

Joint Structures and Common Foundation of Statistical Physics, Information Geometry and Inference for Learning

# Souriau-Casimir Lie Groups Thermodynamics & Machine Learning

SPIGL'2

Grenoble INP UGA

Frédéric BARBARESCO 28/07/2020

THALES

« Tout mathématicien sait qu'il est impossible de comprendre un cours élémentaire en thermodynamique. » Vladimir Arnold



#### **Geometric Theory of Heat:** Gibbs Diagrams sculpted by James Clerk Maxwell (1874)





### Jean-Marie Souriau (1922-2012): New way of thinking Physics



« Il est évident que l'on ne peut définir de valeurs moyennes que sur des objets appartenant à un espace vectoriel (ou affine); donc - si bourbakiste que puisse sembler cette affirmation - que l'on n'observera et ne mesurera de valeurs moyennes que sur des grandeurs appartenant à un ensemble possédant physiquement une structure affine. Il est clair que cette structure est nécessairement unique - sinon les valeurs moyennes ne seraient pas bien définies. » -

Jean-Marie Souriau

« Il n'y a rien de plus dans les théories physiques que les groupes de symétrie si ce n'est la construction mathématique qui permet précisément de montrer qu'il n'y a rien de plus » - Jean-Marie Souriau
 [There is nothing more in physical theories than symmetry groups except the

mathematical construction which allows precisely to show that there is nothing more]



### Gaston Bachelard – Le nouvel esprit scientifique

« La Physique mathématique, en incorporant à sa base la notion de groupe, marque la suprématie rationnelle... Chaque géométrie – et sans doute plus généralement chaque organisation mathématique de l'expérience – est caractérisée par un groupe spécial de transformations.... Le groupe apporte la preuve d'une mathématique fermée sur elle-même. Sa découverte clôt l'ère des conventions, plus ou moins indépendantes, plus ou moins cohérentes » -Gaston Bachelard, Le nouvel esprit scientifique, 1934

« Sous cette aspiration, la physique qui était d'abord une science des "agents" doit devenir une science des "milieux". C'est en s'adressant à des milieux nouveaux que l'on peut espérer pousser la diversification et l'analyse des phénomènes jusqu'à en provoquer la géométrisation fine et complexe, vraiment intrinsèque... Sans doute, la réalité ne nous a pas encore livré tous ses modèles, mais nous savons déjà qu'elle ne peut en posséder un plus grand nombre que celui qui lui est assigné par la théorie mathématique des groupes. » – Guston Bachelard, Etude sur l'Evolution d'un problème de Physique La propagation thermique dans les solides, 1928

http://www.vrin.fr/book.php?title\_url=Etude\_sur\_l\_evolution\_d\_un\_probleme\_de\_physique\_La\_propagation\_thermique\_dans\_les\_solide s\_9782711600434&search\_back=&editor\_back=%&page=2

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TEXTES PHILOSOPHIQUES

GASTON BACHELARD

ÉTUDE SUR L'ÉVOLUTION

D'UN PROBLÈME DE PHYSIQUE

LA PROPAGATION THEEMOUR

V R L N

HALFS

Statistical Mechanics, Lie Group and Cosmology - 1st part: Symplectic Model of Statistical Mechanics Jean-Marie Souriau

Abstract: The classical notion of Gibbs' canonical ensemble is extended to the case of a symplectic manifold on which a Lie group has a symplectic action ("dynamic group"). The rigorous definition given here makes it possible to extend a certain number of classical thermodynamic properties (temperature is here an element of the Lie algebra of the group, heat an element of its dual), notably inequalities of convexity. In the case of non-commutative groups, particular properties appear: the symmetry is spontaneously broken, certain relations of cohomological type are verified in the Lie algebra of the group. Various applications are considered (rotating bodies, covariant or relativistic statistical Mechanics). [These results specify and complement a study published in an earlier work (\*), which will be designated by the initials SSD].

(\*) Souriau, J.-M., Structure des systèmes dynamique. Dunod, collection Dunod Université, Paris 1969. http://www.jmsouriau.com/structure\_des\_systemes\_dynamiques.htm

Souriau, J-M., Mécanique statistique, groupes de Lie et cosmologie, Colloques Internationaux C.N.R.S., n°237 – Géométrie symplectique et physique mathématique, pp.59-113, 1974 English translation by F. Barbaresco: https://www.academia.edu/42630654/Statistical Mechanics Lie Group and Cosmology 1 st par t Symplectic Model of Statistical Mechanics

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# Souriau SSD Chapter IV: Gibbs Equilibrium is not covariant with respect to Dynamic Groups of Physics

#### MÉCANIQUE STATISTIQUE COVARIANTE

Le groupe des translations dans le temps (7.9) est un sous-groupe du groupe de Galilée; mais ce n'est pas un sous-groupe invariant, ainsi que le

montre un calcul trivial. Si un système dynamique est *conservatif* dans un repère d'inertie, il en résulte qu'il peut *ne plus être conservatif dans un autre*. La formulation (17.24) du principe de Gibbs doit donc être élargie, pour devenir compatible avec la relativité galiléenne.

Nous proposons donc le principe suivant :

Si un système dynamique est invariant par un sous-groupe de Lie  $G_{1}^{i}$ du groupe de Galilée, les équilibres naturels du système constituent l'ensemble de Gibbs du groupe dynamique G'.

Soit  $\mathscr{G}'$  l'algèbre de Lie G'; on sait que  $\mathscr{G}$  est une sous-algèbre de Lie de celle de G, notée  $\mathscr{G}$ ; un équilibre du système sera caractérisé par un élément Z de  $\mathscr{G}'$ , donc de  $\mathscr{G}$ ; on pourra écrire

(17.78)

17.79)

6

(17.

 $Z = \begin{bmatrix} j(\omega) & \beta & \gamma \\ 0 & 0 & \varepsilon \\ 0 & 0 & 0 \end{bmatrix}$ 

en utilisant les notations (13.4); Z parcourt l'ensemble  $\Omega$  défini en (16.219); à chaque valeur de Z est associé un élément M du dual  $\mathscr{G}^{**}$  de  $\mathscr{G}^{*}$ , valeur moyenne du moment  $\mu$ ; on peut appliquer les formules (16.219), (16.220), qui généralisent les relations thermodynamiques (17.26), (17.27), (17.28). On voit que c'est Z (17.78) qui généralise la « température »; le théorème d'isothermie (17.32) s'étend immédiatement : l'équilibre d'un système composé de plusieurs parties sans interactions s'obtient en attribuant à chaque composante un équilibre *correspondant à la même valeur de Z*; l'entropie s, le potentiel de Planck z et le moment moyen M sont additifs. W J.M. Souriau, Structure des systèmes dynamiques, Chapitre IV « Mécanique Statistique »



#### Trompette de Souriau

Lorsque le fait qu'on rencontre est en opposition avec une théorie régnante, il faut accepter le fait et abandonner la théorie, alors même que celle-ci, soutenue par de grands noms, est généralement adoptée

- Claude Bernard "Introduction à l'Étude de la Médecine Expérimentale"

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#### Main references for Souriau « Lie Groups Thermodynamics »

SUPPLEMENTO AL NUOVO CIMEN VOLUME IV



N. 1, 1966

1974

Colloques Internationaux C.N.R.S. N° 237 – Géométric symplectique et physique mathématique

Définition covariante des équilibres thermodynamiques.

J.-M. SOURIAU Faculté des Sciences - Marseille

(ricevuto il 5 Novembre 1965)

CONTENTS. — 1. Un problème variationnel. – 2. Mécanique statistique classique. – 3. Equilibres permis par un groupe de Lie. – 4. Exemples. – 5. Localisation de la température vectorielle.

#### MÉCANIQUE STATISTIQUE, GROUPES DE LIE ET COSMOLOGIE

Jean-Marie SOURIAU(I)

Première partie

FORMULATION SYMPLECTIQUE DE LA MECANIQUE STATISTIQUE

#### Référence to Blanc-Lapierre Book in Souriau Book

[7] A. Blanc-Lapierre, P. Casal, and A. Tortrat, Méthodes mathématiques de la mécanique statistique, Masson, Paris, 1959.

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#### Souriau Quinta Essentia (Quinte Essence)

> "Il y a un théorème qui remonte au XXème siècle. Si on prend une orbite coadjointe d'un groupe de Lie, elle est pourvue d'une structure symplectique. Voici un algorithme pour produire des variétés symplectiques : prendre des orbites coadjointes d'un groupe. Donc cela laisse penser que derrière cette structure symplectique de Lagrange, il y avait un groupe caché. Prenons le mouvement classique d'un moment du groupe, alors ce groupe est très «gros» pour avoir tout le système solaire. Mais dans ce groupe est inclus le groupe de Galilée, et tout moment d'un groupe engendre des moments d'un sous-groupe. On va retrouver comme cela les moments du groupe de Galilée, et si on veut de la mécanique relativiste, cela va être celui du groupe de Poincaré. En fait avec le groupe de Galilée, il y a un petit problème, ce ne sont pas les moments du groupe de Galilée qu'on utilise, ce sont les moments d'une extension centrale du groupe de Galilée, qui s'appelle le groupe de Bargmann, et qui est de dimension 11. C'est à cause de cette extension, qu'il y a cette fameuse constante arbitraire figurant dans l'énergie. Par contre quand on fait de la relativité restreinte, on prend le groupe de Poincaré et il n'y a plus de problèmes car parmi les moments il y a la masse et l'énergie c'est mc<sup>2</sup>. Donc le groupe de dimension 11 est un artéfact qui disparait, quand on fait de la relativité restreinte."

#### **SOURIAU: Affine Group and Thermodynamics**



Comment se fait-il qu'un point de vue unitaire, (qui serait nécessairement une véritable Thermodynamique), ne soit pas encore venu couronner le tableau ? Mystère... »

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#### Fundamental Equation of Geometric Thermodynamic: Entropy Function is an Invariant Casimir Function in Coadjoint Representation



#### Lie Groups Thermodynamic Equations and its extension (1/3)



### Lie Groups Thermodynamic Equations and its extension (2/3)

• • •

Entropy Invariance under the action of the Group !  

$$S\left(Ad_{g}^{\#}(Q)\right) = S\left(Q\right)$$

$$Ad_{g}^{\#}(Q) = Ad_{g}^{*}(Q) + \theta\left(g\right)$$
Souriau cocycle
$$ad_{\frac{\partial S}{\partial Q}}^{*}Q + \Theta\left(\frac{\partial S}{\partial Q}\right) = 0$$

$$\Theta(X) = T_{e}\theta\left(X(e)\right)$$

$$\Theta(X) = T_{e}\theta\left(X(e)\right)$$

$$\theta\left(g\right) = Q\left(Ad_{g}(\beta)\right) - Ad_{g}^{*}(Q)$$
Entropy & Poisson Bracket
$$\left\{S, H\right\}_{\tilde{\Theta}}\left(Q\right) = \left\langle Q, \left[\frac{\partial S}{\partial Q}, \frac{\partial H}{\partial Q}\right] \right\rangle + \tilde{\Theta}\left(\frac{\partial S}{\partial Q}, \frac{\partial H}{\partial Q}\right) = 0$$

$$12$$

$$\tilde{\Theta}\left(X, Y\right) = \left\langle \Theta(X), Y \right\rangle = J_{[X,Y]} - \left\{J_{X}, J_{Y}\right\} = -\left\langle d\theta(X), Y \right\rangle$$
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#### Lie Groups Thermodynamic Equations and its extension (3/3)

Entropy  
Production 
$$dS = \tilde{\Theta}_{\beta} \left( \frac{\partial H}{\partial Q}, \beta \right) dt$$
  $2^{nd}$  principle is related to  
positivity of Fisher tensor  $\frac{\partial S}{\partial t} = \tilde{\Theta}_{\beta} \left( \frac{\partial H}{\partial Q}, \beta \right) \ge 0$   
Metric Tensor related to Fisher Metric  $\tilde{\Theta}_{\beta} \left( \frac{\partial H}{\partial Q}, \beta \right) = \tilde{\Theta} \left( \frac{\partial H}{\partial Q}, \beta \right) + \left\langle Q, \left[ \frac{\partial H}{\partial Q}, \beta \right] \right\rangle$   
Time Evolution of Heat  
wrt to Hamiltonian H  $\frac{dQ}{dt} = \{Q, H\}_{\tilde{\Theta}} = ad_{\frac{\partial H}{\partial Q}}^* Q + \Theta \left( \frac{\partial H}{\partial Q} \right)$   
Stochastic  
Equation  $dQ + \left[ ad_{\frac{\partial H}{\partial Q}}^* Q + \Theta \left( \frac{\partial H}{\partial Q} \right) \right] dt + \sum_{i=1}^{N} \left[ ad_{\frac{\partial H_i}{\partial Q}}^* Q + \Theta \left( \frac{\partial H_i}{\partial Q} \right) \right] \circ dW_i(t) = 0$   
Is rescuently 2783\* AUX202

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#### Euler-Poincaré Equation in case of Non-Null Cohomology

$$\frac{dQ}{dt} = ad_{\frac{\partial H}{\partial Q}}^*Q + \Theta\left(\frac{\partial H}{\partial Q}\right)$$

 $= ad_{\partial H}^*$ 

« Ayant eu l'occasion de m'occuper du mouvement de rotation d'un corps solide creux, dont la cavité est remplie de liquide, j'ai été conduit à mettre les équations générales de la mécanique sous une forme que je crois nouvelle et qu'il peut être intéressant de faire connaître » - Henri Poincaré, CRAS, 18 Février 1901

SÉANCE DU LUNDI 18 FÉVRIER 1901,

PRÉSIDENCE DE M. FOUQUÉ.

#### MEMOIRES ET COMMUNICATIONS

DES MEMBRES ET DES CORRESPONDANTS DE L'ACADÉMIE.

MÉCANIQUE RATIONNELLE. — Sur une forme nouvelle des équations de la Mécanique. Note de M. H. POINCABE.

« Ayant eu l'occasion de m'occuper du mouvement de rotation d'un corps solide creux, dont la cavité est remplie de liquide, j'ai été conduit à mettre les équations générales de la Mécanique sous une forme que je crois nouvelle et qu'il peut être intéressant de faire connaître.

$$rac{d}{dt}rac{d\mathrm{T}}{d\eta_s}=\sum c_{ski}rac{d\mathrm{T}}{d\eta_i}\eta_k+\Omega_s.$$

« Elles sont surtout intéressantes dans le cas où U étant nul, T ne dépend que des  $\eta$  » - Henri Poincaré

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$$\left(ad_{\frac{\partial H}{\partial Q}}^{*}\frac{\partial \Phi}{\partial \beta}\right)_{j} + \Theta\left(\frac{\partial H}{\partial Q}\right)_{j} = C_{ij}^{k}ad_{\left(\frac{\partial H}{\partial Q}\right)^{i}}^{*}\left(\frac{\partial G}{\partial \beta}\right)_{j}^{k}$$
  
de Saxcé, G. Euler-Poincaré equation for Lie groups with  
non null symplectic cohomology. Application to the  
mechanics. In GSI 2019. LNCS; Nielsen, F., Barbaresco,  
F., Eds.; Springer: Berlin, Germany, 2019; Volume 11712

**(H)** 

 $+\Theta_{i}$ 

#### Souriau Model Variational Principle : Poincaré-Cartan Integral Invariant on Massieu Characteristic Function

Extension of Poincaré-Cartan Integral Invariant for Souriau Model

$$\omega = \langle Q, (\beta.dt) \rangle - S.dt = (\langle Q, \beta \rangle - S).dt = \Phi(\beta).dt$$

$$g(t) \in G \qquad \beta(t) = g(t)^{-1} \dot{g}(t) \in \mathfrak{g}$$

Variational Model for arbitrary path  $\eta(t)$ 

$$\delta\beta = \dot{\eta} + [\beta, \eta]$$
$$\delta \int_{t_0}^{t_1} \Phi(\beta(t)) dt = 0$$



#### Legendre Transform as Reciprocal Polar with respect to a paraboloid

 $\Phi(\beta) = \langle \beta, Q \rangle - S(Q)$ 

 $\Phi(\beta)$  Reciprocal Polar with respect to the Paraboloid  $Q^2 = 2S(Q)$ 

#### Darboux Lecture on Legendre Transform based on Chasles remark

178. La méthode de Legendre est élégante et irréprochable. Elle consiste à remplacer les variables x, y, z par p, q et

v = px + qy - z,

en considérant v comme une fonction de p et de q; ce qui revient, suivant une remarque de Chasles, à substituer à la surface sa polaire réciproque par rapport au paraboloïde ayant pour équation

 $2z = x^2 + y^2$ .



## THALES

#### Koszul Book on Souriau Work: The Little Green Book



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#### Koszul Book on Souriau Work: The Little Green Book

#### Jean-Louis Koszul · Yiming Zou Introduction to Symplectic Geometry Forewords by Michel Nguiffo Boyom, Frédéric Barbaresco and Charles-Michel Marle

This introductory book offers a unique and unified overview of symplectic geometry, highlighting the differential properties of symplectic manifolds. It consists of six chapters: Some Algebra Basics, Symplectic Manifolds, Cotangent Bundles, Symplectic G-spaces, Poisson Manifolds, and A Graded Case, concluding with a discussion of the differential properties of graded symplectic manifolds of dimensions (o,n). It is a useful reference resource for students and researchers interested in geometry, group theory, analysis and differential equations.

$$\mu : M \longrightarrow \mathfrak{g}^*$$

$$\mu (sx) = s\mu (x) = \operatorname{Ad}^*(s)\mu(x) + \varphi_{\mu}(s), \quad \forall s \in G, x \in M$$

$$c_{\mu}(a,b) = \{ \langle \mu, a \rangle, \langle \mu, b \rangle \} - \langle \mu, [a,b] \rangle = \langle \operatorname{d} \varphi_{\mu}(a), b \rangle, \quad \forall a, b \in \mathfrak{g}$$

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Jean-Louis Koszul Yiming Zou

# Introduction to Symplectic Geometry

$$\begin{split} & \left[ \mu \left[ \mathbf{h} \mathbf{q} \right] = i \mu \left( \mathbf{x} \right) - \mathbf{A} \mathbf{d}^* (\mathbf{s}) \mu (\mathbf{x}) + \mathbf{q}_{\mu} (\mathbf{s}), \quad \forall \ \mathbf{s} \in G, \mathbf{s} \in \mathbf{M} \\ & c_{\mu} \left( a, b \right) = \left\{ \left( \mu, a \right), \left( \mu, b \right) \right\} - \left( \mu, \left( a, b \right) \right) = \left\{ \mathbf{d} \ \mathbf{q}_{\mu} (a), b \right\}, \quad \forall \ a, b \in \mathfrak{g}. \end{split}$$

Science Press
Beijing

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#### **Massieu Potential versus Gibbs Potentials**

# **GIBBS Potential: Free Energy**



#### Joseph Louis François Bertrand gave to François Massieu a bad advice:

<sup>(1)</sup> Dans le mémoire dont un extrait est inséré aux Comptes rendus de l'Académie des sciences du 18 octobre 1869, ainsi que dans la Note additionnelle insérée le 22 novembre suivant, j'avais adopté pour fonction caractéristique  $\frac{H}{T}$ , ou S  $-\frac{U}{T}$ ; c'est d'après les bons conseils de M. Bertrand que j'y ai substitué la fonction H. dont l'emploi réalise quelques simplifications dans les formules.

# **MASSIEU** Potential : characteristic function



[X] Roger Balian, François Massieu and the thermodynamic potentials, Comptes Rendus Physique Volume 18, Issues 9–10, November–December 2017, Pages 526-530 https://www.sciencedirect.com/science/article/pii/\$1631070517300671

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### Preamble: Souriau Lie Groups Thermodynamics

- Lie groups are in common use in robotics, but still seem to be little used in machine learning.
- We present a model from Geometric Mechanics, developed by Jean-Marie Souriau as part of Mechanical Statistics, allowing to define an invariant Fisher-type metric and covariant statistical densities under Lie group action.
- This new approach makes it possible to <u>extend supervised and un-</u> <u>supervised machine learning</u>, jointly:
  - > to elements belonging to a (matrix) Lie group
  - > to elements belonging to a homogeneous manifold on which a group acts transitively.

## Other models are under study also using the theory of representations of Lie groups [see Tojo keynote].

### Preamble: Souriau Lie Group Statistics & Machine Learning

#### It could be applied for Lie Groups Statistical Analysis for:

- > Rigid Objects Trajectories via the SE(3) Lie group
- > Articulated objects via the SO(3) Lie group
- > Moving parts Dynamics via the SU(1,1) Lie group
- The Souriau model makes it possible in particular to define:
  - a Gibbs density of Maximum Entropy on the Lie group coadjoint orbits (in the dual space of their Lie algebra)
  - > with **coadjoint orbits** considered as a homogeneous symplectic manifold.

These densities are parameterized via the Souriau "Moment Map" :

- > map from the symplectic manifold to the dual space of Lie algebra
- > tool geometrizing Noether's theorem
- > on which the group acts via the coadjoint operator

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#### Preamble: New Geometric Entropy Definition

This model is also very useful in control and navigation, because it makes it possible to extend concept of "Gaussian" noises (in the sense of the maximum Entropy) on the Lie algebra.

In this new model, Entropy is defined as an invariant Casimir function in coadjoint representation (this fact gives a natural geometrical definition to Entropy via the structural coefficients).

This Souriau Model of Lie Groups Thermodynamics:

- Is developed in a MDPI "Entropy" Special Issue on "Lie Group Machine Learning and Lie Group Structure Preserving Integrators"
- will be presented at Les Houches SPIGL'20 on "Joint Structures and Commun Foundation of Statistical Physics, Information Geometry and Inference for Learning"

will be presented at IRT SystemX workshop on "Topological and geometric approaches for statistical learning"

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# Motivations for Lie Group Machine Learning



## AI/Machine Learning Evolution: ALGEBRA COMPUTATION STRUCTURES



### **GEOMSTATS: PYTHON Library for Lie Group Machine Learning**



hal-02536154, version 1

ΖJ

Pré-publication, Document de travail 🕄

#### Geomstats https://github.com/geomstats/geomstats

pypi package 2.1.0 build passing codecov (Coverages for: numpy, tensorflow, pytorch)

Geomstats is an open-source Python package for computations and statistics on manifolds. The package is organized into two main modules: geometry and learning.

The module geometry implements concepts in differential geometry, and the module learning implements statistics and learning algorithms for data on manifolds.

To get started with geomstats, see the examples directory.

For more in-depth applications of geomstats, see the applications repository.

- The documentation of geomstats can be found on the documentation website.
- If you find geomstats useful, please kindly cite our paper.

#### Install geomstats via pip3

#### Video: https://m.youtube.com/watch?v=Ju-Wsd84uG0

pip3 install geomstats

https://hal.inria.fr/hal-02536154

#### Geomstats: A Python Package for Riemannian Geometry in Machine Learning

Nina Miolane<sup>1</sup>, Alice Le Brigant, Johan Mathe<sup>2</sup>, Benjamin Hou<sup>3</sup>, Nicolas Guigui<sup>4,5</sup>, Yann Thanwerdas<sup>4,5</sup>, Stefan Hevder<sup>6</sup>, Olivier Peltre, Niklas Koep, Hadi Zaatiti<sup>7</sup>, Hatem Hajri<sup>7</sup>, Yann Cabanes, Thomas Gerald, Paul Chauchat<sup>8</sup>, Christian Shewmake, Bernhard Kainz, Claire Donnat<sup>9</sup>, Susan Holmes<sup>1</sup>, Xavier Pennec<sup>4, 5</sup> Détails



#### Motivation for Lie Group Machine Learning : Data as Lie Groups



### Motivation for Lie Group Machine Learning: Data as Lie Groups



#### Path Signatures on Lie Groups

#### Path Signatures on Lie Groups

#### Darrick Lee

Department of Mathematics University of Pennsylvania Philadelphia, PA 19104, USA

#### **Robert Ghrist**

LDARRICK@SAS.UPENN.EE

GHRIST@SEAS.UPENN.ED

Departments of Mathematics and Electrical & Systems Engineering University of Pennsylvania Philadelphia, PA 19104, USA

Editor:

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#### Abstract

Path signatures are powerful nonparametric tools for time series analysis, shown to form a universal and characteristic feature map for Euclidean valued time series data. We lift the theory of path signatures to the setting of Lie group valued time series, adapting these tools for time series with underlying geometric constraints. We prove that this generalized path signature is universal and characteristic. To demonstrate universality, we analyze the human action recognition problem in computer vision, using SO(3) representations for the time series, providing comparable performance to other shallow learning approaches, while offering an easily interpretable feature set. We also provide a two-sample hypothesis test for Lie group-valued random walks to illustrate its characteristic property. Finally we provide algorithms and a Julia implementation of these methods.

Keywords: path signature, Lie groups, universal and characteristic kernels



Figure 5: Numbering of the primary pairs of body parts

300

250

200

150

100

an Test Distribution M false

0.4

0.4 0.6

MMD

MMD

Euclidean Null Distribution (H, false









Figure 9: The von-Mises Fisher density on  $S^2$  with mean direction x = (0, 0, 1) and  $\kappa = 0.1$ .



### Extension of Mean-Shift for Lie Group (e.g. with SO(3))



#### Motivation for Lie Group Machine Learning: Data in Homogenous Space where a Lie Group acts homogeneously

Poincaré/Hyperbolic Embedding in Poincaré Unit Disk for NLP (Natural Langage Processing)



#### Image Processing by SE(2) Group: hypoelliptic diffusion in SE(2): analogy with Brain Orientation Maps V1

#### Image Processing with oriented gradient by SE(2)

$$\begin{bmatrix} Z \\ 1 \end{bmatrix} = \begin{bmatrix} \Omega & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} Z \\ 1 \end{bmatrix}, \quad \begin{cases} \Omega \in SO(2) \\ t \in R^2 \end{cases}$$

$$Z : \mathbb{R}^2 \to \mathbb{C}, a = \rho e^{i\theta} \mapsto r(a) e^{i\varphi(a)} \\ \omega_S = -\sin(\theta) dx + \cos(\theta) dy \\ \cos(\theta) (dy - pdx) = \cos(\theta) \omega. \end{cases}$$

$$Z : \mathbb{R}^2 \to \mathbb{C}, a = \rho e^{i\theta} \mapsto r(a) e^{i\varphi(a)} \\ \omega_S = -\sin(\theta) dx + \cos(\theta) dy \\ \cos(\theta) (dy - pdx) = \cos(\theta) \omega. \end{cases}$$

$$Z : \mathbb{R}^2 \to \mathbb{R}$$

$$\begin{bmatrix} \nabla \nabla (t) = \frac{1}{2} \arg Z_{exp}(x) \\ \nabla (t) = \frac{1}{2} \arg Z_{exp}(x) \\$$

Plane of the image



SE(2) double covering of  $PTR^2$ 



A scheme of the primary visual cortex V1

(horizontal)

connections among orientation columns belonging to different hypercolumns



### Lie Group Machine Learning for Drone Recognition

#### Drone Recognition on Micro-Doppler by SU(1,1) Lie Group Machine Learning

- Verblunsky/Trench Theorem: all Toeplitz Hermitian Positive Definite Covariance matrices of stationary Radar Time series could be coded and parameterized in a product space with a real positive axis (for signal power) and a Poincaré polydisk (for Doppler Spectrum shape).
- Poincaré Unit Disk is an homogeneous space where SU(1,1) Lie Group acts transitively. Each data in Poincaré unit disk of this polydisk could be then coded by SU(1,1) matrix Lie group element.
- > Micro-Doppler Analysis can be achieved by SU(1,1) Lie Group Machine Learning.

#### Drone Recognition on Kinematics by SE(3) Lie Group Machine Learning

- Trajectories could be coded by SE(3) Lie group time series provided through Invariant Extended Kalman Filter (IEKF) Radar Tracker based on local Frenet-Seret model.
- > Drone kinematics will be then coded by time series of SE(3) matrix Lie Groups characterizing local rotation/translation of Frenet frame along the drone trajectory.

#### Drone Recognition by Lie Group Machine Learning: SU(1,1) & SE(3)



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 $D = \{ z = x + iy \in C / |z| < 1 \}$ 

#### Motivation for Lie Group Machine Learning: Data in Homogenous Space where a Lie Groups act homogeneously

(Micro-)Doppler & Space-Time wave Learning in Poincaré/Siegel Polydisks

$$SU(1,1) = \left\{ \begin{pmatrix} \alpha & \beta \\ \beta^* & \alpha^* \end{pmatrix} / \alpha, \beta \in C, |\alpha|^2 - |\beta|^2 = 1 \right\}$$

$$\varphi: THDP(n) \rightarrow R_+^* \times D^{n-1}$$

$$R_n \mapsto (P_0, \mu_1, \dots, \mu_{n-1})$$

$$R_n = (h, \mu_1, \dots, \mu_{n-1}) \in R^{+*} \times D^{n-1}$$
with  $(D^{n-1} = D \times \dots \times D)$ 

$$ds_{Poincaré}^2 = \frac{|dz|^2}{(1-|z|^2)^2} = (1-zz^*) dz (1-z^*z) dz^*$$

$$\varphi: TBTHPD_{n \times n} \rightarrow THPD_n \times SD^{n-1}$$

$$R \mapsto (R_0, A_1^1, \dots, A_{n-1}^{n-1})$$

$$34 \text{ with } SD = \left\{ Z \in Herm(n) / ZZ^+ < L \right\}$$



F. Barbaresco, Lie Group Machine Learning and Gibbs Density on Poincaré Unit Disk from Souriau Lie Groups Thermodynamics and SU(1,1) Coadjoint Orbits. In: Nielsen, F., Barbaresco, F. (eds.) GSI 2019. LNCS, vol. 11712, SPRINGER, 2019

 $ds_{Siegel}^{2} = Tr \left[ \left( I - ZZ^{+} \right)^{-1} dZ \left( I - Z^{+}Z \right)^{-1} dZ^{+} \right]$ 

#### Matrix Lie Group SU(1,1) for Doppler Data

#### Lie Group Structure for Doppler Data

Lie Group structure appears naturally on Doppler data, if we consider time series of locally stationary signal and their associated covariance matrix. Covariance matrix is Toeplitz Hermitian Positive Definite. We can then use a Theorem due to Verblunsky and Trench, that this structure of covariance matrix could be coded in product space involving the Poincaré unit Polydisk:

$$\varphi: THDP(n) \to R_{+}^{*} \times D^{n-1}$$
$$R_{n} \mapsto (P_{0}, \mu_{1}, ..., \mu_{n-1})$$

> where D is the Poincaré Unit Disk:  $D = \{z = x + iy \in C / |z| < 1\}$ 

### Matrix Lie Group SU(1,1) for Doppler Data

#### Poincaré Unit Disk as a Homogeneous Manifold

The Poincaré unit disk is an homogeneous bounded domain where the Lie Group SU(1,1) act transitively. This Matrix Group is given by

$$SU(1,1) = \left\{ \begin{bmatrix} a & b \\ b^* & a^* \end{bmatrix} / |a|^2 - |b|^2 = 1, \ a, b \in C \right\}$$

$$az + b$$

- > where SU(1,1) acts on the Poincaré Unit Disk by:  $g \in SU(1,1) \Rightarrow g.z = \frac{az + c}{b^*z + a}$
- > with Cartan Decomposition of SU(1,1):

$$\begin{pmatrix} a & b \\ b^* & a^* \end{pmatrix} = |a| \begin{pmatrix} 1 & z \\ z^* & 1 \end{pmatrix} \begin{pmatrix} a/|a| & 0 \\ 0 & a^*/|a| \end{pmatrix}$$

with 
$$z = b(a^*)^{-1}, |a| = (1 - |z|^2)^{-1/2}$$

> We can observe that  $z = b(a^*)^{-1}$  could be considered as action of  $g \in SU(1,1)$ on the centre on the unit disk  $z = g.0 = b(a^*)^{-1}$ .

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### Matrix Lie Group SU(1,1) for Doppler Data

### Coding Doppler Spectrum by data on SU(1,1) Lie group

> The principal idea is that we can code any point  $z = b(a^*)^{-1}$  in the unit disk by an element of the Lie Group SU(1,1). Main advantage is that the point position is no longer coded by coordinates but intrinsically by transformation from 0 to this point. Finally, a covariance matrix of a stationary signal could be coded by (n-1) Matrix SU(1,1) Lie Group elements:

$$\begin{aligned} \text{THPD} &\to R_{+}^{*} \times D^{n-1} &\to R_{+}^{*} \times SU(1,1)^{n-1} \\ R_{n} &\mapsto \left(P_{0}, \mu_{1}, ..., \mu_{n-1}\right) \mapsto \left(P_{0}, \begin{bmatrix} a_{1} & b_{1} \\ b_{1}^{*} & a_{1}^{*} \end{bmatrix}, ..., \begin{bmatrix} a_{n-1} & b_{n-1} \\ b_{n-1}^{*} & a_{n-1}^{*} \end{bmatrix} \end{aligned}$$



### Extension for Space-Time Processing: Siegel Disk



Matrix Extension of Trench/Verblunsky Theorem: Existence of diffeomorphism φ and Siegel Polydisk (matrix extension of Poincaré Disk)

$$\varphi: TBTHPD_{n \times n} \to THPD_n \times SD^{n-1}$$

$$R \mapsto \left(R_0, A_1^1, \dots, A_{n-1}^{n-1}\right)$$
with  $SD = \{Z \in Herm(n) / ZZ^+ < I\}$ 

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$$ds_{Siegel}^{2} = Tr \left[ \left( I - ZZ^{+} \right)^{-1} dZ \left( I - Z^{+}Z \right)^{-1} dZ^{+} \right]$$



### **3D trajectory and Frenet-Serret Frame**

When we consider a 3D trajectory of a mobile target, we can describe this curve by a time evolution of the local Frenet–Serret frame (local frame with tangent vector, normal vector and binormal vector). This frame evolution is described by the Frenet-Serrtet formula that gives the kinematic properties of the target moving along the continuous, differentiable curve in 3D Euclidean space R<sup>3</sup>. More specifically, the formulas describe the derivatives of the so-called tangent, normal, and binormal unit vectors in terms of each other.

$$\frac{d}{dt} \begin{pmatrix} \vec{t} \\ \vec{n} \\ \vec{b} \end{pmatrix} = \begin{bmatrix} 0 & \kappa & 0 \\ -\kappa & 0 & \gamma \\ 0 & -\gamma & 0 \end{bmatrix} \begin{pmatrix} \vec{t} \\ \vec{n} \\ \vec{b} \end{pmatrix} \text{ with } \begin{cases} \kappa : \text{curvature} \\ \gamma : \text{torsion} \end{cases}$$



THALES

### **3D trajectory curve**

> we will consider motions determined by exponentials of paths in the Lie algebra. Such a motion is determined by a unit speed space-curve  $\tau(t)$ . Now in a Frenet-Serret motion a point in the moving body moves along the curve and the coordinate frame in the moving body remains aligned with the tangent  $\vec{t}$ , normal  $\vec{n}$ , and binormal  $\vec{b}$ , of the curve. Using the 4-dimensional representation of the Lie Group SE(3), the motion can be specified as :

$$G(t) = \begin{pmatrix} R(t) & \tau(t) \\ 0 & 1 \end{pmatrix} \in SE(3)$$

> where  $\tau(t)$  is the curve and the rotation matrix has the unit vectors  $\vec{t}$ ,  $\vec{n}$ , and  $\vec{b}$  as columns:

$$R(t) = \begin{pmatrix} \vec{t} & \vec{n} & \vec{b} \end{pmatrix} \in SO(3)$$

### Time evolution of Frenet-Serret Frame

- > If we introduce the Darboux vector  $\vec{\omega} = \gamma \vec{t} + \kappa \vec{b}$  that we can rewritte from Frenet-Serret Formulas :  $\frac{d\vec{t}}{dt} = \vec{\omega} \times \vec{t}$ ,  $\frac{d\vec{n}}{dt} = \vec{\omega} \times \vec{n}$ ,  $\frac{d\vec{b}}{dt} = \vec{\omega} \times \vec{b}$
- Then, we can write with  $\Omega$  is the 3×3 anti-symmetric matrix corresponding to  $\vec{\omega}$ :  $\frac{dR}{dt} = \Omega R$
- > We note that  $\frac{d\tau(t)}{dt} = \vec{t}$  and  $\frac{d\vec{\omega}}{dt} = \frac{d\gamma}{dt}\vec{t} + \frac{d\kappa}{dt}\vec{b}$
- > The instantaneous twist of the motion G(t) is given by:

$$S_d = \frac{dG(t)}{dt}G^{-1}(t) = \begin{pmatrix} \Omega & \upsilon \\ 0 & 0 \end{pmatrix}$$

### Instantaneous twist

> This is the Lie algebra element corresponding to the tangent vector to the curve G(t). It is well known that elements of the Lie algebra se(3) can be described as lines with a pitch. The fixed axode of a motion  $G(t) \in SE(3)$  is given by the axis of  $S_d$  as t varies. The instantaneous twist in the moving reference frame is given by  $S_b = G^{-1}(t)S_dG(t)$ , that is, by the adjoint action on the twist in the fixed frame. The

instantaneous twist  $S_b$  can also be found from the relation:

$$S_{b} = G^{-1}(t) \frac{dG(t)}{dt}$$
$$S_{b} = G^{-1} \frac{dG}{dt} = \begin{pmatrix} R^{T} & -R^{T}\tau \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \Omega R & \vec{t} \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} R^{T}\Omega R & R\vec{t} \\ 0 & 0 \end{pmatrix}$$

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### Trajectory as a time series of Matrix SE(3) Lie groups

We can observe that we could describe a 3D trajectory by a time series of SE(3) Lie group elements:

$$SE(3) = \left\{ \begin{bmatrix} R & \tau \\ 0 & 1 \end{bmatrix} / R \in SO(3), \tau \in R^3 \right\}$$
  
with  $SO(3) = \left\{ R / R^T R = RR^T = I, \det^2 R = 1 \right\}$ 

> Then, the trajectory will be given by the following time series :

$$\left\{ \begin{bmatrix} R_1 & \tau_1 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} R_2 & \tau_2 \\ 0 & 1 \end{bmatrix}, \dots, \begin{bmatrix} R_n & \tau_n \\ 0 & 1 \end{bmatrix} \right\} \in SE(3)^n$$

# THALES

## Lie Group Co-adjoint Orbits & Homogeneous Symplectic Manifold



### **Structuring Principles for Learning : Calculus of Variations**



### SOURIAU 2019

### SOURIAU 2019

- Internet website : <u>http://souriau2019.fr</u>
- In 1969, 50 years ago, Jean-Marie Souriau published the book "Structure des système dynamiques", in which using the ideas of J.L. Lagrange, he formalized the "Geometric Mechanics" in its modern form based on Symplectic Geometry
- Chapter IV was dedicated to "Thermodynamics of Lie groups" (ref André Blanc-Lapierre)
- Testimony of Jean-Pierre Bourguignon at Souriau'19 (IHES, director of the European ERC)



Jean-Marie SOURIAU and Symplectic Geometry

Jean-Pierre BOURGUIGNON (CNRS-IHÉS)

# SOURIAU 2019

Conference May 27-31 2019, Paris-Diderot University

#### https://www.youtube.com/watch?v=beM2pUK1H7o



Frédéric Barbaresco **Deniel Benneguin** Jean-Pierre Bourguignor Plant Cartist Dan Christenser Maurice Courtrage Thibeuit Denour Paul Donato Paolo Giordano Seing Gürer Patrick (glenne-Zerminous theil Camboo Jean-Fierte Mathot Yeattle Kosman-Schwartzbach Marc Lachiete-Res Martin Processed Elisti Prato Un Schweiber lash, larings Szone injurt Robarid Triay Jordan Watte Emin'Wu San Wil Moter Alan Weinsteir

#### JEAN-MARIE SOURIAU

In 1969, the groundbreaking book of Jean-Marie Souriau appeared "Structure des Systèmes Dynamiques". We will celebrate, in 2019, the jubilee of its publication, with a conference in honour of the work of this great scientist.

Symplectic Mechanics, Geometric Quantization, Relativity, Thermodynamics, Cosmology, Diffeology & Philosophy



SOHERF

PARIS DIDEROT

### Le Livre de J.M. Souriau « Structure des systèmes dynamiques », 1969



#### http://www.jmsouriau.com/structure\_des\_systemes\_dynamiques.htm http://www.springer.com/us/book/9780817636951

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THALES

### Lagrange 2-form rediscovered by Jean-Marie Souriau

> Rewriting equations of classical mechanics in phase space

$$m\frac{d^2r}{dt^2} = F$$
  $m\frac{dv}{dt} = F$  et  $v = \frac{dr}{dt}$ 

- Souriau rediscovered that Lagrange had considered the evolution space:  $y = \begin{bmatrix} r \\ v \end{bmatrix} \in V$  $\begin{cases} m\delta v - F\delta t = 0 \\ \delta r - v\delta t = 0 \end{cases}$
- > A dynamic system is represented by a foliation. This foliation is determined by an antisymmetric covariant 2<sup>nd</sup> order tensor  $\sigma$ , called the Lagrange (-Souriau) form, a bilinear operator on the tangent vectors of V.

$$\sigma(\delta y)(\delta' y) = \langle m\delta v - F\delta t, \delta' r - v\delta' t \rangle - \langle m\delta' v - F\delta' t, \delta r - v\delta t \rangle \quad \delta y = \begin{pmatrix} \delta t \\ \delta r \\ \delta v \end{pmatrix} \text{ et } \delta' y = \begin{pmatrix} \delta' t \\ \delta' r \\ \delta' v \end{pmatrix}$$

> In the Lagrange-Souriau model,  $\sigma$  is a 2-form on the evolution space V, and the differential equation of motion implies:  $\delta y \in \varepsilon$ 

 $\sigma(\delta y)(\delta' y) = 0, \forall \delta' y$ int Structures and Common Foundations of Statistical Physics,

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 $\sigma(\delta y) = 0$  ou  $\delta y \in \ker(\sigma)$ 

### THALES

### **Evolution space of Lagrange-Souriau**





### Gallileo Group & Algebra & V. Bargman Central extensions

Symplectic cocycles of the Galilean group: V. Bargmann (Ann. Math. 59, 1954, pp 1–46) has proven that the symplectic cohomology space of the Galilean group is one-dimensional.



### Souriau Work Roots: François Gallissot Theorem

> Gallissot Theorem: There are 3 types of differential forms generating the equations of a material point motion, **invariant by the action of the Galileo group** 

$$A: \begin{cases} s = \frac{1}{2m} \sum_{i=1}^{3} (mdv_i - F_i dt) \\ e = \frac{m}{2} \sum_{j=1}^{3} (dx_j - v_j dt)^2 \end{cases}$$

F. GALLISSOT, Les formes extérieures en Mécanique (Thèse), Durand, Chartres, 1954.

$$B: f = \sum_{1}^{3} \delta_{ij} (dx_i - v_i dt) (m dv_j - F_j dt) \text{ with } \delta_{ij} \text{ krönecker symbol}$$

C: 
$$\omega = \sum_{i=1}^{3} \delta_{ij} (mdv_i - F_i dt) \wedge (dx_j - v_j dt)$$

- >  $d\omega = 0$  constrained the Pfaff form  $\delta_{ii}F_i dx_i$  to be closed and to be reduced to the differential of  $U: C \Rightarrow \omega = m\delta_{ij}dv_i \wedge dx_j - dH \wedge dt$  with H = T - U and  $T = 1/2 \sum m(v_i)^2$
- > It proves that  $\omega$  has an exterior differential  $d\omega$  generating Poincaré-Cartan Integral invariant:  $d\omega = \sum mv_i dx_j - \omega dx_j$ Les Houches 27th-31st July 2020

Information Geometry and Inference for Learning (SPIGL'20)

### François Gallissot Work in 1952 based on Elie and Henri Cartan works

S'affranchir

de la servitude

des coordonnées

 $i(\mathbf{E})\Omega = \mathbf{o}$ 

#### LES FORMES EXTÉRIEURES EN MÉCANIQUE

par F. GALLISSOT.

#### 1952

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#### INTRODUCTION

La mécanique des systèmes paramétriques développée traditionnellement d'après les idées de Lagrange s'est toujours heurtée à des difficultés notables lorsqu'elle a désiré aborder les questions de frottement entre solides (impossibilité et indétermination) ou la notion générale de liaison (asservissement de M. Béghin), d'autre part la forme lagrangienne des équations du mouvement ne nous donne aucune indication sur la nature du problème de l'intégration.

Dans ces célèbres leçons sur les invariants intégraux Élie Cartan a montré que toutes les propriétés des équations différentielles de la dynamique des sytèmes holonomes résultaient de l'existence de l'invariant intégral  $\int \omega, \ \omega = p_i dq^i - H dt$ . Ainsi à tout système holonome dont les forces dérivent d'une fonction de forces est associé une forme  $\omega$ , les équations du mouvement étant les caractéristiques de la forme extérieure  $d\omega$ . Au cours de ces dix dernières années, sous l'influence des topologistes s'est édifiée sur des bases qui semblent définitives la théorie des formes extérieures sur les variétés différentiables. Il est alors naturel de se demander si la mécanique classique ne peut pas bénéficier largement de ce courant d'idées, si elle ne peut pas être construite en plaçant à sa base une forme extérieure de degré deux, si grâce à la notion de variétés, la notion de liaison ne peut pas être envisagée sous un angle plus intelligible, si les indéterminations et impossibilités qui paraissent paradoxales dans le cadre lagrangien n'ont pas une explication naturelle, enfin s'il n'est pas possible de considérer sous un jour nouveau le problème de l'intégration des équations du mouvement, ces dernières étant engendrées par une forme  $\Omega$  de degré deux.

Pour atteindre ces divers objectifs il m'a semblé utile de reprendre dans le chapitre 1 l'étude des bases logiques sur lesquelles est édifiée la mécanique galiléenne. Je montre ainsi dans le § 1 que lorsqu'on se propose de trouver des formes génératrices des équations du mouvement d'un point matériel invariantes dans les transformations du groupe galiléen, la forme la plus intéressante est une forme extérieure de degré deux définie sur une variété  $V_{.} = E_{.} \otimes E \otimes .T$ (E, espace euclidien, T droite numérique temporelle) (<sup>1</sup>). Dans le § 11 on montre qu'à tout système paramétrique holonome à n degrés de liberté est associé une forme  $\Omega$  de degré deux de rang 2n définie sur une variété différentiable dont les caractéristiques sont les équations du mouvement (2). Cette forme s'exprime si l'on veut au moyen de 2n formes de Pfaff et de dt, la forme hamiltonienne n'étant qu'un cas particulier simple. Dans le § 3 j'indique sommairement comment on peut s'affranchir de la servitude des coordonnées dans l'étude des systèmes dynamiques et le rôle important joué par l'opérateur i() antidérivation de M. H. Cartan (3), le champ caractéristique E de la forme  $\Omega$  étant défini par la relation  $i(\mathbf{E})\Omega = \mathbf{0}$ .

(1) M. KRAVTCHENKO a présenté cette conception au VIII<sup>e</sup> Congrès de Mécanique.

(\*) Dès 1946 M. LICHNEROWICZ au Bulletin des Sciences Mathémaliques tome LXX, p. 90 a déjà introduit les formes extérieures pour la formation des équations des systèmes holonomes et linéairement non holonomes.

(3) M. H. CARTAN, Colloque de Topologie, Bruxelles, 1950. Masson, Paris, 1951.

F. GALLISSOT, Les formes extérieures en Mécanique (Thèse), Durand, Chartres, 1954.



### Interior/Exterior Products and Lie derivative

 $i_{V}\omega$  is the (p-1)-form on X obtained by inserting V(x) as the first argument of  $\omega$  :

Interior product : 
$$i_V \omega(v_2, \cdots v_p) = \omega(V(x), v_2, \cdots, v_p)$$

▶  $\theta \land \omega$  is the (p + 1)-form on X where  $\omega$  is a p-form and  $\theta$  is a 1-form on X:

Exterior product: 
$$\theta \wedge \omega(v_0, \dots, v_p) = \sum_{i=0}^p (-1)^i \theta(v_i) \omega(v_0, \dots, \hat{v}_i, \dots, v_p)$$

(where the hat indicates a term to be omitted).

>  $L_V \omega$  is a p-form on X , and  $L_V \omega = 0$  if the flow of V consists of symmetries of  $\omega$ :

Lie derivative : 
$$L_V \omega(v_1, \dots, v_p) = \frac{d}{dt} e^{tV^*} \omega(v_1, \dots, v_p) \Big|_{t=0}$$

### Exterior derivative and E.Cartan, H. Cartan & S. Lie formulas

>  $d\omega$  is the (p+1)-form on X defined by taking the ordinary derivative of  $\omega$  and then antisymmetrizing:

**Exterior derivative** : 
$$d\omega(v_0, \dots, v_p) = \sum_{i=0}^p (-1)^i \frac{\partial \omega}{\partial x}(v_i)(v_0, \dots, \hat{v}_i, \dots, v_p)$$
  
 $p = 0, [d\omega]_i = \partial_i \omega \; ; \; p = 1, [d\omega]_{ij} = \partial_i \omega_j - \partial_j \omega_i \; ; \; p = 2, [d\omega]_{ijk} = \partial_i \omega_{jk} + \partial_j \omega_{ki} + \partial_k \omega_{ij}$   
> The properties of the exterior and Lie Derivative are the following:

$$L_V \omega = di_V \omega + i_V d\omega$$
 (E. Cartan)

$$\dot{i}_{[U,V]}\omega = i_V L_U \omega - L_U i_V \omega$$
 (H.Cartan)

$$L_{[U,V]}\omega = L_V L_U \omega - L_U L_V \omega \quad (S. Lie)$$

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THALES

### Souriau Moment Map (1/2)

- > Let  $(X, \sigma)$  be a connected symplectic manifold.
- > A vector field  $\eta$  on X is called symplectic if its flow preserves the 2-form :  $L_n \sigma = 0$

> If we use Elie Cartan's formula, we can deduce that :

$$L_{\eta}\sigma = di_{\eta}\sigma + i_{\eta}d\sigma = 0$$

> but as  $d\sigma = 0$  then  $di_{\eta}\sigma = 0$ . We observe that the 1-form  $i_{\eta}\sigma$  is closed. > When this 1-form is exact, there is a smooth function  $x \mapsto H$  on X with:

$$i_{\eta}\sigma = -dH$$

> This vector field  $\eta$  is called Hamiltonian and could be define as s symplectic gradient:

$$\eta = \nabla_{Symp} H$$

### Souriau Moment Map (2/2)

$$di_{\eta}\sigma = 0 \qquad \qquad i_{\eta}\sigma = -dH$$

> We define the Poisson bracket of two functions H ,  $H^{\,\prime}$  by :

$$\{H, H'\} = \sigma(\eta, \eta') = \sigma(\nabla_{Symp}H', \nabla_{Symp}H)$$

with 
$$i_\eta \sigma = -dH$$
 and  $i_{\eta'} \sigma = -dH'$ 

- > Let a Lie group  $\,G\,$  that acts on X and that also preserve  $\sigma$  .
- A moment map exists if these infinitesimal generators are actually hamiltonian, so that a map exists:

: 
$$X \to \mathfrak{g}^*$$
 with  $i_{Z_X} \sigma = -dH_Z$  where  $H_Z = \langle \Phi(x), Z \rangle$ 

### Souriau Model of Lie Groups Thermodynamics

- Souriau Geometric (Planck) Temperature is an element of Lie Algebra of Dynamical Group (Galileo/Poincaré groups) acting on the system
- Generalized Entropy is Legendre Transform of minus logarithm of Laplace Transform
- Fisher(-Souriau) Metric is a Geometric Calorific Capacity (hessian of Massieu Potential)
- Higher Order Souriau Lie Groups Thermodynamics is given by Günther's Poly-Symplectic Model (vector-valued model in non-equivariant case)





Souriau formalism is fully covariant, with no special coordinates (covariance of Gibbs density wrt Dynamical Groups)



### Lie Groups Tools Development: From Group to Co-adjoint Orbits



#### Lie Group & Statistical Physics

Jean-Michel Bismut – Random Mechanics Jean-Marie Souriau – Lie Group Thermodynamics, Souriau Metric Jean-Louis Koszul – Affine Lie Group & Algebra representation

#### Harmonic Analysis on Lie Group & Orbits Method

*Pierre Torasso & Michèle Vergne* – Poisson-Plancherel Formula *Michel Duflo* – Extension of Orbits Method, Plancherel & Character *Alexandre Kirillov* – Coadjoint Orbits, Kirillov Character *Jacques Dixmier* – Unitary representation of nilpotent Group

#### Lie Group Representation

**Bertram Kostant** – KKS 2-form, Geometric Quantization **Alexandre Kirillov** – Representation Theory, KKS 2-form **Jean-Marie Souriau** – Moment Map, KKS 2-form, Souriau Cocycle **Valentine Bargmann** – Unitary representation, Central extension

#### **Lie Group Classification**

*Carl-Ludwig Siegel* – Symplectic Group *Hermann Weyl* – Conformal Geometry, Symplectic Group *Elie Cartan* – Lie algebra classification, Symmetric Spaces *Willem Killing* – Cartan-Killing form, Killing Vectors

#### **Group/Lie Group Foundation**

Henri Poincaré – Fuchsian Groups Felix Klein – Erlangen Program (Homogeneous Manifold) Sophus Lie – Lie Group Evariste Galois/Louis Joseph Lagange – Substitution Group

### Lie Group

### **GROUP (Mathematics)**

A set equipped with a binary operation with 4 axioms:

- > Closure  $\forall a, b \in G$  then  $a \bullet b \in G$
- Associativity  $\forall a, b, c \in G$  then  $(a \bullet b) \bullet c = a \bullet (b \bullet c)$
- > Identity  $\exists e \in G \text{ such that } e \bullet a = a \bullet e = a$
- > invertibility  $\forall a \in G, \exists b \in G \text{ such that } b \bullet a = a \bullet b = e$



### LIE GROUP

- A group that is a differentiable manifold, with the property that the group operations of multiplication and inversion are smooth maps:
   ∀x, y ∈ G then φ: G×G → G then φ(x, y) = x<sup>-1</sup>y is smooth
- ▶ A Lie algebra  $\mathfrak{g} = T_e G$  is a vector space with a binary operation called the Lie bracket  $[.,.]:\mathfrak{g} \times \mathfrak{g} \to \mathfrak{g}$  that satisfies axioms: [ax+by,z]=a[x,z]+b[y,z]; [x,x]=0; [x,y]=-[y,x]

Jacobi Identity: [x, [y, z]] + [z, [x, y]] + [y, [z, x]] = 0

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Jant Structurel and Common Foundation of Statinical Physics Info X grioy Comment X and Intere XX or FOI no Matrix Lie Group



### **Lie Group Notation**

### Lie Algebra of Lie Group and Adjoint operators

> Let G a Lie Group and  $T_eG$  tangent space of G at its neutral element e

- 
$$Ad$$
Adjoint representation of  $G$   
 $Ad: G \rightarrow GL(T_eG)$  with  $i_g: h \mapsto ghg^{-1}$   
 $g \in G \mapsto Ad_g = T_e i_g$ 

- ad Tangent application of Ad at neutral element e of G  $ad = T_eAd : T_eG \rightarrow End(T_eG)$   $X, Y \in T_eG \mapsto ad_X(Y) = [X, Y]$ > For  $G = GL_n(K)$  with K = R or C  $T_eG = M_n(K)$   $X \in M_n(K), g \in G$   $Ad_g(X) = gXg^{-1}$   $X, Y \in M_n(K)$   $ad_X(Y) = (T_eAd)_X(Y) = XY - YX = [X, Y]$ - Curve from  $e = I_d = c(0)$  tangent to X = c(1):  $c(t) = \exp(tX)$ and transform by  $Ad : \gamma(t) = Ad \exp(tX)$ 

$$ad_X(Y) = (T_eAd)_X(Y) = \frac{d}{dt}\gamma(t)Y\Big|_{t=0} = \frac{d}{dt}\exp(tX)Y\exp(tX)^{-1}\Big|_{t=0} = XY - YX$$

### Coadjoint operator and Coadjoint Orbits (Kirillov Representation)

### Lie Group Adjoint Representation

> the adjoint representation of a Lie group  $Ad_g$  is a way of representing its elements as linear transformations of the Lie algebra, considered as a vector space

$$Ad_{g} = \left(d\Psi_{g}\right)_{e} : \mathfrak{g} \to \mathfrak{g}$$
$$X \mapsto Ad_{g}(X) = gXg^{-1}$$

$$\Psi: G \to Aut(G)$$

$$g \mapsto \Psi_g(h) = ghg^{-1}$$

$$ad = T_eAd: T_eG \to End(T_eG)$$

$$X, Y \in T_eG \mapsto ad_X(Y) = [X, Y]$$

### Lie Group Co-Adjoint Representation

> the coadjoint representation of a Lie group  $Ad_g^*$ , is the dual of the adjoint representation ( $\mathfrak{g}^*$  denotes the dual space to  $\mathfrak{g}^*$ ):

$$\forall g \in G, Y \in \mathfrak{g}, F \in \mathfrak{g}^*, \text{ then } \left\langle Ad_g^*F, Y \right\rangle = \left\langle F, Ad_{g^{-1}}Y \right\rangle$$

$$K = Ad_{g}^{*} = \left(Ad_{g^{-1}}\right)^{*} \text{ and } K_{*}(X) = -\left(ad_{X}\right)^{*}$$

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### Coadjoint operator and Coadjoint Orbits (Kirillov Representation)

### Co-adjoint Orbits as Homogeneous Symplectic Manifold by KKS 2-form

 $O_F = \{Ad_g^*F, g \in G\}$  subset of  $\mathfrak{g}^*, F \in \mathfrak{g}^*$ > A coadjoint orbit: carry a natural homogeneous symplectic structure by a closed G-invariant 2-form:

$$\sigma_{\Omega}(K_{*X}F,K_{*Y}F) = B_F(X,Y) = \langle F, [X,Y] \rangle, X, Y \in \mathfrak{g}$$

- > The coadjoint action on  $O_F$  is a Hamiltonian G-action with moment map  $\Omega \rightarrow \mathfrak{g}^*$
- Souriau Foundamental Theorem « Every symplectic manifold is a coadjoint orbit » is based on classification of symplectic homogeneous Lie group actions by Souriau, Kostant and Kirillov

$$g \in G$$

Lie Group

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$$O_F = \left\{ Ad_g^*F, g \in G, F \in \mathfrak{g}^* \right\}$$

**Coadjoint Orbit** (action of Lie Group on dual Lie algebra)  $\sigma_{\Omega}(ad_{F}X, ad_{F}Y) = \langle F, [X, Y] \rangle$ 

 $X, Y \in \mathfrak{a}, F \in \mathfrak{a}^*$ 

### Homogeneous Symplectic Manifold

(a smooth manifold with a closed differential 2-form  $\sigma$ , such that  $d\sigma=0$ , where the Lie Group acts transitively)

# THALES

## Fisher-Koszul-Souriau Metric and Geometric Structures of Inference and Learning



### **Elementary Structures of Information Geometry**

= L ( ( x) gjdx'd ( x)

Geometry of Homogeneous RiemanBiounded Domains Complex Geometry Geometry

**Symplectic** 

Information Geometry

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 $\omega = ig_{\alpha\beta}dz^{\alpha} \wedge dz^{\beta}$ 

 $d\omega = 0$ 

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 $g_{\alpha\beta} = \partial \partial^* U$ 

#### Seminal work of Elie Cartan

**Geometry of Jean-Marie Souriau** Study of homogeneous symplectic manifolds geometry with the action of dynamical groups. Introduction of the Lagrange-Souriau 2-form and Lie Groups Thermodynamics.

**Geometry of Jean-Louis Koszul** Study of homogeneous bounded domains geometry, symmetric homogeneous spaces and sharp convex cones. Introduction of an invariant 2-form.

**Geometry of Erich Kähler** Study of differential manifolds geometry equipped with a unitary structure satisfying a condition of integrability. The homogeneous Kähler case studied by André Lichnerowicz.



### Fisher Metric and Fréchet-Darmois (Cramer-Rao) Bound

Cramer-Rao –Fréchet-Darmois Bound has been introduced by Fréchet in 1939 and by Rao in 1945 as inverse of the Fisher Information Matrix:  $I(\theta)$ 

$$R_{\hat{\theta}} = E\left[\left(\theta - \hat{\theta}\right)\left(\theta - \hat{\theta}\right)^{+}\right] \ge I(\theta)^{-1} \qquad [I(\theta)]_{i,j} = -E\left|\frac{\partial^{2}\log p_{\theta}(z)}{\partial \theta_{i}\partial \theta_{i}^{*}}\right|$$

Rao has proposed to introduced an invariant metric in parameter space of density of probabilities (axiomatised by N. Chentsov):  $ds_{\theta}^{2} = Kullback \_Divergence(p_{\theta}(z), p_{\theta+d\theta}(z))$ 

$$ds_{\theta}^{2} = -\int p_{\theta}(z) \log \frac{p_{\theta+d\theta}(z)}{p_{\theta}(z)} dz$$

 $w = W(\theta)$  $\Rightarrow ds_w^2 = ds_\theta^2$ 

$$ds_{\theta}^{2} \underset{Taylor}{\approx} \sum_{\substack{j:j \neq 0 \\ \text{Joint Structures and Common Foundations of Statistical Physics,}}} g_{ij} d\theta_{i} d\theta_{j}^{*} = \sum_{i,j} [I(\theta)]_{i,j} d\theta_{i} d\theta_{j}^{*} = d\theta^{+} . I(\theta) . d\theta$$

### **Distance Between Gaussian Density with Fisher Metric**

### Fisher Matrix for Gaussian Densities:

$$I(\theta) = \begin{bmatrix} \frac{1}{\sigma^2} & 0\\ 0 & \frac{2}{\sigma^2} \end{bmatrix} \text{ avec } E\left[\left(\theta - \hat{\theta}\right)\left(\theta - \hat{\theta}\right)^T\right] \ge I(\theta)^{-1} \text{ et } \theta = \begin{pmatrix} m\\ \sigma \end{pmatrix}$$

> Fisher matrix induced the following differential metric :

$$ds^{2} = d\theta^{T} I(\theta) . d\theta = \frac{dm^{2}}{\sigma^{2}} + 2 . \frac{d\sigma^{2}}{\sigma^{2}} = \frac{2}{\sigma^{2}} \left[ \left( \frac{dm}{\sqrt{2}} \right)^{2} + \left( d\sigma \right)^{2} \right]$$

Poincaré Model of upper half-plane and unit disk





### 1 monovariate gaussian = 1 point in Poincaré unit disk



### Machine Learning & Gradient descent

### Gradient descent for Learning

- Information geometry has been derived from invariant geometrical structure involved in statistical inference. The Fisher metric defines a Riemannian metric as the Hessian of two dual potential functions, linked to dually coupled affine connections in a manifold of probability distributions. With the Souriau model, this structure is extended preserving the Legendre transform between two dual potential function parametrized in Lie algebra of the group acting transitively on the homogeneous manifold.
- > Classically, to optimize the parameter  $\theta$  of a probabilistic model, based on a sequence of observations  $y_t$ , is an online gradient descent with learning rate  $\eta_t$ , and the loss function  $l_t = -\log p(y_t / \hat{y}_t)$ :

$$\theta_{t} \leftarrow \theta_{t-1} - \eta_{t} \frac{\partial l_{t} \left(y_{t}\right)^{T}}{\partial \theta}$$

### **Information Geometry & Machine Learning**

### Information Geometry & Natural Gradient

> This simple gradient descent has a first drawback of using the same non-adaptive learning rate for all parameter components, and a second drawback of non invariance with respect to parameter re-encoding inducing different learning rates. S.I. Amari has introduced the **natural gradient** to preserve this invariance to be insensitive to the characteristic scale of each parameter direction. The gradient descent could be corrected by  $I(\theta)^{-1}$  where I is the Fisher information matrix with respect to parameter  $\theta$ , given by:

$$I(\theta) = \begin{bmatrix} g_{ij} \end{bmatrix}$$
  
with  $g_{ij} = \begin{bmatrix} -E_{y \approx p(y/\theta)} \begin{bmatrix} \frac{\partial^2 \log p(y/\theta)}{\partial \theta_i \partial \theta_j} \end{bmatrix}_{ij} = \begin{bmatrix} E_{y \approx p(y/\theta)} \begin{bmatrix} \frac{\partial \log p(y/\theta)}{\partial \theta_i} & \frac{\partial \log p(y/\theta)}{\partial \theta_j} \end{bmatrix}_{ij}$ 

### THALES

### Natural Gradient & Stochastic Gradient: Natural Langevin Dynamics

# Natural Langevin Dynamics: Natural Gradient with Langevin Stochastics descent

- To regularize solution and avoid over-fitting, Stochastic gradient is used, as Langevin Stochastic Gradients
- Yann Ollivier (FACEBOOK FAIR, previously CNRS LRI Orsay) and Gaëtan Marceau-Caron (MILA, previously CNRS LRI Orsay and THALES LAS/ATM & TRT PhD) have proposed to coupled Natural Gradient with Langevin Dynamics: Natural Langevin Dynamics (Best SMF/SEE GSI'17 paper)

$$\theta_{t} \leftarrow \theta_{t-1} - \eta_{t} I(\theta_{t-1})^{-1} \frac{\partial \left( l_{t} \left( y_{t} \right)^{T} - \frac{1}{N} \log \alpha \left( \theta_{t-1} \right) \right)}{\partial \theta} + \sqrt{\frac{2\eta_{t}}{N}} I(\theta_{t-1})^{-1/2} N(0, I_{d})$$

The resulting natural Langevin dynamics combines the advantages of Amari's natural gradient descent and Fisher-preconditioned Langevin dynamics for large neural networks

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### **Dual Entropic Natural Gradient**

We can define a natural gradient with dual potential given by Shannon Entropy H (Legendre transform of characteristic fonction G, logarithm of partitiun function).

$$\theta_{t} \leftarrow \theta_{t-1} - \eta_{t} I(\theta_{t-1})^{-1} \frac{\partial l_{t} \left( y_{t} / \theta \right)^{T}}{\partial \theta}$$

$$\theta = \nabla S(\eta) = h(\eta) \qquad \eta = \nabla \Phi(\theta) = g(\theta) \qquad S(\eta) = \sup_{\theta \in \Theta} \left\{ \langle \theta, \eta \rangle - \Phi(\theta) \right\}$$

$$\eta_t \leftarrow \eta_{t-1} - \alpha_t \frac{\partial l_t \left( y_t / h(\eta_{t-1}) \right)^T}{\partial \theta} \qquad S(\eta) = \sum_{\theta \in \Theta} \left\{ \langle \theta, \eta \rangle - \Phi(\theta) \right\}$$

$$\nabla_\eta l_i \left( h(\eta) \right) = \nabla_\eta h(\eta) \nabla_\theta l_t \left( h(\eta) \right) \Rightarrow \nabla_\theta l_t \left( h(\eta) \right) = \left[ \nabla_\eta h(\eta) \right]^{-1} \nabla_\eta l_i \left( h(\eta) \right)$$

$$\eta_t \leftarrow \eta_{t-1} - \alpha_t \left[ \nabla^2 S(\eta_{t-1}) \right]^{-1} \frac{\partial l_t \left( y_t / h(\eta_{t-1}) \right)^T}{\partial \eta} \qquad \text{Natural Dual Entropic Gradient}$$

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### Information Geometry & Machine Learning : Legendre structure

Legendre Transform, Dual Potentials & Fisher Metric

S.I. Amari has proved that the Riemannian metric in an exponential family is the Fisher information matrix defined by:

$$g_{ij} = -\left[\frac{\partial^2 \Phi}{\partial \theta_i \partial \theta_j}\right]_{ij}$$
 with  $\Phi(\theta) = -\log \int_R e^{-\langle \theta, y \rangle} dy$ 

> and the dual potential, the Shannon entropy, is given by the Legendre transform:

$$S(\eta) = \langle \theta, \eta \rangle - \Phi(\theta)$$
 with  $\eta_i = \frac{\partial \Phi(\theta)}{\partial \theta_i}$  and  $\theta_i = \frac{\partial S(\eta)}{\partial \eta_i}$ 



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## Fisher Metric and Koszul 2 form on sharp convex cones

#### Koszul-Vinberg Characteristic Function, Koszul Forms

> J.L. Koszul and E. Vinberg have introduced an affinely invariant Hessian metric on a sharp convex cone through its characteristic function

 $\Phi_{\Omega}(\theta) = -\log \int_{\mathbb{R}^{*}} e^{-\langle \theta, y \rangle} dy = -\log \psi_{\Omega}(\theta) \text{ with } \theta \in \Omega \text{ sharp convex cone}$ 

 $\psi_{\Omega}(\theta) = \int e^{-\langle \theta, y \rangle} dy$  with Koszul-Vinberg Characteristic function

- > 1<sup>st</sup> Koszul form  $\alpha$ :  $\alpha = d\Phi_{\Omega}(\theta) = -d\log\psi_{\Omega}(\theta)$
- For KOSZULION *u*, *u*2<sup>nd</sup> Koszul form *γ*:  $\gamma = D\alpha = Dd \log \psi_{\Omega}(\theta)$   $(Dd \log \psi_{\Omega}(x))(u) = \frac{1}{\psi_{\Omega}(u)^{2}} \left[ \int_{\Omega^{*}} F(\xi)^{2} d\xi \int_{\Omega^{*}} G(\xi)^{2} d\xi \left( \int_{\Omega^{*}} F(\xi) \cdot G(\xi) d\xi \right)^{2} \right] > 0$  with  $F(\xi) = e^{-\frac{1}{2}\langle x,\xi \rangle}$  and  $G(\xi) = e^{-\frac{1}{2}\langle x,\xi \rangle} \langle u,\xi \rangle$ > Diffeomorphism:  $\eta = \alpha = -d \log \psi_{\Omega}(\theta) = \int_{\Omega^{*}} \xi p_{\theta}(\xi) d\xi$  with  $p_{\theta}(\xi) = \frac{e^{-\langle \xi,\theta \rangle}}{\int_{\Omega^{*}} e^{-\langle \xi,\theta \rangle} d\xi}$ Jean-Louis Koszul Legendre transform:  $S_{\Omega}(\eta) = \langle \theta, \eta \rangle - \Phi_{\Omega}(\theta)$  with  $\eta = d\Phi_{\Omega}(\theta)$  and  $\theta = dS_{\Omega}(\eta)$

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## Fisher Metric and Souriau 2-form: Lie Groups Thermodyamics

#### Statistical Mechanics, Dual Potentials & Fisher Metric

In geometric statistical mechanics, J.M. Souriau has developed a "Lie groups thermodynamics" of dynamical systems where the (maximum entropy) Gibbs density is covariant with respect to the action of the Lie group. In the Souriau model, previous structures of information geometry are preserved:

$$I(\beta) = -\frac{\partial^2 \Phi}{\partial \beta^2} \text{ with } \Phi(\beta) = -\log \int_M e^{-\langle \beta, U(\xi) \rangle} d\lambda \qquad U: M \to \mathfrak{g}^*$$

$$S(Q) = \langle \beta, Q \rangle - \Phi(\beta) \text{ with } Q = \frac{\partial \Phi(\beta)}{\partial \beta} \in \mathfrak{g}^* \text{ and } \beta = \frac{\partial S(Q)}{\partial Q} \in \mathfrak{g}$$



Jean-Marie Souriau

> In the Souriau Lie groups thermodynamics model,  $\beta$  is a "geometric" (Planck) temperature, element of Lie algebra  $\mathfrak{g}$  of the group, and Q is a "geometric" heat, element of dual Lie algebra  $\mathfrak{g}^*$  of the group.



## Fisher-Souriau Metric and its invariance

#### Statistical Mechanics & Invariant Souriau-Fisher Metric

> In Souriau's Lie groups thermodynamics, the invariance by re-parameterization in information geometry has been replaced by invariance with respect to the action of the group. When an element of the group g acts on the element  $\beta \in \mathfrak{g}$  of the Lie algebra, given by adjoint operator  $Ad_g$ . Under the action of the group g and the Fisher metric  $I(\beta)$  are invariant:

$$\beta \in \mathfrak{g} \to Ad_g(\beta) \Rightarrow \begin{cases} S \Big[ Q \Big( Ad_g(\beta) \Big) \Big] = S \Big( Q \Big) \\ I \Big[ Ad_g(\beta) \Big] = I \Big( \beta \Big) \end{cases}$$

$$I(\beta) = -\frac{\partial^2 \Phi}{\partial \beta^2} \text{ with } \Phi(\beta) = -\log \int_M e^{-\langle \beta, U(\xi) \rangle} d\lambda$$

$$S(Q) = \langle \beta, Q \rangle - \Phi(\beta) \text{ with } Q = \frac{\partial \Phi(\beta)}{\partial \beta} \in \mathfrak{g}^* \text{ and } \beta = \frac{\partial S(Q)}{\partial Q} \in \mathfrak{g}$$



## Fisher-Souriau Metric Definition by Souriau Cocycle & Moment Map

#### Statistical Mechanics & Fisher Metric

Souriau has proposed a Riemannian metric that we have identified as a generalization of the Fisher metric:

$$I(\beta) = \begin{bmatrix} g_{\beta} \end{bmatrix} \text{ with } g_{\beta}([\beta, Z_1], [\beta, Z_2]) = \tilde{\Theta}_{\beta}(Z_1, [\beta, Z_2])$$
  
with  $\tilde{\Theta}_{\beta}(Z_1, Z_2) = \tilde{\Theta}(Z_1, Z_2) + \langle Q, ad_{Z_1}(Z_2) \rangle$  where  $ad_{Z_1}(Z_2) = \begin{bmatrix} Z_1, Z_2 \end{bmatrix}$ 

> The tensor  $\tilde{\Theta}$  used to define this extended Fisher metric is defined by the moment map J(x), from M (homogeneous symplectic manifold) to the dual Lie algebra  $\mathfrak{g}^*$ , given by:

$$\tilde{\Theta}(X,Y) = J_{[X,Y]} - \{J_X, J_Y\} \text{ with } J(X): M \to \mathfrak{g}^* \text{ such that } J_X(X) = \langle J(X), X \rangle, X \in \mathfrak{g}$$

> This tensor  $\tilde{\Theta}$  is also defined in tangent space of the cocycle  $\theta(g) \in \mathfrak{g}^*$  (this cocycle appears due to the non-equivariance of the coadjoint operator  $Ad_g^*$ , action of the group on the dual lie algebra):  $Q(Ad_g(\beta)) = Ad_g^*(Q) + \theta(g)$ 

$$\tilde{\Theta}(X,Y): \mathfrak{g} \times \mathfrak{g} \to \mathfrak{R}$$
 with  $\Theta(X) = T_e \theta(X(e))$ 

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Joint Structures and Common Foundations of Statistical Physics,  $Y\mapstoigl\langle\Theta(X),Y
ight
angle$ 

### Fisher-Souriau Metric as a non-null Cohomology extension of KKS 2 form (Kirillov-Kostant-Souriau 2 form)

Souriau definition of Fisher Metric is related to the extension of KKS 2-form (Kostant-Kirillov-Souriau) in case of non-null Cohomogy:

with  $\Theta_{\beta}(Z_1, Z_2) = \Theta(Z_1, Z_2) + \langle Q, [Z_1, Z_2] \rangle$ 

Souriau-Fisher Metric

$$I(\beta) = \left[g_{\beta}\right] \text{ with } g_{\beta}\left(\left[\beta, Z_{1}\right], \left[\beta, Z_{2}\right]\right) = \tilde{\Theta}_{\beta}\left(Z_{1}, \left[\beta, Z_{2}\right]\right)$$

 $Q(Ad_g(\beta)) = Ad_g^*(Q) + \theta(g)$ 

Non-null cohomology: aditional term from Souriau Cocycle

Equivariant KKS 2 form

$$\begin{split} \tilde{\Theta}(X,Y) &= J_{[X,Y]} - \left\{ J_X, J_Y \right\} & \text{with } J(x) : M \to \mathfrak{g}^* & \text{such that } J_X(x) = \left\langle J(x), X \right\rangle, X \in \mathfrak{g} \\ \tilde{\Theta}(X,Y) : \mathfrak{g} \times \mathfrak{g} \to \mathfrak{R} & \text{with } \Theta(X) = T_e \theta \left( X(e) \right) & \tilde{\Theta}(\beta,Z) + \left\langle Q, [\beta,Z] \right\rangle = 0 \\ & X, Y \mapsto \left\langle \Theta(X), Y \right\rangle & \beta \in Ker \, \tilde{\Theta}_{\beta} \end{split}$$

OPEN

Souriau Fundamental **Equation** of Lie Group Thermodynamics Structures and Common Foundations of Statistical Physics

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#### **Fundamental Souriau Theorem**



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## Non-equivariance of Coadjoint operator

> Non-equivariance of Coadjoint operator:  $Q(Ad_g(\beta)) = Ad_g^*(Q) + \theta(g)$ 

> This is the action of Lie Group on Moment map:

$$J(\Phi_g(x)) = a(g, J(x)) = Ad_g^*(J(x)) + \theta(g)$$

> By noting the action of the group on the dual space of the Lie algebra:  $G \times \mathfrak{g}^* \to \mathfrak{g}^*, (s, \xi) \mapsto s\xi = Ad_s^*\xi + \theta(s)$ 

Associativity is given by:

$$(s_1s_2)\xi = Ad_{s_1s_2}^*\xi + \theta(s_1s_2) = Ad_{s_1}^*Ad_{s_2}^*\xi + \theta(s_1) + Ad_{s_1}^*\theta(s_2)$$

$$(s_{1}s_{2})\xi = Ad_{s_{1}}^{*}\left(Ad_{s_{2}}^{*}\xi + \theta(s_{2})\right) + \theta(s_{1}) = s_{1}\left(s_{2}\xi\right) , \ \forall s_{1}, s_{2} \in G, \xi \in \mathfrak{g}^{*}$$

## Souriau Cocycle

▶  $\theta(g) \in \mathfrak{g}^*$  is called nonequivariance one-cocycle, and it is a measure of the lack of equivariance of the moment map.

$$\theta(st) = J((st).x) - Ad_{st}^*J(x)$$
  

$$\theta(st) = \left[J(s.(t.x)) - Ad_s^*J(t.x)\right] + \left[Ad_s^*J(t.x) - Ad_s^*Ad_t^*J(x)\right]$$
  

$$\theta(st) = \theta(s) + Ad_s^*\left[J(t.x) - Ad_t^*J(x)\right]$$
  

$$\theta(st) = \theta(s) + Ad_s^*\theta(t)$$

### Souriau one-cocycle and compute 2-cocycle

$$\tilde{\Theta}(X,Y):\mathfrak{g}\times\mathfrak{g}\to\mathfrak{R}$$
 with  $\Theta(X)=T_e\theta(X(e))$ 

 $X, Y \mapsto \langle \Theta(X), Y \rangle$ > We can also compute tangent of one-cocycle  $\theta$  at neutral element, to compute 2-cocycle  $\Theta$ :

$$\begin{aligned} \zeta \in \mathfrak{g} \quad , \quad \theta_{\zeta}\left(s\right) &= \left\langle \theta\left(s\right), \zeta \right\rangle = \left\langle J\left(s.x\right), \zeta \right\rangle - \left\langle Ad_{s}^{*}J\left(x\right), \zeta \right\rangle \\ \theta_{\zeta}\left(s\right) &= \left\langle J\left(s.x\right), \zeta \right\rangle - \left\langle J\left(x\right), Ad_{s^{-1}}\zeta \right\rangle \\ T_{e}\theta_{\zeta}\left(\xi\right) &= \left\langle T_{x}J.\xi_{p}\left(x\right), \zeta \right\rangle + \left\langle J\left(x\right), ad_{\xi}\zeta \right\rangle \quad \text{with} \quad \xi_{p} = X_{\langle J,\xi \rangle} \\ T_{e}\theta_{\zeta}\left(\xi\right) &= X_{\langle J(x),\xi \rangle} \left[ \left\langle J(x), \zeta \right\rangle \right] + \left\langle J\left(x\right), \left[\xi,\zeta\right] \right\rangle \end{aligned}$$

$$T_{e} \theta_{\zeta}(\xi) = -\left\{ \langle J, \xi \rangle, \langle J, \zeta \rangle \right\} + \langle J(x), [\xi, \zeta] \rangle = \Theta(\xi)$$

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#### Souriau Tensor

$$\tilde{\Theta}(X,Y) = J_{[X,Y]} - \{J_X, J_Y\} = -\langle d\theta(X), Y \rangle , X, Y \in \mathfrak{g}$$
$$\tilde{\Theta}([X,Y],Z) + \tilde{\Theta}([X,Y],Z) + \tilde{\Theta}([X,Y],Z) = 0 , X,Y,Z \in \mathfrak{g}$$

> By differentiating the equation on affine action, we have:

$$T_{x}J(\xi_{p}(x)) = -ad_{\xi}^{*}J(x) + \Theta(\xi,.)$$

$$dJ(Xx) = ad_X J(x) + d\theta(X) , x \in M, X \in \mathfrak{g}$$
  
$$\langle dJ(Xx), Y \rangle = \langle ad_X J(x), Y \rangle + \langle d\theta(X), Y \rangle, x \in M, X, Y \in \mathfrak{g}$$
  
$$\langle dJ(Xx), Y \rangle = \langle J(x), [X, Y] \rangle + \langle d\theta(X), Y \rangle = \{\langle J, X \rangle, \langle J, Y \rangle\}(x)$$
  
$$\langle J(x), [X, Y] \rangle - \{\langle J, X \rangle, \langle J, Y \rangle\}(x) = -\langle d\theta(X), Y \rangle$$

#### Souriau-Fisher Metric & Souriau Lie Groups Thermodynamics: Bedrock for Lie Group Machine Learning



## Link with Classical Thermodynamics

### We have the reciprocal formula:

$$Q = \frac{\partial \Phi}{\partial \beta} \qquad \qquad \beta = \frac{\partial s}{\partial Q}$$
$$s(Q) = \left\langle \frac{\partial \Phi}{\partial \beta}, \beta \right\rangle - \Phi \qquad \qquad \Phi(\beta) = \left\langle Q, \frac{\partial s}{\partial Q} \right\rangle - s$$

For Classical Thermodynamics (Time translation only), we recover the definition of Boltzmann-Clausius Entropy:

$$\begin{cases} \beta = \frac{\partial s}{\partial Q} \\ \beta = \frac{1}{T} \end{cases} \Rightarrow ds = \frac{dQ}{T} \end{cases}$$

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## Covariant Souriau-Gibbs density

> Souriau has then defined a Gibbs density that is covariant under the action of the group:  $\Phi(\beta)=/U(\xi),\beta)$   $e^{-\langle U(\xi),\beta\rangle}$ 

$$p_{Gibbs}(\xi) = e^{\Phi(\beta) - \langle U(\xi), \beta \rangle} = \frac{C}{\int_{M} e^{-\langle U(\xi), \beta \rangle} d\lambda_{\omega}}$$
  
with  $\Phi(\beta) = -\log \int_{M} e^{-\langle U(\xi), \beta \rangle} d\lambda_{\omega}$   
$$Q = \frac{\partial \Phi(\beta)}{\partial \beta} = \frac{\int_{M} U(\xi) e^{-\langle U(\xi), \beta \rangle} d\lambda_{\omega}}{\int_{M} e^{-\langle U(\xi), \beta \rangle} d\lambda_{\omega}} = \int_{M} U(\xi) p(\xi) d\lambda_{\omega}$$

> We can express the Gibbs density with respect to Q by inverting the relation  $Q = \frac{\partial \Phi(\beta)}{\partial \beta} = \Theta(\beta)$ . Then  $p_{Gibbs,Q}(\xi) = e^{\Phi(\beta) - \langle U(\xi), \Theta^{-1}(Q) \rangle}$  with  $\beta = \Theta^{-1}(Q)$ 

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### Souriau Lie Groups Thermodynamics: Geometric Calorific Capacity

Nous prenons désormais Z dans C . La valeur moyenne du moment  $\Psi(x)$  dans l'état de Gibbs est égal à la dérivée

Q = z'(Z);

 $Z \mapsto Q$  est un difféomorphisme analytique de C sur un ouvert convexe de  $\mathcal{G}^*$ ; la transformée de Legendre s de z :

s(0) = 0Z-z

 $K \neq z^{22}(Z)$ 

y est convexe et vérifie I = s'(0): la dérivée seconde:

est un tenseur positif. dont l'inverse est égal à s''(Q).

K munit l'ensemble C d'une structure riemannienne invariante par l'action du groupe; pour cette structure, l'application linéaire Ad(Z) est antihermitienne.

```
L'application f_{Z'}, définie par:

f_{Z'}(Z',Z'') = K([Z,Z'],Z'') \quad \forall Z',Z'' \in \mathcal{G}_{T}
```

Souriau-Fisher Metric is a Geometrization of Thermodynamical «Calorific Capacity» (Pierre Duhem has deeply developed this idea of « generalized capacities »)

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Dans le cas classique, on ne considère que le groupe de dimension 1 des translations temporelles (qui n'est défini qu'après avoir choisi un référentiel - par exemple celui de la boîte qui contient le gaz). Alors, avec des unités convenables, Z est l'inverse de la TEMPERATURE ABSOLUE; z est le POTENTIEL THERMODYNAMIQUE DE PLANCK: -s est l'ENTROPIE; Q est l'ENERGIE INTERNE; K caractérise la CAPACITE CALORIFIQUE.

est un cocycle symplectique, cohomologue à f [formule (2,7 C)]; son noyau est l'orthogonal de l'orbite adjointe de Z pour la structure riemannienne de C.

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## Souriau Lie Group Thermodynamics: Geometric Calorific Capacity

Il faut bien entendu que cette intégrale soit convergente ; nous définirons l'ensemble canonique de Gibbs  $\Omega_{
m comme}$  comme le plus grand ouvert (dans l'algèbre de Lie) où cette intégrale est localement normalement conver- $\odot$  ). On montre que  $\Omega$  est convexe, et que z est une cente (en ; que la dérivée Q = <u>dr</u> C<sup>m</sup> sur  $\Omega$ fonction coïncide avec la valeur moyenne de l'énergie E ( Q cénéralise donc la chaleur) : que le est synétrique et positif (il généralise la capacité calotenseur en résulte que z est <u>fonction convexe</u> de 👄 ; la transrifique). formation de Legendre lui associe une fonction concave. à savoir

(7.3)  $Q \mapsto s = z - Q \Theta$ 

#### s est l'entropie.

#### Souriau-Fisher Metric based on cocycle

pour chaque "température"  $\bigcirc$ , définissons un tenseur  $f_{\bigcirc}$ , somme du cocycle f (défini en (3.2)) et du <u>cobord de la chaleur</u>:

(7.4) 
$$f_{\Theta}(z,z') = f(z,z') + \varphi[z,z']$$

 $f_{\Theta} \text{ jouit alors des propriétés suivantes :}$   $f_{\Theta} \text{ est un <u>cocycle symplectique</u> ;}$   $f_{\Theta} \text{ est un <u>cocycle symplectique</u> ;}$   $f_{\Theta} \text{ (c) Le tenseur symétrique } g_{\Theta} \text{ , défini sur l'ensemble de valeurs de ad($\Theta$)}$   $g_{\Theta}([\Theta, Z], [\Theta, L']) = f_{\Theta}(Z, [\Theta, Z'])$ 

est positif (et même défini positif si l'action du groupe est effective).

Ces formules sont <u>universelles</u>, en ce sens qu'elles ne mettent pas en jeu la variété symplectique U - mais seulement le groupe G, son cocycle symplectique f et les couples  $\Theta$ , Q. Peut-être cette "thermojynamique des groupes de Lie" a-t-elle un intérêt mathématique.

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#### Souriau Lie Groups Thermodynamics: General Equations

Ces formules sont <u>universelles</u>, en ce sens qu'elles ne mettent pas en jeu la variété symplectique U - mais seulement le groupe G, son cocycle symplectique f et les couples  $\Theta$ , Q. Peut-être cette "thermodynamique des groupes de Lie" a-t-elle un intérêt mathématique.



## SOURIAU GEOMETRIC THEORY OF HEAT

#### Multivariate Gaussian Density as 1st order Maximum Entropy in Souriau Book (Chapter IV)

Exemple : (loi normale) :

Prenons le cas 
$$V = R^n$$
,  $\lambda =$  mesure de Lebesgue,  $\Psi(x) \equiv$ 

$$\begin{pmatrix} x \\ x \otimes x \end{pmatrix};$$

un élément Z du dual de E peut se définir par la formule

$$Z(\Psi(x)) \equiv \overline{a} \cdot x + \frac{1}{2} \overline{x} \cdot H \cdot x$$

 $[a \in R^*; H = \text{matrice symétrique}]$ . On vérifie que la convergence de l'intégrale  $I_0$  a lieu si la matrice H est positive (<sup>1</sup>); dans ce cas la loi de Gibbs s'appelle *loi normale de Gauss*; on calcule facilement  $I_0$  en faisant le changement de variable  $x^* = H^{1/2} x + H^{-1/2} a$  (<sup>2</sup>); il vient

 $z = \frac{1}{2} \left[ \overline{a} \cdot H^{-1} \cdot a - \log (\det (H)) + n \log (2 \pi) \right]$ 

alors la convergence de  $I_1$  a lieu également; on peut donc calculer M, qui est défini par les moments du premier et du second ordre de la loi (16.196); le calcul montre que le moment du premier ordre est égal à  $-H^{-1}$ . a et que les composantes du tenseur variance (16.196) sont égales aux éléments de la matrice  $H^{-1}$ ; le moment du second ordre s'en déduit immédiatement.

La formule (16.200 ) donne l'entropie :

$$s = \frac{n}{2}\log(2 \pi e) - \frac{1}{2}\log(\det(H))$$
;

(<sup>1</sup>) Voir Calcul linéaire, tome II.
 (<sup>2</sup>) C'est-à-dire en recherchant l'image de la loi par l'application x → x<sup>\*</sup>.

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DÉPARTEMENT MATHÉMATIQUE Dirigi par le Professeur P. LELONG

#### STRUCTURE <sup>des</sup> SYSTÈMES DYNAMIQUES

Maîtrises de mathématiques

J.-M. SOURIAG

DUNOD

http://www.jmsouriau.com/structure\_ des\_systemes\_dynamiques.htm

#### Example of Multivariate Gaussian Law (real case)

Multivariate Gaussian law parameterized by moments

$$p_{\xi}(\xi) = \frac{1}{(2\pi)^{n/2} \det(R)^{1/2}} e^{-\frac{1}{2}(z-m)^{T}R^{-1}(z-m)}$$

$$\frac{1}{2}(z-m)^{T}R^{-1}(z-m) = \frac{1}{2} \begin{bmatrix} z^{T}R^{-1}z - m^{T}R^{-1}z - z^{T}R^{-1}m + m^{T}R^{-1}m \end{bmatrix}$$

$$= \frac{1}{2} z^{T}R^{-1}z - m^{T}R^{-1}z + \frac{1}{2}m^{T}R^{-1}m$$

$$p_{\xi}(\xi) = \frac{1}{(2\pi)^{n/2} \det(R)^{1/2}} e^{\frac{1}{2}m^{T}R^{-1}m} e^{-\left[-m^{T}R^{-1}z + \frac{1}{2}z^{T}R^{-1}z\right]} = \frac{1}{Z} e^{-\langle\xi,\beta\rangle}$$

$$p_{\xi}(\xi) = \frac{1}{(2\pi)^{n/2} \det(R)^{1/2}} e^{\frac{1}{2}m^{T}R^{-1}m} e^{-\left[-m^{T}R^{-1}z + \frac{1}{2}z^{T}R^{-1}z\right]} = \frac{1}{Z} e^{-\langle\xi,\beta\rangle}$$

$$\xi = \begin{bmatrix} z \\ zz^{T} \end{bmatrix} \text{ and } \beta = \begin{bmatrix} -R^{-1}m \\ \frac{1}{2}R^{-1} \end{bmatrix} = \begin{bmatrix} a \\ H \end{bmatrix} \text{ with } \langle\xi,\beta\rangle = a^{T}z + z^{T}Hz = Tr[za^{T} + H^{T}zz^{T}]$$

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THALES

# THALES

## SOURIAU ENTROPY AS INVARIANT CASIMIR FUNCTION IN COADJOINT REPRESENTATION



## Gromov question: Are there « entropies » associated to moment maps

#### **Bernoulli Lecture - What is Probability?**

- > 27 March 2018 CIB EPFL Switzerland
- > Lecturer: Mikhail Gromov
- https://bernoulli.epfl.ch/images/website/What is Probability\_v2(2).mp4
- http://forum.cs-dc.org/uploads/files/1525172771489-alternative-probabilities-2018.pdf

Fisher Metric. Recall (Archimedes, 287-212 BCE) the real moment map from the unit sphere  $S^n \subset \mathbb{R}^{n+1}$  to the probability simplex  $\triangle^n \subset \mathbb{R}^{n+1}$  for

 $(x_0, ..., x_n) \mapsto (p_0 = x_0^2, ..., p_n = x_n^2)$ and observe following R. Fisher that the spherical metric (with constant curvature +1) thus transported to  $\triangle^n$ , call it  $ds^2$  on  $\triangle^n$ , is equal, up to a scalar multiple, to the Hessian of the entropy  $ent\{p_0, ..., p_n\} = -\sum_i p_i \log p_i.$  $ds^2 = const \frac{\partial^2 ent(p_i)}{dp_i dp_i}.$  If, accordingly, we take the "inverse Hessian" – a kind of double integral " $\int \int ds^2$ " for the *definition* of entropy – we arrive at

Question 2. Are there *interesting* "entropies" associated to (real and complex) moment maps of general toric varieties? Is there a *meaningful* concept of "generalised probability" grounded in positivity encountered in algebraic geometry?



## Souriau Entropy Invariance

## Casimir Invariant Function in coadjoint representation

> We conclude the paper by a deeper study of Souriau model structure. We observe that Souriau Entropy S(Q) defined on coadjoint orbit of the group has a property of invariance :

$$S\left(Ad_g^{\#}(Q)\right) = S\left(Q\right)$$

> with respect to Souriau affine definition of coadjoint action:

$$Ad_g^{\#}(Q) = Ad_g^{*}(Q) + \theta(g)$$

> where  $\theta(g)$  is called the Souriau cocyle.

$$Q(Ad_g(\beta)) = Ad_g^*(Q) + \theta(g)$$

$$Q\left(Ad_{g}(\beta)\right) = S\left(Q\right)$$

New Entropy Definition: Function in Coadjoint Representation Invariant under the action of the Group



Hendrik Casimir (Thesis supervised by Niels Bohr & Paul Ehrenfest)

H.B.G. Casimir, On the Rotation of a Rigid Body in Quantum Mechanics, Doctoral Thesis, Leiden, 1931.

## Entropy as Invariant Casimir Function in Coadjoint Representation

NEW GEOMETRIC DEFINITION OF ENTROPY

$$\{S,H\}_{\tilde{\Theta}}(Q)=0$$

$$ad_{\frac{\partial S}{\partial Q}}^*Q + \Theta\left(\frac{\partial S}{\partial Q}\right) = 0$$

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$$\{S,H\}(Q) = \left\langle Q, \left[\frac{\partial S}{\partial Q}, \frac{\partial H}{\partial Q}\right] \right\rangle = -C_{ij}^{k}Q_{k}\frac{\partial S}{\partial Q_{i}}\cdot\frac{\partial H}{\partial Q_{j}}$$
$$\left[e_{i}, e_{j}\right] = C_{ij}^{k}e_{k} , C_{ij}^{k} \text{ structure coefficients}$$

$$\left\{S,H\right\}_{\tilde{\Theta}}\left(Q\right) = \left\langle Q, \left[\frac{\partial S}{\partial Q}, \frac{\partial H}{\partial Q}\right] \right\rangle + \tilde{\Theta}\left(\frac{\partial S}{\partial Q}, \frac{\partial H}{\partial Q}\right) = 0 \quad , \quad \forall H : \mathfrak{g}^* \to R, \quad Q \in \mathfrak{g}^*$$

$$\tilde{\Theta}(X,Y) = J_{[X,Y]} - \{J_X, J_Y\} \text{ where } J_X(X) = \langle J(X), X \rangle$$
$$\tilde{\Theta}(X,Y) = \langle \Theta(X), Y \rangle \text{ with } \Theta(X) = T_e \theta(X(e))$$

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## Geometric Definition of Entropy

In the framework of Souriau Lie groups Thermodynamics, we can characterize the Entropy as a generalized Casimir invariant function in coadjoint representation,

## Geometric Definition of Massieu Characteristic Function

Massieu characteristic function (or log-partition function), dual of Entropy by Legendre transform, as a generalized Casimir function in adjoint representation.

## Casimir Function Definition

When M is a Poisson manifold, a function on M is a Casimir function if and only if this function is constant on each symplectic leaf (the non-empty open subsets of the symplectic leaves are the smallest embedded manifolds of M which are Poisson submanifolds)

#### Entropy Invariance under the action of the Group (1/2)

$$\begin{split} \beta &\in \mathfrak{g} \to Ad_g(\beta) \Longrightarrow \Psi(Ad_g(\beta)) = \int_M e^{-\langle U(Ad_g(\beta) \rangle} d\lambda_{\omega} \\ \Psi(Ad_g(\beta)) &= \int_M e^{-\langle Ad_{g^{-1}}^*U, \beta \rangle} d\lambda_{\omega} = \int_M e^{-\langle U(Ad_{g^{-1}}\beta) - \theta(g^{-1}), \beta \rangle} d\lambda_{\omega} \\ \Psi(Ad_g(\beta)) &= e^{\langle \theta(g^{-1}), \beta \rangle} \Psi(\beta) \end{split}$$

$$\theta(g^{-1}) = -Ad_{g^{-1}}^*\theta(g) \Longrightarrow \Psi(Ad_g(\beta)) = e^{-\langle Ad_{g^{-1}}^*\theta(g),\beta \rangle} \Psi(\beta)$$
  
$$\Phi(\beta) = -\log \Psi(\beta)$$

 $\Rightarrow \Phi\left(Ad_{g^{-1}}(\beta)\right) = \Phi\left(\beta\right) - \left\langle\theta\left(g^{-1}\right),\beta\right\rangle = \Phi\left(\beta\right) + \left\langle Ad_{g^{-1}}^{*}\theta(g),\beta\right\rangle$ 97 Information Geometry and Inference for Learning (SPIGL'20)

## Entropy Invariance under the action of the Group (2/2)

$$S(Q) = \langle Q, \beta \rangle - \Phi(\beta) \Rightarrow S(Q(Ad_{g}\beta)) = \langle Q(Ad_{g}\beta), Ad_{g}\beta \rangle - \Phi(Ad_{g}\beta)$$

$$\begin{cases} Q(Ad_{g}(\beta)) = Ad_{g}^{*}(Q) + \theta(g) \\ \Phi(Ad_{g}(\beta)) = -\log \Psi(Ad_{g}(\beta)) = -\langle \theta(g^{-1}), \beta \rangle + \Phi(\beta) \\ \Rightarrow S(Q(Ad_{g}\beta)) = \langle Ad_{g}^{*}(Q) + \theta(g), Ad_{g}\beta \rangle + \langle \theta(g^{-1}), \beta \rangle - \Phi(\beta) \\ \Rightarrow S(Q(Ad_{g}\beta)) = \langle Ad_{g}^{*}(Q) + \theta(g), Ad_{g}\beta \rangle - \langle Ad_{g^{-1}}^{*}\theta(g), \beta \rangle - \Phi(\beta) \\ \Rightarrow S(Q(Ad_{g}\beta)) = \langle Ad_{g}^{*}(Q) + \theta(g), Ad_{g}\beta \rangle - \langle Ad_{g^{-1}}^{*}\theta(g), \beta \rangle - \Phi(\beta) \\ \Rightarrow S(Q(Ad_{g}\beta)) = \langle Ad_{g^{-1}}^{*}Ad_{g}^{*}(Q) + Ad_{g^{-1}}^{*}\theta(g), \beta \rangle - \langle Ad_{g^{-1}}^{*}\theta(g), \beta \rangle - \Phi(\beta) \\ \Rightarrow Ad_{g^{-1}}^{*}Ad_{g}^{*}(Q) = Q \Rightarrow S(Q(Ad_{g}\beta)) = \langle Q, \beta \rangle - \Phi(\beta) = S(\beta) \\ \xrightarrow{Q_{g^{-1}}} HALEES$$

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### Casimir Function and Entropy

- Classically, the Entropy is defined axiomatically as Shannon or von Neumann Entropies without any geometric structures constraints.
- > Entropy could be built by Casimir Function Equation:

$$\left(ad_{\frac{\partial S}{\partial Q}}^{*}Q\right)_{j} + \Theta\left(\frac{\partial S}{\partial Q}\right)_{j} = C_{ij}^{k}ad_{\left(\frac{\partial S}{\partial Q}\right)^{i}}^{*}Q_{k} + \Theta_{j} = 0$$

$$\tilde{\Theta}(X,Y) = \langle \Theta(X),Y \rangle = J_{[X,Y]} - \{J_X,J_Y\} = -\langle d\theta(X),Y \rangle \quad , \ X,Y \in \mathfrak{g}$$

$$\Theta(X) = T_e \theta (X(e))$$
$$\theta (g) = Q (Ad_g(\beta)) - Ad_g^*(Q)$$

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## Demo

- > if we consider the heat expression  $Q = \frac{\partial \Phi}{\partial \beta}$ , that we can write  $\partial \Phi \langle Q, \partial \beta \rangle = 0$ . > For each  $\partial \beta$  tangent to the orbit, and so generated by an element Z of the Lie algebra, if we consider the relation  $\Phi(Ad_g(\beta)) = \Phi(\beta) \langle \theta(g^{-1}), \beta \rangle$ , we differentiate it at g = e using the property that:

$$\widetilde{\Theta}(X,Y) = -\langle d\theta(X),Y \rangle , X,Y \in \mathfrak{g}$$

> we obtain : 
$$\langle Q, [\beta, Z] \rangle + \tilde{\Theta}(\beta, Z) = 0$$

From last Souriau equation, if we use the identities  $\beta = \frac{\partial S}{\partial Q}$ ,  $ad_{\beta}Z = [\beta, Z]$  and  $\tilde{\Theta}(\beta, Z) = \langle \Theta(\beta), Z \rangle$ Then we can deduce that:  $\langle ad_{\frac{\partial S}{\partial Q}}^* Q + \Theta(\frac{\partial S}{\partial Q}), Z \rangle = 0, \forall Z$ So, Entropy S(Q) should verify:

$$\begin{cases} \frac{\partial S}{\partial Q} \\ \frac{\partial Q}{\partial Q} \end{cases} = 0 \quad \{S, H\}_{\tilde{\Theta}} (Q) = 0 \quad \forall H : \mathfrak{g}^* \to R, \quad Q \in \mathfrak{g}^* \\ \{S, H\}_{\tilde{\Theta}} (Q) = \left\langle Q, \left[ \frac{\partial S}{\partial Q}, \frac{\partial H}{\partial Q} \right] \right\rangle + \tilde{\Theta} \left( \frac{\partial S}{\partial Q}, \frac{\partial H}{\partial Q} \right) = 0 \quad \mathbf{1} \\ \begin{cases} \frac{\partial Q}{\partial Q} \\ \frac{\partial$$

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 $ad_{\partial S}^*Q+\Theta$ 

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Demo

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$$\{S,H\}_{\tilde{\Theta}}(Q) = \left\langle Q, \left[\frac{\partial S}{\partial Q}, \frac{\partial H}{\partial Q}\right] \right\rangle + \left\langle \Theta\left(\frac{\partial S}{\partial Q}\right), \frac{\partial H}{\partial Q} \right\rangle = 0$$

$$\{S,H\}_{\tilde{\Theta}}(Q) = \left\langle Q, ad_{\frac{\partial S}{\partial Q}} \frac{\partial H}{\partial Q} \right\rangle + \left\langle \Theta\left(\frac{\partial S}{\partial Q}\right), \frac{\partial H}{\partial Q} \right\rangle = 0$$

$$\{S,H\}_{\tilde{\Theta}}(Q) = \left\langle ad_{\frac{\partial S}{\partial Q}}^*Q, \frac{\partial H}{\partial Q} \right\rangle + \left\langle \Theta\left(\frac{\partial S}{\partial Q}\right), \frac{\partial H}{\partial Q} \right\rangle = 0$$

$$\forall H, \{S,H\}_{\tilde{\Theta}}(Q) = \left\langle ad_{\frac{\partial S}{\partial Q}}^*Q + \Theta\left(\frac{\partial S}{\partial Q}\right), \frac{\partial H}{\partial Q} \right\rangle = 0 \Rightarrow ad_{\frac{\partial S}{\partial Q}}^*Q + \Theta\left(\frac{\partial S}{\partial Q}\right) = 0$$

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#### Link with Souriau development

> Souriau property:  $\beta \in Ker\tilde{\Theta}_{\beta} \Rightarrow \langle Q, [\beta, Z] \rangle + \tilde{\Theta}(\beta, Z) = 0$ 

$$\Rightarrow \left\langle Q, ad_{\beta}Z \right\rangle + \tilde{\Theta}(\beta, Z) = 0 \Rightarrow \left\langle ad_{\beta}^{*}Q, Z \right\rangle + \tilde{\Theta}(\beta, Z) = 0$$

$$\beta = \frac{\partial S}{\partial Q} \Rightarrow \left\langle ad_{\frac{\partial S}{\partial Q}}^{*}Q, Z \right\rangle + \tilde{\Theta}\left(\frac{\partial S}{\partial Q}, Z\right) = \left\langle ad_{\frac{\partial S}{\partial Q}}^{*}Q + \Theta\left(\frac{\partial S}{\partial Q}\right), Z \right\rangle = 0, \forall Z$$

$$\Rightarrow ad_{\frac{\partial S}{\partial Q}}^{*}Q + \Theta\left(\frac{\partial S}{\partial Q}\right) = 0$$

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## Dynamic equation

- The dual space of the Lie algebra foliates into coadjoint orbits that are also the level sets on the entropy.
- The information manifold foliates into level sets of the entropy that could be interpreted in the framework of Thermodynamics by the fact that motion remaining on this complex surfaces is non-dissipative, whereas motion transversal to these surfaces is dissipative, where the dynamic is given by:

$$\frac{dQ}{dt} = \left\{Q, H\right\}_{\tilde{\Theta}} = ad_{\frac{\partial H}{\partial Q}}^* Q + \Theta\left(\frac{\partial H}{\partial Q}\right)$$

> with stable equilibrium when:

$$H = S \Longrightarrow \frac{dQ}{dt} = \{Q, S\}_{\tilde{\Theta}} = ad_{\frac{\partial S}{\partial Q}}^* Q + \Theta\left(\frac{\partial S}{\partial Q}\right) = 0$$

## Entropy Production and 2<sup>nd</sup> Principle

## 2<sup>nd</sup> Principle

- > We can observe that:  $dS = \tilde{\Theta}_{\beta} \left( \frac{\partial H}{\partial O}, \beta \right) dt$
- > Where:

$$\tilde{\Theta}_{\beta}\left(\frac{\partial H}{\partial Q},\beta\right) = \tilde{\Theta}\left(\frac{\partial H}{\partial Q},\beta\right) + \left\langle Q, \left[\frac{\partial H}{\partial Q},\beta\right] \right\rangle$$

- > showing that Entropy production is linked with Souriau tensor related to Fisher metric:  $\frac{dS}{dt} = \tilde{\Theta}_{\beta} \left( \frac{\partial H}{\partial Q}, \beta \right) \ge 0$
- It allows to introduce the stochastic extension based on a Stratonovich differential equation for the stochastic process given by the following relation by mean of Souriau's symplectic cocycle

$$dQ + \left[ad_{\frac{\partial H}{\partial Q}}^{*}Q + \Theta\left(\frac{\partial H}{\partial Q}\right)\right]dt + \sum_{i=1}^{N} \left[ad_{\frac{\partial H_{i}}{\partial Q}}^{*}Q + \Theta\left(\frac{\partial H_{i}}{\partial Q}\right)\right] \circ dW_{i}(t) = 0$$

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## **Entropy Production and 2nd Principle**

**Demo**  

$$S(Q) = \langle Q, \beta \rangle - \Phi(\beta) \quad \text{with } \frac{dQ}{dt} = ad_{\frac{\partial H}{\partial Q}}^* Q + \Theta\left(\frac{\partial H}{\partial Q}\right)$$

$$\frac{dS}{dt} = \left\langle Q, \frac{d\beta}{dt} \right\rangle + \left\langle ad_{\frac{\partial H}{\partial Q}}^* Q + \Theta\left(\frac{\partial H}{\partial Q}\right), \beta \right\rangle - \frac{d\Phi}{dt} = \left\langle Q, \frac{d\beta}{dt} \right\rangle + \left\langle ad_{\frac{\partial H}{\partial Q}}^* Q, \beta \right\rangle + \left\langle \Theta\left(\frac{\partial H}{\partial Q}\right), \beta \right\rangle - \frac{d\Phi}{dt}$$

$$\frac{dS}{dt} = \left\langle Q, \frac{d\beta}{dt} \right\rangle + \left\langle Q, \left[\frac{\partial H}{\partial Q}, \beta\right] \right\rangle + \tilde{\Theta}\left(\frac{\partial H}{\partial Q}, \beta\right) - \frac{d\Phi}{dt} = \left\langle Q, \frac{d\beta}{dt} \right\rangle + \tilde{\Theta}_{\beta}\left(\frac{\partial H}{\partial Q}, \beta\right) - \left\langle \frac{\partial\Phi}{\partial\beta}, \frac{d\beta}{dt} \right\rangle$$

$$\frac{dS}{dt} = \left\langle Q, \frac{d\beta}{dt} \right\rangle + \tilde{\Theta}_{\beta}\left(\frac{\partial H}{\partial Q}, \beta\right) - \left\langle \frac{\partial\Phi}{\partial\beta}, \frac{d\beta}{dt} \right\rangle \quad \text{with } \frac{\partial\Phi}{\partial\beta} = Q$$

$$\frac{dS}{dt} = \tilde{\Theta}_{\beta}\left(\frac{\partial H}{\partial Q}, \beta\right) \ge 0, \forall H \quad (\text{link to positivity of Fisher metric)}$$

$$\begin{array}{c}
 \text{if } H = S \\
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 \text{Information Geometry} \partial Q \\
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 \end{array}$$

THALES

# THALES

## Souriau Lie Groups Thermodynamics & Covariant Gibbs Density



## Geometric (Planck) Temperature in the Lie Algrbra

- Let a Group G of a Manifold M with a moment map E, the Geometric (Planck) Temperature  $\beta$  is all elements of Lie Agebra g of G such that the following integrals converges in a neighborhood of  $\beta$ :  $I_0(\beta) = \int e^{-\langle \beta, U \rangle} d\lambda$ 
  - $ig>\langleeta,U
    angle$  notes the duality of  $\mathfrak{g}$  and  $\mathfrak{g}^*$
  - >  $d\lambda$  is the Liouville density on M
- **<u>Theorem</u>:** The function  $I_0$  is infinitly differentiable  $C^{\infty}$  in  $\Omega$  (the largest open proper subset of  $\mathfrak{g}$ ) and is n<sup>th</sup> derivative for all  $\beta \in \Omega$ , the tensor integral is convergent:  $I_n(\beta) = \int e^{-\langle \beta, U \rangle} U^{\otimes^n} d\lambda$

To each temperature  $\beta$ , we can associate probability law on M with distribution function (such that the probability law has a mass equal to 1):  $e^{\Phi(\beta) - \langle \beta, U(\xi) \rangle}$  with  $\Phi(\beta) = -\log(I_0) = -\log \int e^{-\langle \beta, U(\xi) \rangle} d\lambda$  and  $Q(\beta) = \int e^{\Phi(\beta) - \langle \beta, U(\xi) \rangle} U d\lambda = \frac{I_1}{I_0}$ > The set of these probalities law is Gibbs Ensemble of the Dynamic Group,  $\Phi$  is the formed on a common Foundations of Statistical Physics. Dirit Structures and Common Foundations of Statistical Physics. Previous Structures and Phys

## Geometric Fisher Metric: Geometric Heat Capacity

We can observe that the Geometric Heat Q is  $C^{\infty}$  function of Geometric Temperature  $\beta$  in Dual Lie Algebra  $\mathfrak{g}^*$ :  $\beta \in \mathfrak{a} \mapsto O \in \mathfrak{a}^*$  $Q(\beta) = \int_{M} e^{\Phi(\beta) - \langle \beta, U(\xi) \rangle} U d\lambda = \frac{I_1}{I}$ We have:  $Q = \frac{\partial \Phi}{\partial \beta}$  $\Phi(\beta) = -\log \int_{\Omega} e^{-\langle \beta, U(\xi) \rangle} d\lambda$ Its derivative is a 2<sup>nd</sup> order symmetric tensor:  $\frac{\partial Q}{\partial \beta} = \frac{I_2}{I_0} - \frac{I_1 \otimes I_1}{I_0} = \frac{I_2}{I_1} - Q \otimes Q$  $\frac{\partial Q}{\partial \beta} = \int_{M} e^{\Phi(\beta) - \langle \beta, U(\xi) \rangle} [U - Q] \otimes [U - Q] d\lambda \qquad -\frac{\partial Q}{\partial \beta} \ge 0$ This quatratic form is positive, and positive definite for each  $x \in M$  unless there exist a non null element  $Z \in \mathfrak{g}$  such that  $\langle U - Q, Z \rangle = 0$  (means that the moment U varies in an affine sub-manifold of  $\mathfrak{a}^*$ )

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#### Distribution of probability by Group action

The distribution density under the action of the Lie Group is given by:

$$\mu^*: e^{\Phi^*-\left}$$

$$\Phi^{*} = \Phi(\beta^{*}) = \Phi - \langle \theta(g^{-1}), \beta \rangle$$
  
$$\Phi^{*} = \Phi + \langle \theta(g), Ad_{g}\beta \rangle$$
  
(\*\*)

$$p = Ad_g(\beta)$$
  

$$\theta(g^{-1}) = -Ad_g^*\theta(g)$$
  

$$\Phi(\beta) = -\log \int e^{-\langle \beta, U(\xi) \rangle} d\lambda$$

 $\mathbf{A} = \mathbf{I} (\mathbf{O})$ 

 $o^*$ 

The set  $\Omega$  of Geometric Temperature is invariant by the<sup>M</sup>adjoint action of G $\Psi_g(\mu_\beta) = \mu_{Ad_g(\beta)}$ 

If we use  $Q = \frac{\partial \Phi}{\partial \beta}$ , we have the constraint  $\delta \Phi - \langle Q, \delta \beta \rangle = 0$ By derivation of (\*\*), we have:  $\tilde{\Theta}(\beta, Z) + \langle Q, [\beta, Z] \rangle = 0$   $\tilde{\Theta}(X, Y) : \mathfrak{g} \times \mathfrak{g} \to \mathfrak{R}$   $X, Y \mapsto \langle \Theta(X), Y \rangle$  $\Theta(X) = T_e \theta(X(e))$ 

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# Geometric (Planck) Temperature

We have previously observed that: 
$$\tilde{\Theta}(\beta, Z) + \langle Q, [\beta, Z] \rangle = 0$$

 $\widetilde{\Theta}(X,Y)$  is called the Symplectic Cocycle of Lie algebra g associated to the momentum map J

 $\widetilde{\Theta}(X,Y) = J_{[X,Y]} - \{J_X, J_Y\}$  with  $\{.,.\}$  Poisson Bracket and J the Moment Map

> where  $J_X$  linear application from  $\mathfrak{g}$  to differential function on M: and the associated differentiable application  $J_i$  called moment (um) map:  $\begin{array}{c} \mathfrak{g} \to C^{\infty}(M,R) \\ X \to J_X \end{array}$ 

$$J: M \to \mathfrak{g}^*$$
 with  $x \mapsto J(x)$  such that  $J_X(x) = \langle J(x), X \rangle, X \in \mathfrak{g}$ 

 $\widetilde{\Theta}(X,Y)$  is a 2-form of  $\mathfrak{g}$  and verify:

 $\widetilde{\Theta}([X,Y],Z) + \widetilde{\Theta}([Y,Z],X) + \widetilde{\Theta}([Z,X],Y) = 0$ 

If we define:  $\widetilde{\Theta}_{\beta}(Z_1, Z_2) = \widetilde{\Theta}(Z_1, Z_2) + \langle Q, ad_{Z_1}(Z_2) \rangle$  with  $ad_{Z_1}(Z_2) = [Z_1, Z_2]$ 

We can observe that :  $\beta \in Ker \, \widetilde{\Theta}_{\beta}$   $\widetilde{\Theta}_{\beta}(\beta,\beta) = 0$  ,  $\forall \beta \in \mathfrak{g}$ 

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# Associated Riemannian Metric: Geometric Fisher Metric

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We can compute the image of Geometric Heat by the Lie Group action:  $Q^* = Ad_{_{g}}^*(Q) + \theta(g)$ 

By tangential derivative to the orbit with respect to  $Z \in \mathfrak{g}$  and by using positivity of  $\underline{\partial Q}_{\geq 0}$ , we find:  $\widetilde{\Theta}_{\beta}(Z, [\beta, Z]) = \widetilde{\Theta}(Z, [\beta, Z]) + \langle Q, [Z, [\beta, Z]] \rangle \ge 0$  $\widetilde{\Theta}_{\beta}$  is a 2-form of  $\mathfrak{g}$  that verifies:  $\widetilde{\Theta}([X,Y],Z) + \widetilde{\Theta}([Y,Z],X) + \widetilde{\Theta}([Z,X],Y) = 0$ Then, there exists a symmetric tensor  $g_{\beta}$  defined on  $ad_{\beta}(Z)$  $g_{\beta}([\beta, Z_1], [\beta, Z_2]) = \widetilde{\Theta}_{\beta}(Z_1, [\beta, Z_2])$  $I(Ad_g(\beta)) = -\frac{\partial^2 (\Phi - \langle \theta(g^{-1}), \beta \rangle)}{\partial \beta^2} = -\frac{\partial^2 \Phi}{\partial \beta^2} = I(\beta)$ With the following invariances:  $s \left| Q(Ad_{o}(\beta)) \right| = s(Q(\beta))$ OPEN

#### Fisher Metric of Souriau Lie Group Thermodynamics

# Souriau has introduced the Riemannian metric $g_{\beta}([\beta, Z_1], [\beta, Z_2]) = \widetilde{\Theta}_{\beta}(Z_1, [\beta, Z_2])$ $\beta \in Ker \widetilde{\Theta}_{\beta}$ $\widetilde{\Theta}_{\beta}(Z_1, Z_2) = \widetilde{\Theta}(Z_1, Z_2) + \langle Q, ad_{Z_1}(Z_2) \rangle$ with $ad_{Z_1}(Z_2) = [Z_1, Z_2]$ This metric is an extension of Fisher metric, an hessian metric: If we

differentiate the relation  $Q(Ad_g(\beta)) = Ad_g^*(Q) + \theta(g)$ 

$$\frac{\partial Q}{\partial \beta} (-[Z_1,\beta],.) = \widetilde{\Theta}(Z_1,[\beta,.]) + \langle Q, Ad_{Z_1}([\beta,.]) \rangle = \widetilde{\Theta}_{\beta}(Z_1,[\beta,.])$$

$$-\frac{\partial Q}{\partial \beta}([Z_1,\beta],Z_2) = \widetilde{\Theta}(Z_1,[\beta,Z_2]) + \langle Q, Ad_{Z_1}([\beta,Z_2]) \rangle = \widetilde{\Theta}_{\beta}(Z_1,[\beta,Z_2])$$
$$\Rightarrow -\frac{\partial^2 \Phi}{\partial \beta^2} = -\frac{\partial Q}{\partial \beta} = g_{\beta}([\beta,Z_1],[\beta,Z_2]) = \widetilde{\Theta}_{\beta}(Z_1,[\beta,Z_2])$$

 $\frac{\partial T}{\partial t} = \frac{\kappa}{C.D} \Delta T \quad \text{with} \quad \frac{\partial Q}{\partial T} = C.D$ 

The Fisher Metric is then a generalization of "Heat Capacity":

# **Continuous Medium Thermodynamics**

- For Continuous Medium Thermodynamics, « Temperature Vector » is no longer constrained to be in Lie Algebra, but only contrained by phenomenologic equations (e.g. Navier equations, ...).
- For Thermodynamic equilibrium, the « Temperature Vector » is a Killing vector of Space-Time.
- For each point X, there is a « Temperature Vector »  $\beta(X)$ , such it is an infinitesimal conformal transform of the metric of the univers  $g_{ij}$ :

Conservation equations can be deduced for components of Impulsion-Energy tensor  $T^{ij}$  and Entropy flux  $S^{j}: \hat{\partial}_{i}T^{ij} = 0$   $\hat{\partial}_{i}S^{j} = 0$ 

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## Poincaré-Cartan Integral Invariant of Lie Group Thermodynamics

Analogies between Geometric Mechanics & Geometric Lie Group Thermodynamics, provides the following <u>similarities of structures</u>:

$$\begin{cases} \dot{q} \leftrightarrow \beta \\ p \leftrightarrow Q \end{cases} \begin{cases} L(\dot{q}) \leftrightarrow \Phi(\beta) \\ H(p) \leftrightarrow S(Q) \\ H = p.\dot{q} - L \leftrightarrow S = \langle Q, \beta \rangle - \Phi \end{cases} \begin{cases} \dot{q} = \frac{dq}{dt} = \frac{\partial H}{\partial p} \leftrightarrow \beta = \frac{\partial S}{\partial Q} \\ p = \frac{\partial L}{\partial \dot{q}} \leftrightarrow Q = \frac{\partial \Phi}{\partial \beta} \end{cases}$$

We can then consider a similar Poincaré-Cartan-Souriau Pfaffian form:  $\omega = p.dq - H.dt \leftrightarrow \omega = \langle Q, (\beta.dt) \rangle - S.dt = (\langle Q, \beta \rangle - S).dt = \Phi(\beta).dt$ This analogy provides an associated Poincaré-Cartan Integral Invariant:

 $\int_{C_a} p dq - H dt = \int_{C_b} p dq - H dt \text{ transforms in } \int_{C_a} \Phi(\beta) dt = \int_{C_b} \Phi(\beta) dt$ For Thermodynamics, we can then deduce an Euler-Poincaré Variational Principle: The Variational Principle holds on  $\mathfrak{g}$ , for variations  $\delta\beta = \dot{\eta} + [\beta, \eta]$ , where  $\eta(t)$  is an arbitrary path that vanishes at the endpoints,  $\eta(a) = \eta(b) = 0$ :

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$$\delta \int_{t_0}^{t_1} \Phi(\beta(t)) dt = 0$$

# Souriau Gibbs states for Hamiltonian actions of subgroups of the Galilean group

> Galilean Transformation on position and speed:

$$\begin{pmatrix} \vec{r}' & \vec{v}' \\ t' & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} A & \vec{b} & \vec{d} \\ 0 & 1 & e \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \vec{r} & \vec{v} \\ t & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} A\vec{r} + t\vec{b} + \vec{d} & A\vec{v} + \vec{b} \\ t + e & 1 \\ 1 & 0 \end{pmatrix}$$

Souriau Result: this action is Hamiltonian, with the map J, defined on the evolution map  $J(\vec{r},t,\vec{v},m) = m \begin{pmatrix} \vec{r} \times \vec{v} & 0 & 0 \\ \vec{r} - t\vec{v} & 0 & 0 \\ \vec{v} & \frac{1}{2} \|\vec{v}\|^2 & 0 \end{pmatrix} = m \{ \vec{r} \times \vec{v}, \vec{r} - t\vec{v}, \vec{v}, \frac{1}{2} \|\vec{v}\|^2 \} \in \mathfrak{g}^*$ > Coupling formula: space of the particle, with value in the dual g\* of the Lie algebra G, as momentum  $\left\langle J(\vec{r},t,\vec{v},m),\beta\right\rangle = \left\langle m\left\{\vec{r}\times\vec{v},\vec{r}-t\vec{v},\vec{v},\frac{1}{2}\|\vec{v}\|^{2}\right\},\left\{\vec{\omega},\vec{\alpha},\vec{\delta},\varepsilon\right\}\right\rangle$  $Z = \begin{pmatrix} j(\vec{\omega}) & \vec{\alpha} & \vec{\delta} \\ 0 & 1 & \varepsilon \\ 0 & 0 & 0 \end{pmatrix} = \{\vec{\omega}, \vec{\alpha}, \vec{\delta}, \varepsilon\} \in \mathfrak{g}$  $\left\langle J\left(\vec{r},t,\vec{v},m\right),\beta\right\rangle = m\left(\vec{\omega}.\vec{r}\times\vec{v}-(\vec{r}\times\vec{v}).\vec{\alpha}+\vec{v}.\vec{\delta}-\frac{1}{2}\left\|\vec{v}\right\|^{2}\varepsilon\right)$ ation Geometry and Inference for Learning (SPIGL'20)

# Souriau Gibbs states for Hamiltonian actions of subgroups of the Galilean group

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Souriau Gibbs states for one-parameter subgroups of the Galilean group

- Souriau Result: Action of the full Galilean group on the space of motions of an isolated mechanical system is not related to any Equilibrium Gibbs state (the open subset of the Lie algebra, associated to this Gibbs state, is empty)
- The 1-parameter subgroup of the Galilean group generated by β element of Lie Algebra, is the set of matrices

$$\exp(\tau\beta) = \begin{pmatrix} A(\tau) & \vec{b}(\tau) & \vec{d}(\tau) \\ 0 & 1 & \tau\varepsilon \\ 0 & 0 & 1 \end{pmatrix} \text{ with } \begin{cases} A(\tau) = \exp(\tau j(\vec{\omega})) \text{ and } \vec{b}(\tau) = \left(\sum_{i=1}^{\infty} \frac{\tau^{i}}{i!} (j(\vec{\omega}))^{i-1}\right) \vec{\alpha} \\ \vec{d}(\tau) = \left(\sum_{i=1}^{\infty} \frac{\tau^{i}}{i!} (j(\vec{\omega}))^{i-1}\right) \vec{\delta} + \varepsilon \left(\sum_{i=2}^{\infty} \frac{\tau^{i}}{i!} (j(\vec{\omega}))^{i-2}\right) \vec{\alpha} \\ \beta = \begin{pmatrix} j(\vec{\omega}) & \vec{\alpha} & \vec{\delta} \\ 0 & 1 & \varepsilon \\ 0 & 0 & 0 \end{pmatrix} \in \mathfrak{g} \end{cases}$$

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# Souriau Thermodynamics of butter churn (device used to convert cream into butter) or "La Thermodynamique de la crémière"

#### If we consider the case of the centrifuge

$$\vec{\omega} = \omega \vec{e}_z$$
,  $\vec{\alpha} = 0$  and  $\vec{\delta} = 0$ 

Rotation speed: 
$$\frac{\omega}{\varepsilon}$$
  
 $f_i(\vec{r}_{i0}) = -\frac{\omega^2}{2\varepsilon^2} \|\vec{e}_z \times \vec{r}_{i0}\|^2$ 

"The angular momentum is imparted to the gas when the molecules collide with the rotating walls, which changes the Maxwell distribution at every point, shifting its origin. The walls play the role of an angular momentum reservoir. Their motion is characterized by a certain angular velocity, and the angular velocities of the fluid and of the walls become equal at equilibrium, exactly like the equalization of the temperature through energy exchanges". – Roger Balian

with 
$$\Delta = \|\vec{e}_z \times \vec{r}_{i0}\|$$
 distance to axis z

$$\rho_i(\beta) = \frac{1}{P_i(\beta)} \exp\left(-\langle J_i, \beta \rangle\right) = cst. \exp\left(-\frac{1}{2m_i\kappa T} \|\vec{p}_{i0}\|^2 + \frac{m_i}{2\kappa T} \left(\frac{\omega}{\varepsilon}\right)^2 \Delta^2\right)$$

The behaviour of a gas made of point particles of various masses in a centrifuge rotating at a constant angular velocity (the heavier particles concentrate farther from the rotation axis than the lighter ones)

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 $\omega$ 

3

# Roger Balian Computation of Gibbs density for centrifuge

Balian made the remarks that "The angular momentum is imparted to the gas when the molecules collide with the rotating walls, which changes the Maxwell distribution at every point, shifting its origin. The walls play the role of an angular momentum reservoir. Their motion is characterized by a certain angular velocity, and the angular velocities of the fluid and of the walls become equal at equilibrium, exactly like the equalization of the temperature through energy exchanges".

#### Lie Group Thermodynamics: Centrifuge for Butter, U235 & Ribo acides

- Physiquement, la théorie donne de bons résultats si on l'applique aux divers sous-groupes du groupe de Galilée qui sont caractéristiques des appareils thermodynamiques : ainsi une boîte cylindrique dans laquelle on enferme un fluide lui laisse un saus-groupe d'invariance de dimension 2 : rotations autour de l'axe, translations temporelles. D'où résulte un vecteur température à deux dimensions, que l'on peut "transmettre" au fluide par l'intermédiaire de la boîte, (en la refroidissant, par exemple, et en la faisant tourner) ; les résultats de la théorie sont ceux-là même que l'on exploite dans les centrifugeuses (par exemple pour fabriquer du beurre, de l'uranium 235 ou des acides ribonucléiques).

- On remarquera que le processus par lequel une centrifuçeuse réfrigérée transmet son propre vecteur-température à son contenu porte deux noms différents : <u>conduction thermique</u> et <u>viscosité</u>, selon la composante du vecteur-température que l'on considère ; conduction et viscosité devraient donc lime <u>unifiées</u> dans une théorie fondamentale des processus irréversibles (théorie qui reste à construire).





# THALES

# Gibbs Density on Poincaré Unit Disk from Souriau Lie Groups Thermodynamics and SU(1,1) Coadjoint Orbits



# Poincaré Unit Disk and SU(1,1) Lie Group

> The group of complex unimodular pseudo-unitary matrices SU(1,1):  $G = SU(1,1) = \left\{ \begin{pmatrix} a & b \\ b^* & a^* \end{pmatrix} / |a|^2 - |b|^2 = 1, \ a,b \in C \right\}$ > the Lie algebra  $\mathfrak{g} = \mathfrak{su}(1,1)$  is given by:  $\mathfrak{g} = \left\{ \begin{pmatrix} -ir & \eta \\ \eta^* & ir \end{pmatrix} / r \in R, \eta \in C \right\}$ with the following bases  $(u_1, u_2, u_3) \in \mathfrak{g}$ :  $u_1 = \frac{1}{2} \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix}, u_2 = \frac{1}{2} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, u_3 = \frac{1}{2} \begin{pmatrix} -i & 0 \\ 0 & i \end{pmatrix}$ 

with the commutation relation:

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$$\begin{bmatrix} u_3, u_2 \\ u_3, u_2 \end{bmatrix} = u_1, \begin{bmatrix} u_3, u_1 \\ u_3, u_1 \end{bmatrix} = u_2, \begin{bmatrix} u_2, u_1 \\ u_2, u_1 \end{bmatrix} = -u_3$$

THALES

# Poincaré Unit Disk and SU(1,1) Lie Group

> Dual base on dual Lie algebra is named

$$\left(u_1^*, u_2^*, u_3^*\right) \in \mathfrak{g}^*$$

> The dual vector space  $\mathfrak{g}^* = \mathfrak{su}^*(1,1)$  can be identified with the subspace of  $\mathfrak{sl}(2,C)$  of the form:

$$\mathfrak{g}^* = \left\{ \begin{pmatrix} z & x + iy \\ -x + iy & -z \end{pmatrix} = x \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} + y \begin{pmatrix} 0 & i \\ i & 0 \end{pmatrix} + z \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} / x, y, z \in R \right\}$$

> Coadjoint action of  $g \in G$  on dual Lie algebra  $\xi \in \mathfrak{g}^*$  is written  $g.\xi$ 

# Coadjoint Orbit of SU(1,1) and Souriau Moment Map

> The torus  $K = \left\{ \begin{pmatrix} e^{i\theta} & 0 \\ 0 & e^{-i\theta} \end{pmatrix}, \theta \in R \right\}$  induces rotations of the unit disk

- > K leaves 0 invariant. The stabilizer for the origin 0 of unit disk is maximal compact subgroup K of SU(1,1).
- ▶ B. Cahen has observed that  $O(ru_3^*) \approx G / K$  and is diffeomorphic to the unit disk  $D = \{z \in C / |z| < 1\}$
- > The **moment map** is given by:

 $J: D \to O(ru_3^*)$ 

Benjamin Cahen, Contraction de SU(1,1) vers le groupe de Heisenberg, Travaux mathématiques, Fascicule XV, pp.19-43, (2004)

$$z \mapsto J(z) = r \left( \frac{z + z^*}{\left(1 - |z|^2\right)} u_1^* + \frac{z - z^*}{i\left(1 - |z|^2\right)} u_2^* + \frac{1 + |z|^2}{\left(1 - |z|^2\right)} u_3^* \right) \text{ALES}$$

# Coadjoint Orbit of SU(1,1) and Souriau Moment Map

$$J: D \to O_n$$

$$z \mapsto J(z) = \frac{n}{2} \left( \frac{z+z^*}{(1-|z|^2)} u_1^* + \frac{z-z^*}{i(1-|z|^2)} u_2^* + \frac{1+|z|^2}{(1-|z|^2)} u_3^* \right)$$

$$\text{Group } G \text{ act on } D \text{ by homography: } g.z = \begin{pmatrix} a & b \\ b^* & a^* \end{pmatrix} \cdot z = \frac{az+b}{a^*z+b^*}$$

$$\text{This action corresponds with coadjoint action of } G \text{ on } O_n \cdot$$

$$\text{The Kirillov-Kostant-Souriau 2-form of } O_n \text{ is given by:}$$

$$\Omega_n \left( \zeta \right) \left( X \left( \zeta \right), Y \left( \zeta \right) \right) = \left\langle \zeta, [X, Y] \right\rangle, X, Y \in \mathfrak{g} \text{ and } \zeta \in O_n$$

$$\text{ and is associated in the frame by } \Psi_n \text{ with:} \qquad \omega_n = \frac{in}{(1-|z|^2)^2} dz \wedge dz^*$$

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## Coadjoint Orbit of SU(1,1) and Souriau Moment Map

$$J(z) = r\left(\frac{z+z^{*}}{\left(1-|z|^{2}\right)}u_{1}^{*} + \frac{z-z^{*}}{i\left(1-|z|^{2}\right)}u_{2}^{*} + \frac{1+|z|^{2}}{\left(1-|z|^{2}\right)}u_{3}^{*}\right) \in O(ru_{3}^{*}), z \in D$$

- J is linked to the natural action of G on D (by fractional linear transforms) but also the coadjoint action of G on O(ru<sub>3</sub><sup>\*</sup>) = G / K
   J<sup>-1</sup> could be interpreted as the stereographic projection from the two-sphere
- >  $J^{-1}$  could be interpreted as the stereographic projection from the two-sphere  $S^2$  onto  $C \cup \infty$ :

The coadjoint action of G = SU(1,1) is the upper sheet  $x_3 > 0$  of the two-sheet hyperboloid

$$\underbrace{125}_{125} \left\{ \xi = x_1 u_1^* + x_2 u_2^* + x_3 u_3^* : -x_1^2 - x_2^2 + x_3^2 = r^2 \right\}$$

Charles-Michel Marle, Projection stéréographique et moments, hal-02157930, version 1, Juin 2019

# THALES

# **Covariant Gibbs Density by Souriau Thermodynamics**

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- We can use Kirillov representation theory and his character formula to compute Souriau covariant Gibbs density in the unit Poincaré disk.
- > For any Lie group G , a coadjoint orbit  $\mathbf{O} \subset \mathfrak{g}^*$  has a canonical symplectic form given by KKS 2-form  $\mathscr{O}_{\mathbf{O}}$  .
- > If is finite dimensional, the corresponding volume element defines a G -invariant measure supported on O, which can be interpreted as a tempered distribution.

> The Fourier transform : 
$$\Im(x) = \int_{\mathcal{O} \subset \mathfrak{g}^*} e^{-i\langle x, \lambda \rangle} \frac{1}{d!} d\omega^d \quad \text{with } \lambda \in \mathfrak{g}^* \quad \text{and} \quad x \in \mathfrak{g}$$

> is Ad G-invariant. When  $O \subset \mathfrak{g}^*$  is an integral coadjoint orbit, Kirillov has proved that this Fourier transform is related to Kirillov character  $\chi_0$  by:

$$\mathfrak{T}(x) = j(x)\chi_{O}\left(e^{x}\right) \text{ where } j(x) = \det^{1/2}\left(\frac{\sinh\left(ad\left(x/2\right)\right)}{ad\left(x/2\right)}\right)$$

## Symplectic Metric of Unit Disk

#### Symplectic Homogeneous Manifold

► Let consider  $D = \{z \in C / |z| < 1\}$  be the open unit disk of Poincaré. For each  $\rho > 0$ , the pair  $(D, \omega_{\rho})$  is a symplectic homogeneous manifold with:  $\omega_{\rho} = 2i\rho \frac{dz \wedge dz^{*}}{(1-|z|^{2})^{2}}$ where  $\omega_{\rho}$  is invariant under the action :  $SU(1,1) \times D \rightarrow D$ 

$$(g,z) \mapsto g.z = \frac{az+b}{b^*z+a^*}$$

This action is transitive and is globally and strongly Hamiltonian. Its generators are the hamiltonian vector fields associated to the functions:

$$J_{1}(z, z^{*}) = \rho \frac{1 + |z|^{2}}{1 - |z|^{2}}, J_{2}(z, z^{*}) = \frac{\rho}{i} \frac{z - z^{*}}{1 - |z|^{2}}, J_{3}(z, z^{*}) = -\rho \frac{z + z^{*}}{1 - |z|^{2}}$$

# Moment Map for SU(1,1)

#### Invariant Moment Map

> The associated moment map  $J: D \rightarrow su^*(1,1)$  defined by  $J(z).u_i = J_i(z, z^*)$ , maps D into a coadjoint orbit in  $su^*(1,1)$ .

 $(u_{3}^{*})u_{3}^{*}$ 

> Then, we can write the moment map as a matrix element of  $su^*(1,1)$ :

$$J(z) = J_{1}(z, z^{*})u_{1}^{*} + J_{2}(z, z^{*})u_{2}^{*} + J_{3}(z, z)$$
$$J(z) = \rho \begin{pmatrix} \frac{1+|z|^{2}}{1-|z|^{2}} & -2\frac{z^{*}}{1-|z|^{2}} \\ 2\frac{z}{1-|z|^{2}} & -\frac{1+|z|^{2}}{1-|z|^{2}} \end{pmatrix} \in \mathfrak{g}^{*}$$

# Moment Map & Stereographic projection

#### One sheet of the two-sheeted hyperboloid

> The moment map J is a diffeomorphism of D onto one sheet of the twosheeted hyperboloid in  $su^*(1,1)$ , determined by the following equation:

$$J_1^2 - J_2^2 - J_3^2 = \rho^2$$
,  $J_1 \ge \rho$  with  $J_1 u_1^* + J_2 u_2^* + J_3 u_3^* \in su^*(1,1)$ 

- > We note  $O_{\rho}^{+}$  the coadjoint orbit  $Ad_{SU(1,1)}^{*}$  of SU(1,1), given by the upper sheet of the two-sheeted hyperboloid given by previous equation.
- > The orbit method of Kostant-Kirillov-Souriau associates to each of these coadjoint orbits a representation of the discrete series of SU(1,1), provided that  $\rho$  is a half integer greater or equal than 1:  $\rho = k/2, k \in N$  and  $\rho \ge 1$
- > When explicitly executing the Kostant-Kirillov construction, the representation Hilbert spaces  $H_{\rho}$  are realized as closed reproducing kernel subspaces of  $L^2(D, \omega_{\rho})$ . The Kostant-Kirillov-Souriau orbit method shows that to each coadjoint orbit of a connected Lie group is associated a unitary irreducible representation of G

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# Gibbs state Equilibrium

#### One parameter subgroup

- Souriau has observed that action of the full Galilean group on the space of motions of an isolated mechanical system is not related to any equilibrium Gibbs state (the open subset of the Lie algebra, associated to this Gibbs state is empty).
- > The main Souriau idea was to define the Gibbs states for one-parameter subgroups of the Galilean group. We will use the same approach, in this case We will consider action of the Lie group SU(1,1) on the symplectic manifold  $(M,\omega)$  (Poincaré unit disk) and its momentum map J are such that the following open subset is not empty:

$$\Lambda_{\beta} = \left\{ \beta \in \mathfrak{g} / \int_{D} e^{-\langle J(z), \beta \rangle} d\lambda(z) < +\infty \right\}$$

> The idea of Souriau is to consider a one parameter subgroup of SU(1,1). To parametrize elements of SU(1,1) is through its Lie algebra. In the neighborhood of the identity element, the elements of  $g \in SU(1,1)$  can be written as the exponential of an element  $\beta$  of its Lie algebra :  $g = \exp(\epsilon\beta)$  with  $\beta \in \mathfrak{g}$ 

#### Gibbs State Equilibrium

#### One parameter subgroup

> We can then exponentiate  $\beta$  with exponential map to get :

$$g = \exp(\varepsilon\beta) = \sum_{k=0}^{\infty} \frac{(\varepsilon\beta)^k}{k!} = \begin{pmatrix} a_{\varepsilon}(\beta) & b_{\varepsilon}(\beta) \\ b_{\varepsilon}^*(\beta) & a_{\varepsilon}^*(\beta) \end{pmatrix}$$

> If we make the remark that we have the following relation

$$\beta^{2} = \begin{pmatrix} ir & \eta \\ \eta^{*} & -ir \end{pmatrix} \begin{pmatrix} ir & \eta \\ \eta^{*} & -ir \end{pmatrix} = \left( \left| \eta \right|^{2} - r^{2} \right) I$$

> we can developed the exponential map :

$$g = \exp(\varepsilon\beta) = \begin{pmatrix} \cosh(\varepsilon R) + ir\frac{\sinh(\varepsilon R)}{R} & \eta\frac{\sinh(\varepsilon R)}{R} \\ \eta^*\frac{\sinh(\varepsilon R)}{R} & \cosh(\varepsilon R) - ir\frac{\sinh(\varepsilon R)}{R} \end{pmatrix} \text{ with } R^2 = |\eta|^2 - r^2$$

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# Gibbs State Equilibrium

## Equilibrium conditions

- > We can observe that one condition is that  $|\eta|^2 r^2 > 0$
- > Then the subset to consider is given by the subset

$$\Lambda_{\beta} = \left\{ \beta = \begin{pmatrix} ir & \eta \\ \eta^* & -ir \end{pmatrix}, r \in R, \eta \in C / \left| \eta \right|^2 - r^2 > 0 \right\}$$

> such that:  $\int_{D} e^{-\langle J(z),\beta \rangle} d\lambda(z) < +\infty$ 

- > The generalized Gibbs states of the full SU(1,1) group do not exist. However, generalized Gibbs states for the one-parameter subgroups  $\exp(\alpha\beta)$ ,  $\beta \in \Lambda_{\beta}$  of the SU(1,1) group do exist.
- > The generalized Gibbs state associated to  $\beta$  remains invariant under the restriction of the action to the one-parameter subgroup of SU(1,1) generated by  $\exp(\epsilon\beta)$ .

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# Souriau Gibbs density for SU(1,1)

#### Covariant Gibbs density

> We can the write the covariant Gibbs density in the unit disk given by moment map of the Lie group SU(1,1) and geometric temperature in its Lie algebra  $\beta \in \Lambda_{\beta}$ :

$$p_{Gibbs}(z) = \frac{e^{-\langle J(z),\beta \rangle}}{\int_{D} e^{-\langle J(z),\beta \rangle} d\lambda(z)} \text{ with } d\lambda(z) = 2i\rho \frac{dz \wedge dz^{*}}{\left(1 - |z|^{2}\right)^{2}}$$

$$p_{Gibbs}(z) = \frac{e^{-\langle \rho(2\Im bb^{+} - Tr(\Im bb^{+})I),\beta \rangle}}{\int_{D} e^{-\langle J(z),\beta \rangle} d\lambda(z)} = \frac{e^{-\langle \rho\left(\frac{2\Im bb^{+} - Tr(\Im bb^{+})I\right),\beta \rangle}{\left(1 - |z|^{2}\right)^{2} - \frac{1 + |z|^{2}}{(1 - |z|^{2})} - \frac{1 + |z|^{2}}{(1 - |z|^{2})}\right]}{\int_{D} e^{-\langle J(z),\beta \rangle} d\lambda(z)}$$

$$J(z) = \rho\left(\frac{2Mbb^{+} - Tr(Mbb^{+})I}{\int_{D} e^{-\langle J(z),\beta \rangle} d\lambda(z)}\right) \text{ with } M = \begin{bmatrix} 1 & 0\\ 0 & -1 \end{bmatrix} \text{ and } b = \frac{1}{1 - |z|^{2}} \begin{bmatrix} 1\\ -z \end{bmatrix} \text{ THALES}$$

# Souriau Gibbs density

**Covariant Gibbs Density** 

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- > To write the Gibbs density with respect to its statistical moments, we have to express the density with respect to Q = E[J(z)]
- Then, we have to invert the relation between Q and  $\beta$ , to replace  $\beta = \begin{bmatrix} u & u \\ \eta^* & -ir \end{bmatrix} \in \Lambda_{\beta}$ by  $\beta = \Theta^{-1}(Q) \in \mathfrak{g}$  where  $Q = \frac{\partial \Phi(\beta)}{\partial \beta} = \Theta(\beta) \in \mathfrak{g}^*$  with  $\Phi(\beta) = -\log \int_{D} e^{-\langle J(z), \beta \rangle} d\lambda(z)$ deduce from Legendre tranform. The mean moment map is given by:

$$Q = E[J(z)] = E\left[\rho\left(\frac{1+|w|^2}{(1-|w|^2)}, \frac{-2w^*}{(1-|w|^2)}\right) - \frac{2w}{(1-|w|^2)}, \frac{1+|w|^2}{(1-|w|^2)}\right]\right]$$

where  $w \in D$ 

# THALES

# THALES

Representation Theory & Orbits Methods: Fourier Transform for Non-Commutative Harmonic analysis



# Fourier/Laplace Transform and Representation Theory

- Fourier analysis, named after Joseph Fourier, who showed that representing a function as a sum of trigonometric functions greatly simplifies the study of heat transfer and addresses classically commutative harmonic analysis. Classical commutative harmonic analysis is restricted to functions defined on a topological locally compact and Abelian group G (Fourier series when G = R<sup>n</sup>/Z<sup>n</sup>, Fourier transform when G = R<sup>n</sup>, discrete Fourier transform when G is a finite Abelian group).
- The modern development of Fourier analysis during XXth century has explored the generalization of Fourier and Fourier-Plancherel formula for non-commutative harmonic analysis, applied to locally compact non-Abelian groups.
- This has been solved by geometric approaches based on "orbits methods" (Fourier-Plancherel formula for G is given by coadjoint representation of G in dual vector space of its Lie algebra) with many contributors (Dixmier, Kirillov, Bernat, Arnold, Berezin, Kostant, Souriau, Duflo, Guichardet, Torasso, Vergne, Paradan, etc.)

# Dixmier/Kirillov/Duflo/Vergne Representation Theory

#### Classical Commutative Harmonic Analysis

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 $G = T^n = R^n / Z^n$ : Fourier series,  $G = R^n$ : Fourier Transform

G group character (linked to 
$$e^{ikx}$$
): 
$$\begin{cases} \chi : G \to U \\ U = \{ z \in C / |z| = 1 \} \end{cases}$$

$$\hat{G} = \left\{ \chi / \chi_1.\chi_2(g) = \chi_1(g)\chi_2(g) \right\}$$
Fourier Transform
$$\varphi: G \to C \qquad \qquad \hat{\varphi}: \hat{G} \to C$$

$$g \mapsto \varphi(g) = \int_{\hat{G}} \hat{\varphi}(\chi)\chi(g)^{-1}d\chi \qquad \qquad \chi \mapsto \hat{\varphi}(\chi) = \int_{G} \varphi(g)\chi(g)dg$$
Fourier-Plancherel formula
$$\varphi(e) = \int_{\hat{G}} \hat{\varphi}(\chi)d\chi$$
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# Dixmier/Kirillov/Duflo/Vergne Representation Theory

#### **Character-Distribution**

> (Schwarz) Distribution on G :  $\chi_{\rm U}(g) = tr {\rm U}_{\varphi}$  with  ${\rm U}_{\varphi} = \int \varphi(g) {\rm U}_{g} dg$ 

Character Formula: Fourier transform on Lie algebra via Exponential map  $U_{\psi} = \int \psi(X) U_{\exp(X)} dX$ 

Kirillov Character: 
$$\chi_U(\exp(X)) = \operatorname{tr} U_{\exp(X)} = j(X)^{-1} \int_O e^{i\langle f,X \rangle} d\mu_O(f)$$

Fourier Transform:

$$\int_{O} e^{i\langle f, X \rangle} d\mu_{O}(f) = j(X) \operatorname{tr} U_{\exp(X)}$$
$$(X) = \left( \operatorname{det} \left( \frac{e^{ad_{X}/2} - e^{-ad_{X}/2}}{ad_{X}/2} \right) \right)^{1/2}$$

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### **Covariant Gibbs Density by Souriau Thermodynamics**



# THALES

# EXTENSION: GIBBS DENSITY FOR SU(n,n) in Siegel Disk



## Symplectic Group(Carl-Ludwig Siegel) : Siegel Upper half space SH<sub>n</sub>

#### Siegel metric on the Siegel Upper Half Space:

- Siegel Upper half Space:  $SH_n = \{Z = X + iY \in Sym(n, C) / Im(Z) = Y > 0\}$
- > Isometries of  $SH_n$  are given by quotient Group:  $PSp(n,R) \equiv Sp(n,R) / \{\pm I_{2n}\}$  with Sp(n,F) symplectic Group:  $M = \begin{pmatrix} A & B \\ C & D \end{pmatrix} \Longrightarrow M(Z) = (AZ + B)(CZ + D)^{-1}$  $M = \begin{pmatrix} A & B \\ C & D \end{pmatrix} \in Sp(n, F) \Leftrightarrow \begin{cases} A^T C \text{ et } B^T D & \text{symmetric} \\ A^T D - C^T B = I_n \end{cases}$  $Sp(n,F) \equiv \left\{ M \in GL(2n,F) / M^T J M = J \right\}, J = \begin{pmatrix} 0 & I_n \\ -I & 0 \end{pmatrix} \in SL(2n,R)$

Z = X + iY

> Metric invariant by the automorpisms M(Z):

$$ds_{Siegel}^2 = Tr(Y^{-1}(dZ)Y^{-1}(d\overline{Z}))$$

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# Extension of homogeneous bounded symmetric domains: Siegel Upper half-space and Siegel disk



#### Extension of homogeneous Bounded symmetric domains: Siegel Upper half-space and Siegel disk



L.K. Hua







**F.** Berezin

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 $\det(R(Z_1,Z_2)-\lambda J)=0$ 

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#### **Cartan-Siegel Symmetric Homogeneous Bounded Domains**

$$\Omega_{1,1}^{I} = \Omega_{1}^{II} = \Omega_{1}^{III} = \Omega_{1}^{IV} = \{z \in C / zz^{*} < 1\}, K(z, w^{*}) = \frac{1}{(1 - zw^{*})^{2}}$$

Z: Complex Rectangular Matrix

 $ZZ^+ < I$  (+: transposed - conjugate) Type I:  $\Omega_{p,q}^{I}$  complex matrices with p lines and q rows Type II:  $\Omega_p^{II}$  complex symmetric matrices of order p Type III :  $\Omega_p^{III}$  complex skew symmetric matrices of order p Type IV :  $\Omega_n^{IV}$  complex matrices with n rows and 1 line :  $|ZZ^{t}| < 1,1 + |ZZ^{t}|^{2} - 2ZZ^{+} > 0$ 



Henri Poincaré

(n=1)



Elie Cartan (n < = 3)

$$K(Z,W^*) = \frac{1}{\mu(\Omega)} \det(I - ZW^+)^{-\nu} \quad \text{for } \begin{cases} \text{Type I}: \Omega_{p,q}^I, \nu = p + q \\ \text{Type II}: \Omega_p^{II}, \nu = p + 1 \\ \text{Type III}: \Omega_p^{III}, \nu = p - 1 \end{cases}$$
$$K(Z,W^*) = \frac{1}{\mu(\Omega)} (1 + ZZ^*W^*W^+ - 2ZW^*)^{-\nu} \text{ for Type IV}: \Omega_n^{IV}, \nu = n \\ \text{where } \mu(\Omega) \text{ is euclidean volume of the domain.} \end{cases}$$




### From Poincaré Unit Disk to Siegel Unit Disk

- > To extend this approach for covariant Gibbs density on Siegel Unit Disk:  $SD = \{Z \in M_{pq}(C) / I_p - ZZ^+ > 0\}$
- > that is a classical matrix extension of Poincaré unit Disk, we have proposed to consider G = SU(p,q) unitary group and the homogeneous space :

$$G/K = SU(p,q)/S(U(p),U(q))$$
  
with  $K = S(U(p) \times U(q)) = \left\{ \begin{pmatrix} A & 0 \\ 0 & D \end{pmatrix} / A \in U(p), D \in U(q), \det(A) \det(D) = 1 \right\}$ 

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> We can use the following decomposition for  $g \in G^C$ 

$$g = \begin{pmatrix} A & B \\ C & D \end{pmatrix} \in G^{C}, g = \begin{pmatrix} I_{p} & BD^{-1} \\ 0 & I_{q} \end{pmatrix} \begin{pmatrix} A - BD^{-1}C & 0 \\ 0 & D \end{pmatrix} \begin{pmatrix} I_{p} & 0 \\ D^{-1}C & I_{q} \end{pmatrix}$$

> and consider the action of  $g \in G^{c}$  on Siegel Unit Disk given by:

$$SD = \left\{ Z \in M_{pq}(C) / I_{p} - ZZ^{+} > 0 \right\}$$

$$\Rightarrow \left\langle \xi_0, \left[ Z, Z^+ \right] \right\rangle = -2i\lambda \left( p+q \right)^2 Tr \left( ZZ^+ \right), \forall Z \in D$$

> Then, we the equivatiant moment map is given by:  $\forall X \in g^{C}, Z \in D, \ \psi(Z) = Ad^{*}(\exp(-Z^{+})\zeta(\exp Z^{+}\exp Z))\xi_{0}$  $\zeta(\exp Z^{+}\exp Z) = \begin{pmatrix} I_{p} & Z(I_{q}-Z^{+}Z)^{-1} \\ 0 & I_{q} \end{pmatrix}, \ \forall g \in G, Z \in D \text{ then}$ 

Les Hour  $\mathcal{U}_2$  (It  $g_1 \neq Z_2 \neq Ad^*_{\psi}(Z): \psi$  is a diffeomorphism from SD onto orbit  $O(\xi_0)$  THALES Joint Structures and Common Foundation & I Statistical Physics, Information Geometry and Inference for Learning (SPIGL'20)

### From Poincaré Unit Disk to Siegel Unit Disk

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with 
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> The moment map given by:

$$J(Z) = i\lambda \begin{pmatrix} (I_p - ZZ^+)^{-1} (-pZZ^+ - qI_p) & (p+q)Z(I_q - Z^+Z)^{-1} \\ -(p+q)(I_q - Z^+Z)^{-1}Z^+ & (pI_q + qZ^+Z)(I_q - Z^+Z)^{-1} \end{pmatrix}$$

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## THALES

## GIBBS DENSITY FOR SE(2) LIE GROUPS FOR MACHINE LEARNING ON KINEMATICS



## **GIBBS DENSITY FOR SE(2) LIE GROUP**

## Coadjoint action of SE(2)

- > We will consider Souriau model for SE(2) Lie group with non-null cohomology and then with introduction of Souriau one-cocycle.
- > We consider  $SE(2) = SO(2) \times R^2$ :

$$SE(2) = \left\{ \begin{bmatrix} R_{\varphi} & \tau \\ 0 & 1 \end{bmatrix} / R_{\varphi} \in SO(2), \tau \in \mathbb{R}^2 \right\}$$

> The Lie algebra se(2) of SE(2) has underlying vector space  $R^3$  and Lie bracket:

$$(\xi, u) \in se(2) = R \times R^2, \begin{bmatrix} -\xi \mathfrak{F} & u \\ 0 & 0 \end{bmatrix} \in se(2) \text{ with } \mathfrak{F} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

> Coadjoint action of SE(2) is given by:

$$Ad_{(R_{\varphi},\tau)}^{*}(m,\rho) = (m + \Im R_{\varphi}\rho.\tau, R_{\varphi}\rho)$$

## **MOMENT MAP FOR SE(2) LIE GROUP**

## Moment map Computation for SE(2)

> Let  $J_{(\xi,u)}(x): \mathbb{R}^2 \to se^*(2)$  be the moment map of this action relative to the symplectic form, we can compute it from its definition:

$$\begin{split} dJ_{(\xi,u)}(x).y &= -2\omega\big((\xi,u)_{R^2},y\big) \\ \text{with} \quad \omega\big((\xi,u)_{R^2},y\big) &= \omega\big(\xi\Im x - u,y\big) = \big(\xi\Im x - u\big).\Im y = \big(\xi x + \Im u\big).y \\ \Rightarrow dJ_{(\xi,u)}(x).y &= -2\big(\xi x + \Im u\big).y \\ \Rightarrow J_{(\xi,u)}(x) &= -2\bigg(\frac{1}{2}\xi \|x\|^2 + \Im u.x\bigg) = -2\bigg(\frac{1}{2}\|x\|^2, -\Im x\bigg).(\xi,u) \\ J_{(\xi,u)}(x) &= J(x).(\xi,u) \Rightarrow J(x) = -2\bigg(\frac{1}{2}\|x\|^2, -\Im x\bigg) , \ x \in R^2 \end{split}$$

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## SOURIAU MOMENT MAP FOR SE(2)

## Souriau moment map for SE(2)

## > The moment map $J: \mathbb{R}^2 \to se^*(2)$ of SE(2) is defined by: $J_{(\xi,u)}(x) = J(x).(\xi,u)$

> with the right action of SE(2) on  $R^2$ :

$$J_{(\xi,u)}(x) = -2\left(\frac{1}{2}\xi \|x\|^2 + \Im u.x\right) = -2\left(\frac{1}{2}\|x\|^2, -\Im x\right).(\xi, u)$$
$$J_{(\xi,u)}(x) = J(x).(\xi, u) \Longrightarrow J(x) = -2\left(\frac{1}{2}\|x\|^2, -\Im x\right), \ x \in \mathbb{R}^2$$

## SOURIAU COCYCLE FOR SE(2) LIE GROUP & COADJOINT ORBITS

## **SOURIAU Cocycle Computation**

> We then compute the one-cocycle of SE(2) from the moment map

$$\theta\left(\left(R_{\varphi,\tau}\right)\right) = J\left(0.\left(R_{\varphi},\tau\right)\right) - Ad_{\left(R_{\varphi},\tau\right)}^{*}J(0) = J\left(-R_{-\varphi}\tau\right)$$
$$\theta\left(\left(R_{\varphi,\tau}\right)\right) = -2\left(\frac{1}{2}\|\tau\|^{2}, \Im R_{-\varphi}\tau\right) = -2\left(\frac{1}{2}\|\tau\|^{2}, R_{-\varphi-\frac{\pi}{2}}\tau\right)$$

> Coadjoint orbit of SE(2) are generated by:

$$O_{(m,\rho)} = \left\{ Ad_{(R_{\varphi,\tau})}^{*}(m,\rho) + \theta\left(\left(R_{\varphi},\tau\right)\right) / \left(R_{\varphi},\tau\right) \in SE\left(2\right) \right\}$$
$$O_{(m,\rho)} = \left\{ \left( x - R_{-\frac{\pi}{2}}\rho.\tau - \left\|\tau\right\|^{2}, R_{-\varphi}\rho - 2R_{-\varphi-\frac{\pi}{2}}\tau \right) / \left(R_{\varphi},\tau\right) \in SE\left(2\right) \right\}$$

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## SOURIAU-FISHER METRIC FOR SE(2) LIE GROUP

## FISHER Metric in SOURIAU Model for SE(2)

> The KKS 2-form in non-null cohomology case is given by:

$$\omega_{(m,\rho)(m',\rho')} \left( ad_{(\xi,u)}^* (m',\rho') - (0,2\Im u), ad_{(\eta,v)}^* (m',\rho') - (0,2\Im v) \right) = \rho' \cdot (-\xi\Im v + \eta\Im u) + 2u \cdot \Im v$$
  
with  $(m',\rho') = \left( x - R_{-\frac{\pi}{2}} \rho \cdot \tau - \|\tau\|^2, R_{-\varphi} \rho - 2R_{-\varphi-\frac{\pi}{2}} \tau \right) \in \mathcal{O}_{(m,\rho)} \subset \mathbb{R}^3$ 



## **GIBBS DENSITY FOR SE(2) LIE GROUP**

## Souriau Gibbs density for SE(2)

> Considering the symplectic form on  $R^2$  $\omega(\zeta, \upsilon) = \zeta \cdot \Im \upsilon$  with  $\Im = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$ 

> the action of SE(2) is symplectic and admits the momentum map:

$$J(x) = -\left(\frac{1}{2} \|x\|^2, -\Im x\right) , \ x \in R^2$$

► For generalized temperature  $\beta \in \Omega = \{(b,B) \in se(2) / b < 0, B \in R^2\}$ , Souriau Gibbs density is given by :

$$p_{Gibbs}(x) = \frac{e^{-\langle J(x),\beta\rangle}}{\int\limits_{R^2} e^{-\langle J(x),\beta\rangle} d\lambda(x)} = \frac{e^{\frac{1}{2}b\|x\|^2 - B.\Im x}}{\int\limits_{R^2} e^{\frac{1}{2}b\|x\|^2 - B.\Im x} d\lambda(x)}$$

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## Gibbs density for SE(2)

> The Massieu Potential could be computed :

$$\Phi(\beta) = \log \int_{R^2} e^{\frac{1}{2}b\|x\|^2 - B.\Im x} d\lambda(x) = \log\left(-\frac{2\pi}{b}e^{-\frac{1}{2b}\|B\|^2}\right)$$

> By derivation of Massieu potential, we can deduce expression of Heat:

$$Q \in \Omega^* = \left\{ \left(m, M\right) \in se^*(2) / m + \frac{\left\|M\right\|^2}{2} < 0 \right\}; Q = \frac{\partial \Phi(\beta)}{\partial \beta} = \left(\frac{1}{b} - \frac{\left\|B\right\|^2}{2b^2}, \frac{1}{b}B\right) = \Theta(\beta)$$

> We can the inverse this relation to express generalized temperature with respect to the heat:  $\beta = \Theta^{-1}(Q) = \left( \left( m + \frac{1}{2} \|M\|^2 \right)^{-1}, \left( m + \frac{1}{2} \|M\|^2 \right)^{-1} M \right)$ 

We can the express the Gibbs density with respect to the Heat Q which is the mean of moment map:

$$p_{Gibbs}\left(x\right) = \frac{e^{\left(m + \frac{1}{2}\|M\|^{2}\right)}}{\left(m + \frac{1}{2}\|M\|^{2}\right)}}$$

$$\lim_{Les Houches 27th-31^{H} July} \frac{1}{L^{20}}$$

$$\lim_{Joint Structures and Common Foundations of Statistical Physics, Information Geometry and Inference for Learning (SPIGL'20)} \frac{\frac{1}{2}\|x\|^{2} - M.3x}{P} d\lambda(x) \text{ with } (m, M) = E(J(x)) = \left[-E(\|x\|^{2}), 2\Im E(x)\right]$$

$$\lim_{Les Houches 27th-31^{H} July} \frac{1}{L^{20}} d\lambda(x) \text{ with } (m, M) = E(J(x)) = \left[-E(\|x\|^{2}), 2\Im E(x)\right]$$

$$\prod_{Les Houches 27th-31^{H} July} \frac{1}{L^{20}} d\lambda(x) \text{ with } (m, M) = E(J(x)) = \left[-E(\|x\|^{2}), 2\Im E(x)\right]$$

## Gibbs density for SE(2)

## Souriau Covariant Gibbs density for SE(2)

$$p_{Gibbs}(x) = \frac{e^{\frac{\frac{1}{2}\|x\|^2 + 2E(x).Ix}{\left(-E\left(\|x\|^2\right) + 2\|E(x)\|^2\right)}}}{\int\limits_{R^2} e^{\frac{\frac{1}{2}\|x\|^2 + 2E(x).Ix}{\left(-E\left(\|x\|^2\right) + 2\|E(x)\|^2\right)}} d\lambda(x)}$$

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## Fisher Metric for SE(2)

> Entropy is given by:

$$S(Q) = \langle Q, \beta \rangle - \Phi(\beta) = 1 + \log(2\pi) + \log\left(-m - \frac{\|M\|^2}{2}\right)$$

- Fisher Metric is given by:  $I_{Fisher}\left(Q\right) = \left(m + \frac{1}{2} \|M\|^{2}\right)^{-1} \begin{bmatrix}I & M^{T} \\ M^{T} & \frac{1}{2} \|M\|^{2} m\end{bmatrix}$ With  $\left(m, M\right) = E\left(J(x)\right) = E\left[-2\left(\frac{1}{2} \|x\|^{2}, -\Im x\right)\right] = \left[-E\left(\|x\|^{2}\right), 2\Im E(x)\right]$
- > Fisher Metric with respect to moments

$$I_{Fisher}(Q) = \left(2\|E(x)\|^{2} - E(\|x\|^{2})\right)^{-1} \begin{bmatrix} I & \left(2\Im E(x)\right)^{T} \\ 2\Im E(x) & 2\|E(x)\|^{2} + E(\|x\|^{2}) \end{bmatrix}$$

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## THALES

## **Conclusion & perspectives**



## **Towards Lie Group & Symplectic Machine Learning**



## Supervised & Non-Supervised Learning on Lie Groups

#### Geodesic Natural Gradient on Lie Algebra

Extension of Neural Network Natural Gradient from Information Geometry on Lie Algebra for Lie Groups Machine Learning

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## Souriau-Fisher Metric

Extension of Fisher Metric for Lie Group through homogeneous Symplectic Manifolds on Lie Group Co-Adjoint Orbits

### Souriau Exponential Map on Lie Algebra

Exponential Map for Geodesic Natural Gradient on Lie Algebra based on Souriau Algorithm for Matrix Characteristic Polynomial

### Souriau Maximum Entropy Density on Co-Adjoint Orbits

Covariant Maximum Entropy Probability Density for Lie Groups defined with Souriau Moment Map, Co-Adjoint Orbits Method & Kirillov Representation Theory

Symplectic Integrator preserving Moment Map

> Extension of Neural Network Natural Gradient to Geometric Integrators as Symplectic integrators that preserve moment map

### IE GROUP SUPERVISED LEARNING

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Joint Structures and Common Foundations of Statistical Physi Information Geometry and Inference for Learning (SPIGL'20) Lie Group Machine Learning

### Fréchet Geodesic Barycenter by Hermann Karcher Flow

Extension of Mean/Median on Lie Group by Fréchet Definition of Geodesic Barycenter on Souriau-Fisher Metric Space, solved by Karcher Flow

### Mean-Shift on Lie Groups with Souriau-Fisher Distance

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Extension of Mean-Shift for Homogeneous Symplectic Manifold and Souriau-Fisher Metric Space

### LIE GROUP NON-SUPERVISED LEARNING



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## **Rational to Use Lie Groups for THALES Machine Learning Applications**



## Esprit de finesse et esprit de géométrie



Pour la théorie de la connaissance mais aussi pour les sciences est fondamentale la notion de perspective.

Or, les expériences faites dans la géométrie algébriques, dans la théorie des nombres, et dans l'algèbre abstraite m'induisent à tenter une formulation mathématique de cette notion pour surmonter ainsi au moyen de raisonnements d'origine géométrique la géométrie. Il me semble en effet, que la tendance vers l'abstraction observée dans les mathématiques d'aujourd'hui, loin d'être l'ennemi de l'intuition ait le sens profond de quitter l'intuition pour la faire renaitre dans une alliance entre « esprit de géométrie » et « esprit de finesse », alliance rendue possible par les réserves énormes des mathématiques pures dont Pascal et Goethe ne pouvaient pas encore se douter.





## References







### Lie Group Machine Learning and Lie Group Structure Preserving Integrators

Guest Editors:

#### Message from the Guest Editors

Frédéric Barbaresco frederic.barbaresco@ thalesgroup.com

Prof. Elena Cellodoni elena.celledoni@ntnu.no

Prof. François Gay-Balmaz francois.gay-balmaz@ Imd.ens.fr

Prof. Joël Bensoam bensoam@ircam.fr

Deadline for manuscript submissions: 6 January 2020 Machine/deep learning explores use-case extensions for more abstract spaces as graphs and differential manifolds. Recent fruitful exchanges between geometric science of information and Lie group theory have opened new perspectives to extend machine learning on Lie groups to develop new schemes for processing structured data.

Structure-preserving integrators that preserve the Lie group structure have been studied from many points of view and with several extensions to a wide range of situations. Structure-preserving integrators are numerical algorithms that are specifically designed to preserve the geometric properties of the flow of the differential equation such as invariants, (multi)symplecticity, volume preservation, as well as the configuration manifold. They also naturally find applications in the extension of machine learning and deep learning algorithms to Lie group data.

This Special Issue will collect long versions of papers from contributions presented during the GSI'19 conference, but it will be not limited to these authors and is open to international communities involved in research on Lie group machine learning and Lie group structure-preserving integrators.

### Special Issue "Lie Group Machine Learning and Lie Group Structure Preserving Integrators"

### Keywords

- · Lie groups machine learning
- orbits method
- symplectic geometry
- · geometric integrator
- symplectic integrator
- Hamilton's variational principle

https://www.mdpi.com/journal/entropy/special\_issues/Lie\_group





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## SOURIAU 2019

**SOURIAU 2019** 

J.M. Souriau Interview: https://www.youtube.com/w atch?v=uz69vWHXzWY

- Internet website : <u>http://souriau2019.fr</u>
- In 1969, 50 years ago, Jean-Marie Souriau published the book "Structure des système dynamiques", in which using the ideas of J.L. Lagrange, he formalized the "Geometric Mechanics" in its modern form based on Symplectic Geometry
- Chapter IV was dedicated to "Thermodynamics of Lie groups" (ref André Blanc-Lapierre)
- Testimony of Jean-Pierre Bourguignon at Souriau'19 (IHES, director of the European ERC)



Jean-Marie SOURIAU and Symplectic Geometry

Jean-Pierre BOURGUIGNON (CNRS-IHÉS)

## SOURIAU 2019

Conference May 27-31 2019, Paris-Diderot University

#### https://www.youtube.com/watch?v=beM2pUK1H7o



Frédéric Barbaresco **Deniel Benneguin** Jean-Pierre Bourguignor Plant Cartist Dan Christenser Maurice Courtrage Thibeuit Denour Paul Donato Paolo Giordano Seing Gürer Patrick (glenne-Zerminous theil Cambon Jean-Fierte Mathot Yeattle Kosman-Schwartzbach Marc Lachiete-Res Martin Processed Elisti Prato. Un Schweiber lash, larings Szone injurt Robarid Triay Jordan Watte Emin'Wu San Wil Moter Alan Weinsteir

#### JEAN-MARIE SOURIAU

DIDERO

In 1969, the groundbreaking book of Jean-Marie Souriau appeared "Structure des Systèmes Dynamiques". We will celebrate, in 2019, the jubilee of its publication, with a conference in honour of the work of this great scientist.

Symplectic Mechanics, Geometric Quantization, Relativity, Thermodynamics, Cosmology, Diffeology & Philosophy



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## FGSI'19 Cartan-Koszul-Souriau Foundations of Geometric Structures of Information

A seminar on Topological and Geometrical Structures of Information has been organized at CIRM in 2017, to gather

engineers, applied and pure mathematicians interested in the geometry of information. This year FGSI'19 conference



Foundations of Geometric Structures of Information



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SYMPLECTIC GEOMETRY IN PHYSICS



Anton ALEKSEEV (Geneva Univ.) Dmitri ALEKSEEVSKY (Moscow IITP) John BAEZ (Riverside UC) Michel BRION (Grenoble Univ.) Misha GROMOV (Paris IHES)

Panel sessions:

Patrick IGLESIAS-ZEMMOUR (Aix-Marseille Univ.) Yann OLLIVIER (Paris Facebook) Vasily PESTUN (Paris IHES) Aissa WADE (Penn State Univ.)

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### https://fgsi2019.sciencesconf.org/

4-6 Feb 2019 Montpellier (France)

Koszul - Souriau and their influence on the field

PRESENTATION



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Information Geometry and Inference for Learning (SPIGL'20)

#### OPEN

## CIRM Seminar, August 2017 TGSI'17 « Topological & Geometrical Structures of Information »

TGSI'17 videos & slides http://forum.cs-dc.org/category/94/tgsi2017

Special Issue **"Topological and Geometrical Structure of Information**", Selected Papers from CIRM conferences 2017"

http://www.mdpi.com/journal/entropy/speci al\_issues/topological\_geometrical\_info





TOPOLOGICAL & GEOMETRICAL STRUCTURES OF INFORMATION

AUGUST 28<sup>TH</sup> - SEPTEMBER 1<sup>ST</sup> 2017 CIRM - MARSEILLE - FRANCE



Talk on Koszul-Souriau Characteristic Function:

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Jean-Louis Koszul was foreign member of São Paulo Academia of Sciences

## Jean Louis Koszul Lectures at Sao Paulo:

- Faisceaux et Cohomologie
- Variétés Kählériennes
- Exposés sur les espaces homogènes

# symétriques

São Paulo Journal of Mathematical Sciences

## Sao Paulo Journal of Mathematical Sciences SPRINGER Editor-in-Chief: Claudio Gorodski

Inference for Learning (SPIGE 20) CS/JOUINAI/40863

### 2nd Workshop São Paulo Journal of Mathematical Sciences



Round-table with the Editorial Board of the São Paulo Journal of Mathematics





\* To be confirmed

EXPOSES SUR LES ESPACES

OMOGENES SYMETRIOUES

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## **GSI'13 Mines ParisTech**

Slides : https://www.see.asso.fr/gsi2013



## **GSI'17 Mines ParisTech**

Videos: <u>https://www.youtube.com/channel/UCnE9-LbfFRqtaes49cN2DVg/videos</u> UNITWIN website (slides & videos): http://forum.cs-dc.org/category/135/gsi2017



## **GSI'15 Ecole Polvtechnique**

Videos:

https://www.youtube.com/channel/UC5HHo1jbQXusNQzU1iekaGA UNITWIN website (slides & videos): http://forum.cs-dc.org/category/90/gsi2015



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https://www.youtube.com/watch?v=v7QuYuoyLQc&t=5935s (at time 01h35min)





## THALES



## Main references



## Koszul Book on Souriau Work

### Jean-Louis Koszul · Yiming Zou Introduction to Symplectic Geometry Forewords by Michel Nguiffo Boyom, Frédéric Barbaresco and Charles-Michel Marle

This introductory book offers a unique and unified overview of symplectic geometry, highlighting the differential properties of symplectic manifolds. It consists of six chapters: Some Algebra Basics, Symplectic Manifolds, Cotangent Bundles, Symplectic G-spaces, Poisson Manifolds, and A Graded Case, concluding with a discussion of the differential properties of graded symplectic manifolds of dimensions (o,n). It is a useful reference resource for students and researchers interested in geometry, group theory, analysis and differential equations.

Les Houches 27th-31<sup>st</sup> July 2020 Joint Structures and Common Foundations of Statistical Physics, Information Geometry and Inference for Learning (SPIGL'20) Jean-Louis Koszul Yiming Zou

## Introduction to Symplectic Geometry

$$\begin{split} & \left[ \mu \left[ \mathfrak{h} \mathfrak{g} \right] = \mathfrak{s} \mu \left( \mathfrak{s} \right) = A d^2 \mathfrak{s} \mathfrak{s} \left[ \mu \left[ \mathfrak{s} \right] + \mathfrak{g}_{\mu} \left( \mathfrak{s} \right) \right], \quad \forall \ \mathfrak{s} \in G, \mathfrak{s} \in M \\ & c_{\mu} \left( \mathfrak{a}, b \right) = \left\{ \left( \mu, \mathfrak{a} \right), \left( \mu, b \right) \right\} - \left( \mu, \left( \mathfrak{a}, b \right) \right) = \left\{ \mathfrak{g} \mathfrak{g}_{\mu} \left( \mathfrak{a}, b \right) \right\}, \quad \forall \ \mathfrak{a}, b \in \mathfrak{g}. \end{split}$$

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Les Houches 27th-31<sup>st</sup> July 2020 Joint Structures and Common Foundations of Statistical Physi Information Geometry and Inference for Learning (SPIGL'20)

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Joseph Fourier 250<sup>th</sup> Birthday Modern Fourier Analysis and Fourier Heat Equation in Information Sciences for the XXIst Century

> Edited by Frédéric Barbaresco and Jean-Pierre Gazeau Printed Edition of the Special Issue Published in Entropy

> > MDPI

## Jean-Marie Souriau Geometric Theory of Heat, 250 years after Joseph Fourier

## MDPI Entropy Book for Joseph Fourier 250th Birthday

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## **Charles-Michel Marle Books**



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## Référence Book: Gery de Saxcé & Claude Vallée



MECHANICAL ENGINEERING AND SOLID MECHANICS

**Galilean Mechanics** and Thermodynamics of Continua

Géry de Saxcé and Claude Vallée



This title proposes a unified approach to continuum mechanics which is consistent with Galilean relativity. Based on the notion of affine tensors, a simple generalization of the classical tensors, this approach allows gathering the usual mechanical entities - mass, energy, force, moment, stresses, linear and angular momentum — in a single tensor.

Starting with the basic subjects, and continuing through to the most advanced topics, the authors' presentation is progressive, inductive and bottom-up. They begin with the concept of an affine tensor, a natural extension of the classical tensors. The simplest types of affine tensors are the points of an affine space and the affine functions on this space, but there are more complex ones which are relevant for mechanics – torsors and momenta. The essential point is to derive the balance equations of a continuum from a unique principle which claims that these tensors are affine-divergence free.

https://www.wiley.com/en-

us/Galilean+Mechanics+and+Thermodynamics+of+Conti nua-p-9781848216426

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## **Poisson Geometry**



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Yvette Kosmann-Schwarzbach Histoire des Mathématiques et des Sciences physiques

## Siméon-Denis Poisson

Les mathématiques au service de la science Grundlehren der mathematischen Wissenschaften 347 A Series of Comprehensive Studies in Mathematics

Camille Laurent-Gengoux Anne Pichereau Pol Vanhaecke

## **Poisson Structures**



### Camille Laurent-Gengoux

Description Springer

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Les Houches 27th-31<sup>st</sup> July 2020 Joint Structures and Common Foundations of Statistical Physics, Information Geometry and Inference for Learning (SPIGL'20)



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Signals and Communication Technology

Frank Nielsen Editor

Geometric Structures of Information

## Geometric Structures of Information

https://www.springer.com/us/book/978303002 5199

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- Barbaresco, F., Jean-Louis Koszul and the Elementary Structures of Information Geometry, Geometric Structures of Information, pp 333-392, SPRINGER, 2018
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# **GSI'17 Springer Proceedings:**

### **GSI'19 Springer Proceedings:**

#### **Geometric Science** of Information

4th International Conference, 451 2919 Toulouse, France, August 27-29, 2019 Proceedings

NCS 11712



#### **Geometric Science** of Information

Ihird International Conference, GSI 2017 Paris, France, Resember 7-8, 2017 Proceeding



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NGS 9389 of Information Second International Conference, GSI 2015 Falalaese, France, October 28-38, 2011 Proceedings

**GSI'15 Springer** 

**Proceedings:** 



**Geometric Science** 

## THALES

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## Seminal Work of Muriel Casalis supervised by Gérard Letac

International National Review (1991), 59, 2, pp. 341–382. Printed in Grout Relation © International Statistical Institute

#### Familles Exponentielles Naturelles sur $\mathbb{R}^d$ Invariantes par un Groupe

#### **Muriel Casalis**

Laboratoire de Statistique, Université Paul Sabatier, 118, route de Narbonne, 31062 Toulouse Cedex, France

#### Résumé

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La caractérisation des familles exponentielles naturelles de R<sup>el</sup> préservées par un groupe d'affinités donad est faite dans toon case celui d'un groupe compact, en particulier du groupe des rotationes, selui du groupe hypertolique et enfin celui d'un groupe quéclessations. La démarche adoptée consiste à traduire la propriété d'invariance de la famille par une propriété portant sur les mesures qui l'engeadrent pais à rechercher es demaires en conséquence.

#### **1** Introduction

Il est courant en statistique d'envisager un modèle  $(\Omega, sf, (P_R)_{R \in \mathbb{N}})$  tel qu'il existe un groupe G de permutations de  $\Omega$  préservant globalement la famille de probabilités  $F = \{P_n, 0 \in \mathbb{O}\}$ , c'est-àdire que pour tout (0, g) de  $\Theta \times G$ , l'image  $g(P_n)$  de  $P_n$  ar g est encore dans F (Barndorff-Nielsen parle alors de modèle de transformations). On pourra consulter Barndorff-Nielsen et al. (1982) et plus récemment le livre de Barndorff-Nielsen (1988).

Un exemple célèbre est celui des distributions de Fisher-Von-Mises pour lequel  $\Omega$  est la sphère unité de l'espace euclidien E.

 $P_{\theta}(dx) = L(\theta)^{-1} \exp{(\theta, x)} \sigma(dx),$ 

 $\sigma$  désignant la probabilité uniforme sur  $\Omega$  et  $L(\theta)$  le coefficient de normalisation, et pour lequel G est le groupe des rotations 0(E) de E.

Dans cet exemple,  $\{P_{\sigma}, \theta \in \Theta\}$  est une familie exponentielle naturelle au sens suivant. Soit E un espace vectoriel de dimension finie,  $E^{*}$  son dual et si  $(\theta, x)$  est dans  $E^{*} \times E$ ,  $(\theta, x)$  désigne le crochet de dualité; soit, de plus,  $\mu$  une mesure de Radon positive sur E; on note alors  $L_{\alpha}$  la transformée de Laplace de  $\mu$  définie par:

$$L_{\mu}: E^* \rightarrow [0, \infty]: \theta \rightarrow [\exp(\theta, x)\mu(dx);$$

 $D_{\mu}$  est l'ensemble { $\theta \in E^*$ ,  $L_{\mu}(\theta) < \infty$ },  $\Theta(\mu)$  son intérieur et  $k_{\mu}$  la fonction définie sur  $\Theta(\mu)$  par:

 $k_{\mu}(\theta) = \text{Log } L_{\mu}(\theta).$ 

(1.1)

On désigne aussi par  $\mathcal{M}(E)$  l'ensemble des mesures de Radon  $\mu$  positives telles que:

(i) μ n'est pas concentrée sur un sous-espace affine strict de E;
 (ii) Θ(μ) est non vide.

N° d'ordre 679

#### THESE

présentée à

L'UNIVERSITE PAUL SABATIER DE TOULOUSE (SCIENCES)

pour obtenir

DOCTORAT DE L'UNIVERSITE PAUL SABATIER Spécialité : MATHEMATIQUES APPLIQUEES

per

Muriel BONNEFOY - CASALIS

FAMILLES EXPONENTIELLES NATURELLES

INVARIANTES PAR UN GROUPE

Soutenue le 11 Juin 1990, devant la Commission d'Examen :

MM.	H. CAUSSINUS	Professeur à l'Université Paul Sabatier
	D. BAKRY	Professeur à l'Université Paul Sabatier
	J. FARAUT	Professeur à l'Université PARIS VI
	Y. GUIVARC'H	Professeur à l'Université PARIS VI
	G. LETAC	Professeur à l'Université Paul Sabatier

Laboratoire de Statistique et Probabilités UNIVERSITE PAUL SABATIER





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https://www.mdpi.com/journal/entropy/special\_issues/Lie\_group

Les Houches Summer Week, Joint Structures and Common Foundation of Statistical Physics, Information Geometry and Inference for Learning (SPIGL'20); 26th July to 31st July 2020; https://franknielsen.github.io/SPIG-LesHouches2020/

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- Workshop IRT SystemX, Approches topologiques et géométriques pour l'apprentissage statistique, théorie et pratique ; 2020; <u>https://www.irt-</u> <u>systemx.fr/evenements/workshop-approches-topologiques-et-geometriques-</u> <u>tes Honoropour-fapprentissage-statistique-theorie-et-pratique/</u>



# Souriau Entropy Definition



## Entropy Definition by Jean-Marie Souriau (1/4)

Let E a vector space of finite size,  $\mu$  a measure on the dual space  $E^*$ , then the function given by:  $\alpha \mapsto \int_{*} e^{M\alpha} \mu(M) dM$ 

for all  $\alpha \in E$  such that the integral is convergent.

This function is called Laplace Transform. This transform F of the measure  $\mu$  is differentiable inside its definition set def(F). Its p-th derivables are given by the following convergent integrals :

$$F^{(p)}(\alpha) = \int_{E^*} M \otimes M \dots \otimes M \mu(M) dM$$

## Souriau Theorem:

Let E a vector space of finite size,  $\mu$  a non zero positive measure of its dual space  $E^*$ , F its Laplace transform, then:

- *F* is a semi-definite convex function,  $F(\alpha) > 0, \forall \alpha \in def(F)$
- $f = \log(F)$  is convex and semi-continuous
- Let  $\alpha$  an interior point of def(F) then:

$$- D^2(f)(\alpha) \ge 0$$

$$- D^{2}(f)(\alpha) = \int_{E^{*}} e^{M\alpha} \left[ M - D(f)(\alpha) \right]^{\otimes^{2}} \mu(M) dM$$

-  $D^2(f)(\alpha)$  inversible  $\iff$  affine Enveloppe (support ( $\mu$ )) =  $E^*$ 

## Entropy Definition by Jean-Marie Souriau (3/4)

### Lemme:

> Let X a locally compact space, Let  $\lambda$  a positive measure of X, with X as support, then the following function  $\Phi$  is convex:

$$\Phi(h) = \log \int_{X} e^{h(X)} \lambda(x) dx , \ \forall h \in C(X)$$

such that the integral is convergent.

## Proof:

- The integral is strictly positive when its converges, insuring existence of its logarithm
- > Epigraph  $\Phi$  is the set of  $\binom{h}{y}$  such that  $\int_{X} e^{h(x)-y} \lambda(x) dx \le 1$ .
- > Convexity of exponential prove that this epigraph is convex.



# THALES

## Balian Computation of Gibbs Density for Dynamical Centrifuge System



- Balian has computed the Boltzmann-Gibbs distribution without knowing Souriau equations. Exercice 7b of :
  - Balian, R. From Microphysics to Macrophysics, 2nd ed.; Springer: Berlin, Germany, 2007; Volume I
- > Balian started by considering the constants of motion that are the energy and the component  $J_z$  of the total angular momentum:  $J = \sum (r_i \times p_i)$
- > Balian observed that he must add to the Lagrangian parameter, given by (Planck) temperature  $\beta$  for energy, an additional one associated with  $J_{z}$ . He identifies this additional multiplier with  $-\beta\omega$  by evaluating the mean velocity at each point.
- He then introduced the same results also by changing the frame of reference, the Lagrangian and the Hamiltonian in the rotating frame and by writing down the canonical equilibrium in that frame. He uses the resulting distribution to find, through integration, over the momenta, an expression for the particles density as the function of the distance from the cylinder axis.

Joint Structures and Common Foundations of Statistical Physics, Information Geometry and Inference for Learning (SPIGL'20)

> The fluid carried along by the walls of the rotating vessel acquires a non-vanishing average angular momentum  $\langle J_z \rangle$  around the axis of rotation, that is a constant of motion. In order to be able to assign to it a definite value, Balian proposed to associate with it a Lagrangian multiplier  $\lambda$ , in exactly the same way as we classically associate the multiplier  $\beta$  with the energy in canonical equilibrium. The average  $\langle J_z \rangle$  will be a function of  $\lambda$ . The Gibbs density for rotating gas is given by Balian as:

$$D = \frac{1}{Z} e^{-\beta H - \lambda J_z} = \frac{1}{Z} \exp\left\{\sum_{i} \left[\frac{\beta p_i^2}{2m} + \lambda \left(x_i p_{y_i} - y_i p_{x_i}\right)\right]\right\}$$

> With the energy and the average angular momentum given by:

$$U = -\frac{\partial \ln Z}{\partial \beta} = \frac{1}{kT} \qquad \langle J_z \rangle = -\frac{\partial \ln Z}{\partial \lambda}$$

Les Houches 27th-31st July 2020 Joint Structures and Common Foundations of Statistical Physic: Information Geometry and Inference for Learning (SPIGL'20)

> The Lagrangian parameter  $\lambda$  has a mechanical nature. To identify this parameter, Balian compared microscopic and macroscopy descriptions of fluid mechanics. He described the single-particle reduced density by:

$$f(r,p) \propto \exp\left\{-\frac{\beta p^2}{2m} - \lambda \left(xp_y - yp_x\right)\right\} = \exp\left\{-\frac{\beta}{2m} \left(p + \frac{m}{\beta} \left[\lambda \times r\right]\right)^2 + \frac{m\lambda^2}{2\beta} \left(x^2 + y^2\right)\right\}$$

Whence Balian finds the velocity distribution at a point r to be proportional to:

$$\exp\left\{-\frac{m}{2kT}\left(v+\frac{1}{\beta}\left[\lambda\times r\right]\right)^{2}\right\}$$

The mean velocity of the fluid at the point r is equal to:  $\langle v \rangle = -\frac{1}{\beta} [\lambda \times r]$ 

and can be identified with the velocity  $\begin{bmatrix} \omega \times r \end{bmatrix}$  in an uniform rotation with angular velocity  $\omega$ . By comparison, Balian put :  $\omega = -\lambda/\beta$ Use Houches 27th-31<sup>st</sup> July 2020
Use Houches 27th-31<sup>st</sup> July 2020
THALES

Balian made the remarks that "The angular momentum is imparted to the gas when the molecules collide with the rotating walls, which changes the Maxwell distribution at every point, shifting its origin. The walls play the role of an angular momentum reservoir. Their motion is characterized by a certain angular velocity, and the angular velocities of the fluid and of the walls become equal at equilibrium, exactly like the equalization of the temperature through energy exchanges".

# THALES

## Compliance with Symplectic Model of Thermodynamics by Balian-Valentin



## **Compatible Balian Gauge Theory of Thermodynamics**

Entropy S is an extensive variable  $q^0 = S(q^1,...,q^n)$  depending on  $q^i$  (i=1,...,n)n independent extensive/conservative quantities characterizing the system

The n intensive variables  $\gamma_i$  are defined as the partial derivatives:  $\gamma_i = \frac{\partial S(q^1, ..., q^n)}{\partial q^i}$ 

Balian has introduced a non-vanishing gauge variable which multiplies all the intensive variables, defining a new set of variables:  $p_i = -p_0.\gamma_i$ , i = 1,...,n

The 2n+1-dimensional space is thereby extended into a 2n+2-dimensional thermodynamic space T spanned by the variables  $p_i$ ,  $q^i$  with i = 0,1,...,n, where the physical system is associated with a n+1-dimensional manifold M in T, parameterized for instance by the coordinates  $q^1,...,q^n$  and  $p_0$ .

## **Compatible Balian Gauge Theory of Thermodynamics**

the contact structure in 2n+1 dimension:  $\widetilde{\omega} = dq^0 - \sum_{i=1}^{n} \gamma_i dq^i$ 

is embedded into a symplectic structure in 2n+2 dimension, with 1-form, as symplectization:  $\omega = \sum_{i=0}^{n} p_i dq^i$ 

The n + 1-dimensional thermodynamic manifolds M are characterized by

- :  $\omega = 0$  . The 1-form induces then a symplectic structure on T :  $d\omega = \sum dp_i \wedge dq^i$
- The concavity of the entropy  $S(q^1,...,q^n)$ , as function of the extensive variables, expresses the stability of equilibrium states. It entails the existence of a metric structure in the n-dimensional space  $q_i$ :  $2^2 \sigma$

$$ds^{2} = -d^{2}S = -\sum_{i,j=1}^{n} \frac{O S}{\partial q^{i} \partial q^{j}} dq^{i} dq^{j}$$

which defines a distance between two neighboring thermodynamic states:

$$d\gamma_{i} = \sum_{j=1}^{n} \frac{\partial^{2}S}{\partial q^{i} \partial q^{j}} dq^{j}$$

$$ds^{2} = -\sum_{i=1}^{n} d\gamma_{i} dq_{i} = \frac{1}{p_{0}} \sum_{i=0}^{n} dp_{i} dq^{i}$$

$$ds^{2} = -\sum_{i=1}^{n} d\gamma_{i} dq_{i} = \frac{1}{p_{0}} \sum_{i=0}^{n} dp_{i} dq^{i}$$
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## **Compatible Balian Gauge Theory of Thermodynamics**

We can observe that this Gauge Theory of Thermodynamics is compatible with Souriau Lie Group Thermodynamics, where we have to consider the Souriau vector :

$$\beta = \begin{bmatrix} \gamma_1 \\ \vdots \\ \gamma_n \end{bmatrix} \text{ transformed in a new vector } p_i = -p_0.\gamma_i$$

$$\begin{bmatrix} -p_0\gamma_1 \end{bmatrix}$$

$$p = \begin{bmatrix} p_0 \gamma_1 \\ \vdots \\ -p_0 \gamma_n \end{bmatrix} = -p_0 . \beta$$

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THALE

# THALES

## Souriau Model & Multivariate Gaussian Model



### Multivariate Gaussian Density as 1st order Maximum Entropy in Souriau Book (Chapter IV)

Exemple : (loi normale) :

Prenons le cas 
$$V = R^n$$
,  $\lambda =$  mesure de Lebesgue,  $\Psi(x) \equiv$ 

$$\begin{pmatrix} x \\ x \otimes x \end{pmatrix};$$

un élément Z du dual de E peut se définir par la formule

$$Z(\Psi(x)) \equiv \overline{a} \cdot x + \frac{1}{2} \overline{x} \cdot H \cdot x$$

 $[a \in R^*; H = \text{matrice symétrique}]$ . On vérifie que la convergence de l'intégrale  $I_0$  a lieu si la matrice H est positive (<sup>1</sup>); dans ce cas la loi de Gibbs s'appelle *loi normale de Gauss*; on calcule facilement  $I_0$  en faisant le changement de variable  $x^* = H^{1/2} x + H^{-1/2} a$  (<sup>2</sup>); il vient

 $z = \frac{1}{2} \left[ \overline{a} \cdot H^{-1} \cdot a - \log \left( \det \left( H \right) \right) + n \log \left( 2 \pi \right) \right]$ 

alors la convergence de  $I_1$  a lieu également; on peut donc calculer M, qui est défini par les moments du premier et du second ordre de la loi (16.196); le calcul montre que le moment du premier ordre est égal à  $-H^{-1}$ . a et que les composantes du tenseur variance (16.196) sont égales aux éléments de la matrice  $H^{-1}$ ; le moment du second ordre s'en déduit immédiatement.

La formule (16.2009) donne l'entropie :

$$s = \frac{n}{2}\log(2 \pi e) - \frac{1}{2}\log(\det(H))$$
;

(<sup>4</sup>) Voir Calcul linéaire, tome II.
 (<sup>2</sup>) C'est-à-dire en recherchant l'image de la loi par l'application x → x\*.

DÉPARTEMENT MATHÉMATIQUE Dirigi par le Professeur P. LELONG

### STRUCTURE <sup>des</sup> SYSTÈMES DYNAMIQUES

Maîtrises de mathématiques

J.-M. SOURIAG

DUNOD

http://www.jmsouriau.com/structure des\_systemes\_dynamiques.htm

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## Example of Multivariate Gaussian Law (real case)

Multivariate Gaussian law parameterized by moments

$$p_{\xi}(\xi) = \frac{1}{(2\pi)^{n/2} \det(R)^{1/2}} e^{-\frac{1}{2}(z-m)^{T}R^{-1}(z-m)}$$

$$\frac{1}{2}(z-m)^{T}R^{-1}(z-m) = \frac{1}{2} \begin{bmatrix} z^{T}R^{-1}z - m^{T}R^{-1}z - z^{T}R^{-1}m + m^{T}R^{-1}m \end{bmatrix}$$

$$= \frac{1}{2} z^{T}R^{-1}z - m^{T}R^{-1}z + \frac{1}{2}m^{T}R^{-1}m$$

$$p_{\xi}(\xi) = \frac{1}{(2\pi)^{n/2} \det(R)^{1/2}} e^{\frac{1}{2}m^{T}R^{-1}m} e^{-\left[-m^{T}R^{-1}z + \frac{1}{2}z^{T}R^{-1}z\right]} = \frac{1}{Z} e^{-\langle\xi,\beta\rangle}$$

$$p_{\xi}(\xi) = \frac{1}{(2\pi)^{n/2} \det(R)^{1/2}} e^{\frac{1}{2}m^{T}R^{-1}m} e^{-\left[-m^{T}R^{-1}z + \frac{1}{2}z^{T}R^{-1}z\right]} = \frac{1}{Z} e^{-\langle\xi,\beta\rangle}$$

$$\xi = \begin{bmatrix} z \\ zz^{T} \end{bmatrix} \text{ and } \beta = \begin{bmatrix} -R^{-1}m \\ \frac{1}{2}R^{-1} \end{bmatrix} = \begin{bmatrix} a \\ H \end{bmatrix} \text{ with } \langle\xi,\beta\rangle = a^{T}z + z^{T}Hz = Tr[za^{T} + H^{T}zz^{T}]$$

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THALES

## Information Geometry for Multivariate Gaussian Density

For multivariate gaussian density of meam m and covariance matrix Rclassical parameterization is given by:

$$p_{m,R}(z) = \frac{1}{(2\pi)^{n/2} \det(R)^{1/2}} e^{-\frac{1}{2}(z-m)^T R^{-1}(z-m)^T R^{-1}(z-m)}$$

New parameterization by Information Geometry as Gibbs density:

$$p_{\overline{\xi}}(\xi) = \frac{1}{\int_{\Omega^*} e^{-\langle \xi, \beta \rangle} . d\xi} e^{-\langle \xi, \beta \rangle} \quad \text{with} \quad \xi = \begin{bmatrix} z \\ zz^T \end{bmatrix} \text{ and } \beta = \begin{bmatrix} -R^{-1}m \\ \frac{1}{2}R^{-1} \end{bmatrix} = \begin{bmatrix} a \\ H \end{bmatrix}$$

Duality bracket given by  $\langle \xi, \beta \rangle = a^T z + z^T H z = Tr \left[ z a^T + H^T z z^T \right]$ 

$$\log\left(\int_{S} e^{-\langle \xi, \beta \rangle} d\xi\right) = \log(Z) = n \log(2\pi) + \frac{1}{2} \log \det(R) + \frac{1}{2} m^T R^{-1} m$$

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### Information Geometry for Multivariate Gaussian Density

### Massieu characteristic function:

$$\psi_{\Omega}(\beta) = \int_{\Omega^*} e^{-\langle \xi, \beta \rangle} d\xi \text{ and } \beta = \begin{bmatrix} -R^{-1}m \\ \frac{1}{2}R^{-1} \end{bmatrix} = \begin{bmatrix} a \\ H \end{bmatrix}$$

$$\Phi(\beta) = -\log \psi_{\Omega}(\beta) = \frac{1}{2} \left[ -Tr \left[ H^{-1} a a^T \right] + \log \left[ (2)^n \det H \right] - n \log \left( 2\pi \right) \right]$$

$$\frac{\partial \Phi(\beta)}{\partial \beta} = \int_{\Omega^*} \xi \cdot p_{\overline{\xi}}(\xi) d\xi = \overline{\xi} = \begin{bmatrix} E[z] \\ E[zz^T] \end{bmatrix} = \begin{bmatrix} m \\ R + mm^T \end{bmatrix}$$

with  $\xi = \begin{bmatrix} z \\ T \\ Joint Structures and Common Four dations Zi Statistical Physics, Information Geometry and Inference for learning (SHGL 20) <math>R = E \begin{bmatrix} (z-m) (z-m)^T \end{bmatrix} = E \begin{bmatrix} zz^T \\ T \end{bmatrix} - mm^T$ 

## Information Geometry for Multivariate Gaussian Density

(Shannon) Entropy, Legendre transform of Massieu characteristic function

$$S\left(\overline{\xi}\right) = \left\langle\overline{\xi},\beta\right\rangle - \Phi\left(\beta\right) \quad with \quad \frac{\partial S\left(\overline{\xi}\right)}{\partial\overline{\xi}} = \beta = \begin{bmatrix}a\\H\end{bmatrix} = \begin{bmatrix}-R^{-1}m\\\frac{1}{2}R^{-1}\end{bmatrix}$$

$$S\left(\overline{\xi}\right) = -\int_{\Omega^*} \frac{e^{-\langle \xi, \beta \rangle}}{\int_{\Omega^*} e^{-\langle \xi, \beta \rangle} . d\xi} \log \frac{e^{-\langle \xi, \beta \rangle}}{\int_{\Omega^*} e^{-\langle \xi, \beta \rangle} . d\xi} . d\xi = -\int_{\Omega^*} p_{\overline{\xi}}(\xi) \log p_{\overline{\xi}}(\xi) . d\xi$$

(Shannon) Entropy with new parameterization:

$$S\left(\overline{\xi}\right) = \frac{1}{2}\left[\log(2)^n \det\left[H^{-1}\right] + n\log(2\pi \cdot e)\right] = \frac{1}{2}\left[\log\det\left[R\right] + n\log(2\pi \cdot e)\right]$$



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# THALES

## Souriau Algorithm for Exponential Map



## Souriau Book on « Calcul Linéaire » & Leverrier-Souriau Algorithm





$$P(\lambda) = \det(\lambda I - A) = \sum_{i=0}^{n} k_i \lambda^{n-1}$$

$$k_{0} = 1 \text{ et } B_{0} = I$$

$$\begin{cases}
A_{i} = B_{i-1}A , & k_{i} = -\frac{1}{i}tr(A_{i}), & i = 1,...,n-1 \\
B_{i} = A_{i} + k_{i}I & \text{ou } B_{i} = B_{i-1}A - \frac{1}{i}tr(B_{i-1}A)I \\
A_{n} = B_{n-1}A & \text{et } k_{n} = -\frac{1}{n}tr(A_{n})
\end{cases}$$

Souriau, J.-M..:Une méthode pour la décomposition spectrale et l'inversion des matrices. Comptes-Rendus hebdomadaires des séances de l'Académie des Sciences 227 (2), 1010– 1011, Gauthier-Villars, Paris (1948).



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## Souriau Algorithm for Characteristic Polynomial Computation

Souriau Algorithm (1948)

$$P(\lambda) = \det(\lambda I - A) = \sum_{i=0}^{n} k_i \lambda^{n-i} \qquad Q(\lambda) = adj(\lambda I - A) = \sum_{i=0}^{n-1} \lambda^{n-i-1} B_i$$

$$k_0 = 1 \quad \text{and} \quad B_0 = I$$

$$\begin{cases} A_i = B_{i-1}A \quad , \quad k_i = -\frac{1}{i} tr(A_i), \quad i = 1, ..., n-1 \\ B_i = A_i + k_i I \quad \text{or} \quad B_i = B_{i-1}A - \frac{1}{i} tr(B_{i-1}A)I \\ A_n = B_{n-1}A \quad \text{and} \quad k_n = -\frac{1}{n} tr(A_n) \end{cases}$$

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## Souriau Algorithm for Exponential Map Computation

### Souriau Extension Algorithm for Exponential Map

$$\begin{bmatrix} \lambda I - A \end{bmatrix}^{-1} = \frac{Q(\lambda)}{P(\lambda)} \Leftrightarrow \begin{bmatrix} \lambda I - A \end{bmatrix} Q(\lambda) = P(\lambda)I \qquad \begin{bmatrix} I \frac{d}{dt} - A \end{bmatrix} Q\left(\frac{d}{dt}\right) = P\left(\frac{d}{dt}\right)I \qquad P\left(\frac{d}{dt}\right)\gamma = 0$$

$$\begin{bmatrix} B_0 = I \text{ and } B_i = B_{i-1}A - \frac{tr(B_{i-1}A)}{i}I \\ k_0 = 1 \ , \ k_i = -\frac{tr(B_{i-1}A)}{i}i = 1, ..., n$$

$$\begin{bmatrix} \gamma \text{ integrated on } [0,h] \text{ such that} \\ k_0\gamma^{(n)} + k_1\gamma^{(n-1)} + ... + k_{n-1}\gamma^{(1)} + k_n\gamma = 0 \\ \text{with } \gamma(0) = ... = \gamma^{(n-2)} = 0 \text{ and } \gamma^{(n-1)}(0) = 1 \end{bmatrix}$$

$$\begin{bmatrix} 3 \text{ Computation of } \Phi(t) = e^{tA} = \sum_{i=0}^{n-1} \gamma^{(n-i-1)}(t)B_i \text{ on } [0,h] \\ 4 \text{ Extension of Computation on } [0, ph] \text{ by } \Phi(pt) = (\Phi(t))^p \end{bmatrix}$$

$$\texttt{THALES}$$

## Souriau algorithm to recover Lie Group Rodrigue's formula

$$SO(3) = \left\{ R / R^{-1} = R^{T} \right\} \qquad \omega_{x} = \begin{pmatrix} 0 & -\omega_{3} & \omega_{2} \\ \omega_{3} & 0 & -\omega_{1} \\ -\omega_{2} & \omega_{1} & 0 \end{pmatrix} = \omega_{1}L_{1} + \omega_{2}L_{2} + \omega_{3}L_{3} \in \mathfrak{so}(3), \\ \omega = (\omega_{1}, \omega_{2}, \omega_{3}) \in \mathbb{C}^{3} \\ \mathfrak{so}(3) = I \quad \mathrm{and} \quad k_{0} = I \\ B_{0} = I \quad \mathrm{and} \quad k_{0} = I \\ B_{1} = I.\omega_{x} - \frac{Tr(I.\omega_{x})}{1}I = \omega_{x} \quad \mathrm{and} \quad k_{1} = -\frac{Tr(I\omega_{x})}{1} = 0 \\ B_{2} = B_{1}.\omega_{x} - \frac{Tr(\omega_{x}.\omega_{x})}{2}I = \omega_{x}.\omega_{x} + \|\omega\|^{2}I \quad \mathrm{and} \quad k_{2} = -\frac{Tr(\omega_{x}.\omega_{x})}{2} = \|\omega\|^{2} \\ B_{2} = \omega_{x}.\omega_{x} + \|\omega\|^{2}I = \omega \otimes \omega^{T} \qquad k_{3} = 0 \\ \gamma^{(3)}(t) + \|\omega\|^{2}\gamma^{(1)}(t) = 0 \quad \mathrm{with} \quad \gamma^{(2)}(0) = I, \gamma^{(1)}(0) = 0, \gamma(0) = 0 \\ \gamma^{(1)}(t) = \frac{1}{\|\omega\|} \sin(\|\omega\|t) \quad \mathrm{and} \quad \gamma(t) = \frac{1}{\|\omega\|^{2}} \left(1 - \cos(\|\omega\|t)\right) \qquad \omega \otimes \omega^{T} = \omega_{x}.\omega_{x} + \|\omega\|^{T} I \\ \omega = I + \frac{1}{\|\omega\|} \sin(\|\omega\|t) = 0 \\ \omega \otimes \omega^{T} = U = U \\ \omega \otimes \omega^{T} \\ \omega \otimes \omega^{T} = U \\ \omega \otimes \omega^{T} = U \\ \omega \otimes \omega^{T} = U \\ \omega \otimes \omega^{T} \\ \omega \otimes \omega^{T} = U \\ \omega \otimes \omega^{T} \\ \omega \otimes \omega^{T} = U \\ \omega \otimes \omega^{T} \\ \omega \otimes \omega$$

## Reference

## Souriau Exponential Map Algorithm for Machine Learning on Matrix Lie Groups

## https://link.springer.com/chapter/10.1007%2F978-3-030-26980-7\_10

Der Link



International Conference on Geometric Science of Information

### Souriau Exponential Map Algorithm for Machine Learning on Matrix Lie Groups



Part of the Lecture Notes in Computer Science book series (LNCS, volume 11712)

THALES

# THALES

## Natural Exponential Families Invariant by a Group: Muriel Casalis & Gérard Letac



## Seminal work of Muriel Casalis (Institut Mathématique de Toulouse)

### Muriel Casalis PhD at Paul Sabatier Toulouse University supervised by Gérard Letac

### **Reference of Muriel Casalis**

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213 Distributions, Bull. Soc. math. France 39 (1), p. 129–144, 2011





## Travaux précurseurs de Muriel Casalis

International Justicial Review (1991), 56, 2, pp. 341–352. Printed in Group Britain © International Statistical Intiliate

#### Familles Exponentielles Naturelles sur $\mathbb{R}^d$ Invariantes par un Groupe

#### **Muriel Casalis**

Laboratoire de Statistique, Université Paul Sabatier, 118, route de Narbonne, 31062 Toulouse Cedex, France

#### Résumé

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La caractérisation des familles exponentielles naturelles de R<sup>el</sup> préservées par un groupe d'affinités donad est faite dans toon case celui d'un groupe compact, en particulier du groupe des rotationes, selui du groupe hypertolique et enfin celui d'un groupe quéclessations. La démarche adoptée consiste à traduire la propriété d'invariance de la famille par une propriété portant sur les mesures qui l'engeadrent pais à rechercher es demaires en conséquence.

#### **1** Introduction

Il est courant en statistique d'envisager un modèle  $(\Omega, sf, (P_x)_{x\in \Omega})$  tel qu'il existe un groupe G de permutations de  $\Omega$  préservant globalement la famille de probabilités  $F = \{P_n, 0 \in \Omega\}$ , c'est-àdire que pour tout (0, g) de  $\Theta \times G$ , l'image  $g(P_n)$  de  $P_n$  ar g est encore dans F (Barndorff-Nielsen parle alors de modèle de transformations). On pourra consulter Barndorff-Nielsen et al. (1982) et plus récemment le livre de Barndorff-Nielsen (1988).

Un exemple célèbre est celui des distributions de Fisher-Von-Mises pour lequel  $\Omega$  est la sphère unité de l'espace euclidien E.

 $P_{\theta}(dx) = L(\theta)^{-1} \exp{(\theta, x)} \sigma(dx),$ 

 $\sigma$  désignant la probabilité uniforme sur  $\Omega$  et  $L(\theta)$  le coefficient de normalisation, et pour lequel G est le groupe des rotations 0(E) de E.

Dans cet exemple,  $\{P_{\sigma}, \theta \in \Theta\}$  est une familie exponentielle naturelle au sens suivant. Soit E un espace vectoriel de dimension finie,  $E^{*}$  son dual et si  $(\theta, x)$  est dans  $E^{*} \times E$ ,  $(\theta, x)$  désigne le crochet de dualité; soit, de plus,  $\mu$  une mesure de Radon positive sur E; on note alors  $L_{\alpha}$  la transformée de Laplace de  $\mu$  définie par:

$$L_{\mu}: E^* \rightarrow [0, \infty]: \theta \rightarrow [\exp(\theta, x)\mu(dx);$$

 $D_{\mu}$  est l'ensemble { $\theta \in E^*$ ,  $L_{\mu}(\theta) \le \infty$ },  $\Theta(\mu)$  son intérieur et  $k_{\mu}$  la fonction définie sur  $\Theta(\mu)$  par:

 $k_{\mu}(\theta) = \text{Log } L_{\mu}(\theta).$ 

(1.1)

On désigne aussi par  $\mathcal{M}(E)$  l'ensemble des mesures de Radon  $\mu$  positives telles que:

(i) μ n'est pas concentrée sur un sous-espace affine strict de E;
 (ii) Θ(μ) est non vide.

N	° d'ordre 679			
		THESE		
		présentée à		
L'UNIVERSITE PAUL SABATIER DE TOULOUSE (SCIENCES)				
		pour obsenir		
	DOCTORAT	DE L'UNIVERSITE PAUL SABATIER		
	Spécialit	é : MATHEMATIQUES APPLIQUEES		
		per		
	Murie	el BONNEFOY - CASALIS		
FAMILLES EXPONENTIELLES NATURELLES				
	INVARIA	ANTES PAR UN GROUPE		
	Soutenue le 11 Jui	a 1990, devant la Commission d'Examen :		
MM.	H. CAUSSINUS	Professeur à l'Université Paul Sabatier		
	D. BAKRY	Professeur à l'Université Paul Sabatier		
	J. FARAUT	Professeur à l'Université PARIS VI		
	Y. GUIVARC'H	Professeur à l'Université PARIS VI		
	G. LETAC	Professour à l'Université Poul Schetter	1	
		rioresseur a romversite Paul Sababer		

Laboratoire de Statistique et Probabilités UNIVERSITE PAUL SABATIER

## THALES

## NEF (Natural Exponential Families): Letac & Casalis

Let E a vector space of finite size,  $E^{*}$  its dual.  $\langle heta, x 
angle$  Duality braket with  $(\theta, x) \in E^* \times E$ .  $\mu$  Positive Radon measure on E, Laplace transform is :  $L_{\mu}: E^* \to [0,\infty] \text{ with } \theta \mapsto L_{\mu}(\theta) = \int e^{\langle \theta, x \rangle} \mu(dx)$ Transformation  $k_u(\theta)$  defined on  $\Theta(u)$  interior of  $D_u = \{ \theta \in E^*, L_u < \infty \}$  $k_u(\theta) = \log L_u(\theta)$ Natural exponential families are given by:  $F(\mu) = \left\{ P(\theta, \mu)(dx) = e^{\langle \theta, x \rangle - k_{\mu}(\theta)} \mu(dx), \theta \in \Theta(\mu) \right\}$ Injective function (domian of means):  $k'_{\mu}(\theta) = \int x P(\theta, \mu) \mu(dx)$ And the inverse function:  $\psi_{\mu}: M_F \to \Theta(\mu) \text{ with } M_F = \text{Im}(k'_{\mu}(\Theta(\mu)))$  $V_{F}(m) = k_{\mu}^{"}(\psi_{\mu}(m)) = (\psi_{\mu}^{'}(m))^{-1}, \ m \in M_{F}$ Covariance operator: Information Geometry and Inference for Learning (SPIGL'20)

## NEF (Natural Exponential Families): Letac & Casalis

Measure generetad by a familly F :

$$F(\mu) = F(\mu') \Leftrightarrow \exists (a,b) \in E^* \times R, \text{ such that } \mu'(dx) = e^{\langle a,x \rangle + b} \mu(dx)$$

Let F an exponential family of E generated by  $\mu$  and  $\varphi: x \mapsto g_{\varphi}x + v_{\varphi}$ 

with 
$$g_{\varphi} \in GL(E)$$
 automorphisms of  $E$  and  $v_{\varphi} \in E$ , then the family  $\varphi(F) = \{\varphi(P(\theta, \mu)), \theta \in \Theta(\mu)\}$  is an exponential family of  $E$  generated by  $\varphi(\mu)$ 

Definition: An exponential family F is invariant by a group G (affine group of E), if  $\forall \varphi \in G, \varphi(F) = F \colon \forall \mu, F(\varphi(\mu)) = F(\mu)$ 

## (the contrary could be false)

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Joint Structures and Common Foundations of Statistical Physics, Information Geometry and Inference for Learning (SPIGL'20)
#### NEF (Natural Exponential Families): Letac & Casalis

#### Theorem (Casalis): Let $F = F(\mu)$ an exponential family of E and G

affine group of 
$$E$$
, then  $F$  is invariant by  $G$  if and only:  
 $\exists a: G \to E^*, \exists b: G \to R$ , such that:  
 $\forall (\varphi, \varphi') \in G^2, \begin{cases} a(\varphi \varphi') = {}^t g_{\varphi}^{-1} a(\varphi') + a(\varphi) \\ b(\varphi \varphi') = b(\varphi) + b(\varphi') - \langle a(\varphi'), g_{\varphi}^{-1} v_{\varphi} \rangle \\ b(\varphi \varphi) = b(\varphi) + b(\varphi') - \langle a(\varphi), g_{\varphi}^{-1} v_{\varphi} \rangle \end{cases}$ 
 $\forall \varphi \in G, \varphi(\mu)(dx) = e^{\langle a(\varphi), x \rangle + b(\varphi)} \mu(dx)$ 

When G is a linear subgroup, b is a character of G, a could be obtained by the help of Cohomology of Lie groups .

#### NEF (Natural Exponential Families): Letac & Casalis

If we define action of 
$$G$$
 on  $E^*$  by:  $g.x=^t g^{-1}x, g \in G, x \in E^*$   
we can verify that:  $a(g_1g_2) = g_1.a(g_2) + a(g_1)$ 

the action a is an inhomogeneous 1-cocycle:  $\forall n > 0$ , let the set of all functions from  $G^n$  to  $E^*$ ,  $\Im(G^n, E^*)$  called inhomogenesous n-cochains, then we can define the operators:  $d^n: \mathfrak{I}(G^n, E^*) \to \mathfrak{I}(G^{n+1}, E^*)$  $d^{n}F(g_{1},\dots,g_{n+1}) = g_{1}F(g_{2},\dots,g_{n+1}) + \sum_{i=1}^{n} (-1)^{i}F(g_{1},g_{2},\dots,g_{i}g_{i+1},\dots,g_{n})$  $+(-1)^{n+1}F(g_1,g_2,\cdots,g_n)$ 

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## **NEF (Natural Exponential Families): Letac & Casalis** Let $Z^n(G, E^*) = Ker(d^n), B(G, E^*) = Im(d^{n-1})$ , with $Z^n$ inhomogneous n-cocycles, the quotient $H^n(G, E^*) = Z^n(G, E^*) / B^n(G, E^*)$ is the Cohomology Group of G with value in $E^*$ . We have: $d^{0}: E^{*} \to \mathfrak{I}(G, E^{*}) \qquad Z^{0} = \{x \in E^{*}; g.x = x, \forall g \in G\}$ $x \mapsto (g \mapsto g.x - x)$ $d^1: \mathfrak{I}(G, E^*) \to \mathfrak{I}(G^2, E^*)$ $F \mapsto d^{1}F$ , $d^{1}F(g_{1}, g_{2}) = g_{1}F(g_{2}) - F(g_{1}g_{2}) + F(g_{1})$ $Z^{1} = \{F \in \mathfrak{I}(G, E^{*}), F(g_{1}g_{2}) = g_{1}.F(g_{2}) + F(g_{1}), \forall (g_{1}, g_{2}) \in G^{2}\}$ $B^{1} = \{F \in \mathfrak{I}(G, E^{*}); \exists x \in E^{*}, F(g) = g.x - x\}$

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#### NEF (Natural Exponential Families): Letac & Casalis

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- When the Cohomology Group  $H^1(G, E^*) = 0$  then  $Z^1(G, E^*) = B^1(G, E^*)$  $\Rightarrow \exists c \in E^*$ , such that  $\forall g \in G, a(g) = (I_d - g^{-1})c$
- Then if  $F = F(\mu)$  is an exponential familly invariant by G ,  $\mu$  verifies  $\forall g \in G, g(\mu)(dx) = e^{\langle c, x \rangle - \langle c, g^{-1}x \rangle + b(g)} \mu(dx)$  $\forall g \in G, g(e^{\langle c, x \rangle} \mu(dx)) = e^{b(g)} e^{\langle c, x \rangle} \mu(dx) \text{ with } \mu_0(dx) = e^{\langle c, x \rangle} \mu(dx)$ For all compact Group,  $H^1(G, E^*) = 0$  and we can express a $\forall (g,g') \in G^2, A_{oo'} = A_o A_{o'}$  $A: G \to GA(E)$  $g \mapsto A_g$ ,  $A_g(\theta) = g^{-1}\theta + a(g)$  A(G) compact sub - group of GA(E) $\exists \text{fixed point} \Rightarrow \forall g \in G, A_g(c) = {}^t g^{-1} c + a(g) = c \Rightarrow a(g) = (I_d - {}^t g^{-1})c$ Les Houches 27th-31st July 2020 OPEN

### THALES

# Bargmann Parameterization of SU(1,1) Lie Group



> SU(1,1) is isomorphic to SL(2,R) = Sp(2,R) through the complex unitary matrix W:

$$SL(2,R) = \left\{ g = \begin{pmatrix} a & b \\ c & d \end{pmatrix} / \det g = ad - bc = 1 \right\}$$
$$Sp(2,R) = \left\{ g = \begin{pmatrix} a & b \\ c & d \end{pmatrix} / gJg^{T} = J, J = \begin{pmatrix} 0 & +1 \\ -1 & 0 \end{pmatrix} \right\}$$
$$W = \frac{1}{\sqrt{2}} \begin{pmatrix} \omega^{-1} & \omega^{-1} \\ -\omega & \omega \end{pmatrix} = \left( W^{+} \right)^{-1} \text{ with } \omega = e^{i\pi/4} = \frac{1}{\sqrt{2}} \left( 1 + \frac{1}{\sqrt{2}} \right)$$

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> If we observe that  $W^{-1}JW = -iM$ , the isomorphism is given explicitly by:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} = g(u) = WuW^{-1} = \begin{pmatrix} \operatorname{Re}(\alpha + \beta) & -\operatorname{Im}(\alpha - \beta) \\ \operatorname{Im}(\alpha + \beta) & \operatorname{Re}(\alpha - \beta) \end{pmatrix}$$
$$\begin{pmatrix} \alpha & \beta \\ \beta^* & \alpha^* \end{pmatrix} = u(g) = W^{-1}gW = \frac{1}{2} \begin{pmatrix} (a+d) - i(b-c) & (a-d) + i(b+c) \\ (a-d) - i(b+c) & (a+d) + i(b-c) \end{pmatrix}$$

$$W = \frac{1}{\sqrt{2}} \begin{pmatrix} \omega^{-1} & \omega^{-1} \\ -\omega & \omega \end{pmatrix} = \left( W^{+} \right)^{-1} \quad \text{with} \quad \omega = e^{i\pi/4} = \frac{1}{\sqrt{2}} \left( 1 + i \right)$$

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> We can also make also a link with 
$$SO(2,1)$$
 of "1+2" pseudo-orthogonal matrices:  
 $SO(2,1) = \begin{cases} \Gamma \in GL(3,3) / \det(\Gamma) = 1, \Gamma K \Gamma^T = \Gamma, K = \begin{pmatrix} +1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \end{cases}$ 

$$\Gamma(g) = \begin{pmatrix} \frac{1}{2} (a^2 + b^2 + c^2 + d^2) & \frac{1}{2} (a^2 - b^2 + c^2 - d^2) & -cd - ab \\ \frac{1}{2} (a^2 + b^2 - c^2 - d^2) & \frac{1}{2} (a^2 - b^2 - c^2 + d^2) & -cd - ab \\ -bd - ac & bd - ac & ad + bc \end{cases}$$
with 276 dF  $Fa(g_{10}) \Gamma(g_{20}) = \Gamma(g_{1}g_{20}), \Gamma(I) = I, \Gamma(g^{-1}) = \Gamma(g)^{-1}$  HALES

> The matrix SO(2,1) corresponds to any SU(1,1) :

$$\Gamma(u) = \begin{pmatrix} |\alpha|^2 + |\beta|^2 & 2\operatorname{Re}\alpha\beta^* & 2\operatorname{Im}\alpha\beta^* \\ 2\operatorname{Re}\alpha\beta & \operatorname{Re}(\alpha^2 + \beta^2) & \operatorname{Im}(\alpha^2 - \beta^2) \\ -2\operatorname{Im}\alpha\beta & -\operatorname{Im}(\alpha^2 + \beta^2) & \operatorname{Re}(\alpha^2 - \beta^2) \end{pmatrix}$$

$$\alpha = \pm \sqrt{\frac{1}{2} (\Gamma_{11} + \Gamma_{12}) + i (\Gamma_{12} - \Gamma_{21})}, \quad \beta = \frac{1}{2\alpha} (\Gamma_{10} - i\Gamma_{20})$$

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- > The properties of connectivity of Sp(2,R) is described by its isomorphy with SU(1,1)
- > Using unimodular condition:

$$|\alpha|^2 - |\beta|^2 = 1 \Longrightarrow \alpha_R^2 + \alpha_I^2 - \beta_R^2 = 1 + \beta_I^2 \ge 1$$
  
with  $\alpha = \alpha_R + i\alpha_I$  and  $\beta = \beta_R + i\beta_I$ 

> If  $\beta_I$  is fixed,  $(\alpha_R, \alpha_I, \beta_R)$  are constrained to define a one-sheeted revolution hyperboloid, with its circular waist in the  $\alpha$  plane.

> To SU(1,1), we can associate the simply-connected universal covering group, using the maximal compact subgroup U(1) and corresponding to the Iwasawa decomposition (factorization of a noncompact semisimple group into its maximal compact subgroup times a solvable subgroup).

$$\begin{pmatrix} \alpha & \beta \\ \beta^* & \alpha^* \end{pmatrix} = \begin{pmatrix} e^{i\omega} & 0 \\ 0 & e^{i\omega} \end{pmatrix} \begin{pmatrix} \lambda & \mu \\ \mu^* & \lambda \end{pmatrix} \text{ with } \begin{cases} \omega = \arg \alpha = \frac{1}{2}i\ln(\alpha^*\alpha^{-1}) \\ \lambda = |\alpha| > 0 \\ \mu = e^{-i\omega}\beta = \sqrt{\frac{\alpha^*}{\alpha}\beta} \end{cases}$$

$$\beta = e^{i\omega} \mu, |\alpha|^2 - |\beta|^2 = \lambda^2 - |\mu|^2 = 1 \text{ so } |\mu| < \lambda$$

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> Bargmann has generalized this parameterization for Sp(2N, R), more convenient but difficult to generalize to N dimensions. For SU(1,1), Bargmann has used  $(\omega, \gamma)$ :  $\gamma = \frac{\mu}{\lambda} = \frac{\beta}{\alpha} (|\gamma| < 1), \ \lambda = \frac{1}{\sqrt{1 - |\gamma|^2}}, \ \mu = \frac{\gamma}{\sqrt{1 - |\gamma|^2}}$ For SL(2, R) = Sp(2, R), the Bargman, parameterization is given by this decomposition of a non-singular matrix into the product of an orthogonal and a singular matrix into the product of an orthogonal and a singular matrix. positive definite symmetric matrix:  $\begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} \cos \omega & -\sin \omega \\ \sin \omega & \cos \omega \end{pmatrix} \begin{pmatrix} \lambda + \operatorname{Re} \mu & \operatorname{Im} \mu \\ \operatorname{Im} \mu & \lambda - \operatorname{Re} \mu \end{pmatrix}$  $\omega = \arg\left[(a+d) - i(b-c)\right], \mu = e^{-i\omega}\left[(a-d) + i(b+c)\right]$ > SU(1,1) and SL(2,R) = Sp(2,R) are described when is counted modulo  $2\pi$ 228 Les Houches 27th-31<sup>#</sup> July 7020 Joint Ourmond of 21177al Physics, Information Geometry and Underscenot Learning (SriG) 20) OPEN