# Information Geometry - Hessian Geometry

Robert A. Wolak, JU joint work with Michel Nguiffo Boyom

Robert Wolak (JU)

Statistical models

2 Dual Connections

3 Hessian structures

Statistical models



3 Hessian structures

Statistical models





Statistical models

2 Dual Connections

3 Hessian structures

Robert Wolak (JU)

## General

Probability distribution on a set X is a non-negative real function function

 $p\colon X \to \mathbb{R}$ 

- 1) if X discrete and countable  $\sum_{x \in X} p(x) = 1$ ;
- 2) if  $X = \mathbb{R}^n \int_X p(x) dx = 1$ .

p is a probability density function.

Let  $\Lambda$  be a domain in  $\mathbb{R}^m$ . We consider families of probability distributions on a set  $\mathcal{X}$  parametrized by  $\lambda \in \Lambda$ .

 $\mathcal{P} = \{ p(x; \lambda) | \lambda \in \Lambda \}$ 

(1)  $\Lambda$  is a domain in  $\mathbb{R}^m$ ,

(2)  $p(x; \lambda)$  for a fixed x is a smooth function in  $\lambda$ ,

(3) the operation of integration with respect to x and differentiation with respect to  $\lambda$  are commutative.

A is called an *m*-dimensional statistical model (parametric model). Notation  $\Lambda = \{p(x; \lambda)\} = \{p_{\lambda}(x)\}, p(x; \lambda) = p_{\lambda}(x)$ 

# Examples

## Example (Normal distribution)

 $X = \mathbb{R}, m=2$ 

$$\Lambda = \{(\mu, \sigma) \colon -\infty < \mu < \infty, \mathbf{0} < \sigma < \infty\}$$

$$p(x; \lambda) = \frac{1}{\sqrt{2\pi\sigma}} exp\{-\frac{(x-\mu)^2}{2\sigma^2}\}$$

### Example (Multivariate normal distribution)

$$X = \mathbb{R}^k, m = k + \frac{k(k+1)}{2}, \lambda = (\mu, \Sigma)$$

 $\Lambda = \{(\mu, \Sigma) \colon \mu \in \mathbb{R}^k, \ \Sigma \in \mathbb{R}^{k^2} \colon \textit{positive definite}\}$ 

$$p(x;\lambda) = (2\pi)^{-k/2} (det\Sigma)^{-1/2} exp\{-\frac{1}{2}(x-\mu)^t \Sigma^{-1}(x-\mu)\}$$

# Examples cont.

### Example (Poisson distribution)

 $X = \mathbb{N}, m = 1, \Lambda = (0, \infty)$ 

$$p(x; \lambda) = e^{-\lambda} \frac{\lambda^x}{x!}$$

## Example (P(X) for finite X)

$$X = \{x_0, x_1, ..., x_n\}, \Lambda = \{(\lambda^1, ..., \lambda^n) \colon \lambda^i > 0, \ \Sigma_{i=1}^n \lambda^i < 1\}$$

$$p(x;\lambda) = \begin{cases} \lambda^{i} & 1 \leq i \leq n \\ 1 - \sum_{i=1}^{n} \lambda^{i} & i = 0 \end{cases}$$

# **Fisher metric**

#### Definition

Let  $\mathcal{P} = \{p(x; \lambda) | \lambda \in \Lambda\}$  be a family of probability distributions on a set  $\mathcal{X}$  parametrized by  $\lambda \in \Lambda$ .

We set  $l_{\lambda} = l(x; \lambda) = logp(x; \lambda)$  and denote by  $E_{\lambda}$  the expectation with respect to  $p_{\lambda}(x) = p(x; \lambda)$ .

Then the matrix  $g_F(\lambda) = [g_{ij}(\lambda)]$  defined by

$$g_{ij}(\lambda) = E_{\lambda} \left[\frac{\partial I_{\lambda}}{\partial \lambda^{i}} \frac{\partial I_{\lambda}}{\partial \lambda^{j}}\right] = \int_{\mathcal{X}} \frac{\partial I(x;\lambda)}{\partial \lambda^{i}} \frac{\partial I(x;\lambda)}{\partial \lambda^{j}} p(x;\lambda) dx$$

is called the Fisher information matrix tensor.

# Fisher metric cont.

Simple calculations show that

$$g_{ij}(\lambda) = -E_{\lambda}[rac{\partial^2 I_{\lambda}}{\partial \lambda^i \partial \lambda^j}].$$

The Fisher information matrix tensor  $g_F(\lambda) = [g_{ij}(\lambda)]$  is positive semi-definite on  $\Lambda$ :

$$\Sigma_{i,j}g_{ij}(\lambda)c^ic^j = \int_{\mathcal{X}} \{\Sigma_i c^i rac{\partial l(x;\lambda)}{\partial \lambda^i}\}^2 p(x;\lambda) dx \geq 0.$$

In information geometry the standard assumption has been:

(4) For a family of probability distributions  $\mathcal{P} = \{p(x; \lambda) | \lambda \in \Lambda\}$  the Fisher information matrix tensor  $g_F(\lambda) = [g_{ij}(\lambda)]$  is positive definite on  $\Lambda$ .

**Remark** The general case seems to be difficult to study if not hopeless. Therefore to develop a meaningful more general theory in our paper **[BW]** we assume that the **Fisher information matrix tensor is parallel with respect to some torsion-free connection** on  $\Lambda$ . The condition permits us to construct a foliation, and under some reasonable assumptions it has a transverse Hessian structure.

Robert Wolak (JU)

# Connections

$$(\Gamma_{ij,k}^{(\alpha)})_{\lambda} = E_{\lambda}[(\partial_i \partial_j l_{\lambda} + \frac{1-\alpha}{2} \partial_i l_{\lambda} \partial_j l_{\lambda})(\partial_k l_{\lambda})]$$

where  $\alpha$  is an arbitrary real number.

We define an affine connection  $\nabla^{(\alpha)}$  on  $\Lambda$  by

$$g(\nabla^{(\alpha)}_{\partial_i}\partial_j,\partial_k)=\Gamma^{(\alpha)}_{ij,k}$$

 $\nabla^{(\alpha)}$  is called the  $\alpha$ -connection.  $\nabla^{(\alpha)}$  is a symmetric connection.

$$\Gamma^{(\beta)}_{ij,k} = \Gamma^{(\alpha)}_{ij,k} + \frac{\alpha - \beta}{2} T_{ijk},$$

where  $T_{ijk}$  is a covariant symmetric tensor of degree 3 defined by

$$(T_{ijk})_{\lambda} = E_{\lambda}[\partial_i I_{\lambda} \partial_j I_{\lambda} \partial_k I_{\lambda}].$$

Moreover,

$$\nabla^{(\alpha)} = (1-\alpha)\nabla^{(0)} + \alpha\nabla^{(1)} = \frac{1+\alpha}{2}\nabla^{(1)} + \frac{1-\alpha}{2}\nabla^{(-1)}$$

Robert Wolak (JU)

Statistical models

## Connections cont.

#### Theorem

The 0-connection is the Riemannian connection with respect to the Fisher metric.

# Exponential family

If an *m*-dimensional model

$$S = \{p_{\theta} \colon \theta \in \Theta\}$$

can be expressed in terms of functions  $\{C, F_1, ..., F_m\}$  on X and a function  $\psi$  on  $\Theta$ :

$$p(x; \theta) = exp[C(x) + \sum_{i=1}^{n} \theta^{i} F_{i}(x) - \psi(\theta)],$$

then *S* is called an **exponential family** and  $\{\theta^i\}$  are called natural or canonical parameters.

From the normality condition

$$\psi(\theta) = \log \int \exp[C(x) + \sum_{i=1}^{n} \theta^{i} F_{i}(x)] dx.$$

The parametrization  $\theta \mapsto p_{\theta}$  is one-to-one iff the m+1 functions  $\{F_1, ..., F_m, 1\}$  are linearly independent. Always assumed!

Robert Wolak (JU)

# Examples

### Example (Normal distribution)

$$C(x) = 0, \quad F_1(x) = x, \quad F_2(x) = x^2, \quad \theta^1 = \frac{\mu}{\sigma^2} \quad \theta^2 = -\frac{1}{2\sigma^2}$$

$$\psi( heta)=rac{\mu^2}{2\sigma^2}+log(\sqrt{2\pi\sigma})=-rac{( heta^1)^2}{4 heta^2}+rac{1}{2}log(-rac{\pi}{ heta^2}).$$

#### Example

Consider an m-dimensional model  $S = \{p_{\theta}\}$  which can be expressed in terms of function  $\{C, F_1, ..., F_m\}$  on *X* as

$$p(x;\theta) = C(x) + \Sigma_i \theta^i F_i(x).$$

*S* forms an affine subspace of P(X). *S* is called a **mixture family** with **mixture parameters**  $\theta^i$ .

< < >> < <</>

# Examples

### Example (Normal distribution)

$$C(x) = 0, \quad F_1(x) = x, \quad F_2(x) = x^2, \quad \theta^1 = rac{\mu}{\sigma^2} \quad \theta^2 = -rac{1}{2\sigma^2}$$

$$\psi(\theta) = \frac{\mu^2}{2\sigma^2} + \log(\sqrt{2\pi\sigma}) = -\frac{(\theta^1)^2}{4\theta^2} + \frac{1}{2}\log(-\frac{\pi}{\theta^2}).$$

### Example

Consider an m-dimensional model  $S = \{p_{\theta}\}$  which can be expressed in terms of function  $\{C, F_1, ..., F_m\}$  on X as

$$p(x;\theta) = C(x) + \Sigma_i \theta^i F_i(x).$$

S forms an affine subspace of P(X).

S is called a mixture family with mixture parameters  $\theta^i$ .

### Example (Multivariate normal distribution)

$$C(x) = 0$$
,  $F_i(x) = x_i$ ,  $F_{ij}(x) = x_i x_j$   $(i \leq j)$ 

$$\theta^{i} = \Sigma_{j}(\Sigma^{-1})^{ij}\mu_{j}, \quad \theta^{ii} = (-1/2)(\Sigma^{-1})^{ii}, \quad \theta^{ij} = -(\Sigma^{-1})^{ij} \quad (i < j)$$

and

$$F_A(x) = x, \quad F_B(x) = xx^t, \quad \theta^A = \Sigma^{-1}\mu, \quad \theta^B = (-1/2)\Sigma^{-1},$$

We have

$$p(x;\theta) = exp[\Sigma_{1 \le i \le k}\theta^{i}F_{i}(x) + \Sigma_{1 \le i \le j \le k}\theta^{ij}F_{ij}(x) - \psi(\theta)]$$
$$= exp[(\theta^{A})^{t}F_{A}(x) + tr(\theta^{B}F_{B}(x)) - \psi(\theta)]$$

where  $\psi(\theta) = \dots$ 

#### Theorem

An exponential family (a mixture family, respectively) is  $\nabla^{(1)}$ -flat ( $\nabla^{(-1)}$ -flat, respectively) and its natural parameters (mixture parameters, respectively) form a  $\nabla^{(1)}$ -affine ( $\nabla^{(-1)}$ -affine, respectively) coordinate system.

#### Theorem

Let *S* be an exponential family (a mixture family, respectively) and *M* a submanifold of *S*. Them *M* is an exponential family (a mixture family, respectively) iff *M* is  $\nabla^{(1)}$ -autoparallel ( $\nabla^{(-1)}$ -autoparallel ) in *S*.

#### Theorem

An exponential family (a mixture family, respectively) is  $\nabla^{(1)}$ -flat ( $\nabla^{(-1)}$ -flat, respectively) and its natural parameters (mixture parameters, respectively) form a  $\nabla^{(1)}$ -affine ( $\nabla^{(-1)}$ -affine, respectively) coordinate system.

#### Theorem

Let *S* be an exponential family (a mixture family, respectively) and *M* a submanifold of *S*. Them *M* is an exponential family (a mixture family, respectively) iff *M* is  $\nabla^{(1)}$ -autoparallel ( $\nabla^{(-1)}$ -autoparallel ) in *S*.





3 Hessian structures

# Dual connections

When investigating the properties of the Fisher metric g and the  $\alpha$ -connection  $\nabla^{(\alpha)}$  it is important to consider them not individually, but rather as the triple  $(g, \nabla^{(\alpha)}, \nabla^{(-\alpha)})$ . The reason for this is that, through g, there exists a kind of duality between  $\nabla^{(\alpha)}$  and  $\nabla^{(-\alpha)}$  which is of fundamental significance. This notion of duality emerges not only when considering statistical models but also in many different problems related to information geometry.

(S,g) a Riemannian manifold,  $\nabla$  and  $\nabla^*$  two connections.

### Definition

If for any  $X, Y, Z \in \mathcal{X}(S)$ 

$$Zg(X, Y) = g(\nabla_Z X, Y) + g(X, \nabla_Z^* Y)$$

holds then the connections  $\nabla$  and  $\nabla^*$  are said to be dual (or conjugate). The triple  $(g, \nabla, \nabla^*)$  is called a dualistic structure on *S*. In local coordinates we have

$$\partial_k g_{ij} = \Gamma_{ki,j} + \Gamma^*_{kj,i}$$

For given g and  $\nabla$  there exists a unique dual connection  $\nabla^*$ 

Moreover,

- $(\nabla^*)^* = \nabla,$
- 2  $(\nabla + \nabla^*)/2$  is a metric connection,
- If a connection ∇' has the same torsion as  $\nabla^*$  and if  $(\nabla + \nabla')/2$  is metric, then  $\nabla' = \nabla^*$ .

#### Theorem

For any statistical model, the  $(\alpha)$ -connection and the  $(-\alpha)$ -connection are dual with respect to the Fisher metric.

#### Theorem

Let  $h_{\gamma} : T_p S \to T_q S$  (resp.  $h_{\gamma}^*$  be the parallel transport along curve  $\gamma$  from p to q with respect to  $\nabla$  (resp.  $\nabla^*$ ), then

$$g(h_{\gamma}(X),h_{\gamma}^{*}(Y))=g(X,Y)$$

for any vectors  $X, Y \in T_{p}S$ .

For any vector fields  $X, Y, Z, W \in \mathcal{X}(X)$ 

$$g(R(X, Y)Z, W) = -g(R^*(X, Y)W, Z)$$

thus

$$R=0$$
 iff  $R^*=0$ .

However, a similar property does not hold for the torsion tensors.

Let  $(g, \nabla, \nabla^*)$  be a dualistic structure on a manifold *S*. If the connections  $\nabla$  and and  $\nabla^*$  are both symmetric ( $T = T^* = 0$ ), then the  $\nabla$ -flatness and  $\nabla^*$ -flatness are equivalent.

Since the  $\alpha$ -connections are always symmetric, for any statistical model *S* and for any real number  $\alpha$  S is  $\alpha$ -flat iff S is  $(-\alpha)$ -flat.

We call  $(S, g, \nabla, \nabla^*)$  a dually flat space if both dual connections are flat.

#### Theorem

Let  $(S, g, \nabla, \nabla^*)$  be a dually flat space. If a submanifold M of S is autoparallel with respect to either  $\nabla$  or  $\nabla^*$ , then M is a dually flat space with respect to the dualistic structure  $(g_M, \nabla_M, \nabla_M^*)$  induced on M by  $(g, \nabla, \nabla^*)$ .

 $\hat{\xi} \colon X \to \mathbb{R}^m$  is called an **estimator**.

 $\hat{\xi}$  is called an unbiased estimator if  $E_{\xi}[\hat{\xi}(X)] = \xi$  for any  $\xi$ .

The mean squared error of an inbiased estimator  $\hat{\xi}$  may be expressed as the variance-covariance matrix  $V_{\xi}[\hat{\xi}] = [v_{\xi}^{ij}]$  where

$$v_{\xi}^{ij} = E_{\xi}[(\hat{\xi}^{i}(X) = \xi^{i})(\hat{\xi}^{j}(X) = \xi^{j})]$$

An unbiased estimator  $\hat{\xi}$  achieving the equality  $V_{\xi}[\hat{\xi}] = G(\xi)^{-1}$  for all  $\xi$  is called an **efficient estimator**.

#### Theorem

A necessary and sufficient condition for a coordinate system  $\xi$  of a model  $S = \{p_{\xi}\}$  to have an efficient estimator is that S is an exponential family and  $\xi$  is (-1)-affine.

Statistical models

2 Dual Connections

3 Hessian structures

# Hessian structures

### Definition

A Riemannian metric g on a flat manifold (M, D) is called a Hessian metric if for any point x of M there exists a local function  $\phi$  defined on an open nbhd of x such that

$$g = Dd\phi$$

If  $(x^1, ..., x^m)$  is an affine coordinate system for *D* then

$$g_{ij}=\frac{\partial^2\phi}{\partial x^i\partial x^j},$$

The pair (D, g) is called a **Hessian structure** on *M*; *M* is called a **Hessian manifold** - notation (M, D, g). A function  $\phi$  is called a (local) potential of (D, g).

### Definition

A Hessian structure (D,g) is said to be of Koszul type if there exists a closed 1-form  $\omega$  such that  $g = D\omega$ .

Robert Wolak (JU)

## Hessian structures cont.

Let  $\nabla$  be the Levi-Civita connection of the Riemannian metric g. Let  $\gamma$  be the difference tensor

$$\gamma_X Y = \nabla_X Y - D_X Y$$

As  $\nabla$  and *D* are torsion-free  $\gamma_X Y = \gamma_Y X$ .

### Proposition

Let (M,D) be a flat manifold and g a Riemannian metric on M. Then the following conditions are equivalent:

$$(D_Xg)(Y,Z) = (D_Yg)(X,Z),$$

$$\mathbf{b} \ \frac{\partial g_{ij}}{\partial x^k} = \frac{\partial g_{kj}}{\partial x^i},$$

 $) \ \gamma_{ijk} = \gamma_{jik}.$ 

## cont.

Let (M, D) be a flat manifold and  $\pi: TM \to M$  its tangent bundle. To an affine chart  $(x^1, ..., x^m)$  we associate a complex chart on TM

$$z^j = \xi^j + i\xi^{m+j}$$

where  $\xi^i = x^i \pi$  and  $\xi^{m+i} = dx^i$  for i = 1, ..., m.  $J_D$  the associated complex structure on *TM*.

On *TM* we define the following Riemannian metric  $g^{T}$ 

$$g^{ au} = \Sigma g_{ij} \pi dz^i dar{z}^j$$

#### Proposition

Let (M, D) be a flat manifold and g a Riemannian metric on M. Then the following conditions are equivalent: (1) g is a Hessian metric on (M, D), (2)  $g^{T}$  is a Kählerian metric on  $(TM, J_{D})$ .

## cont.

#### Theorem

Let (M,D,g) be a Hessian manifold and let  $\nabla$  be the Levi-Civita connection of g. Define a connection D' by

$$D'=2\nabla-D.$$

Then (1) D' is a flat connection, (2)  $Xg(Y,Z) = g(D_XY,Z) + g(Y,D'_XZ)$ , (3) (D',g) is a Hessian structure.

# Codazzi structures

### Proposition

Let *D* be a torsion-free connection and let *g* be a Riemannian metric. Let D' be a new connection defined by

$$Xg(Y,Z) = g(D_XY,Z) + g(Y,D'_XZ)$$

Then the following conditions are equivalent: (1) the connection D' is torsion-free, (2) The pair (D, g) satisfies the Codazzi equation

$$(D_Xg)(Y,Z)=(D_Yg)(X,Z),$$

(3) Let  $\nabla$  be the Levi-Civita connection for g, let  $\gamma_X Y = \nabla_X Y - D_X Y$ . Then

$$g\gamma_X Y, Z) = g(Y, \gamma_X Z).$$

If the pair (D, g) satisfies the Codazzi equation, so does the pair (D', g) and

$$D' = 2\nabla - D$$
 and  $(D_X g)(Y, Z) = 2g(\gamma_X Y, Z).$ 

# Codazzi structures cont.

### Definition

A pair (D, g) where D is a torsion-free connection and g a Riemannian metric on a manifold M is called a **Codazzi structure** if it satisfies the Codazzi

 $(D_Xg)(Y,Z)=(D_Yg)(X,Z),$ 

For a Codazzi structure (D, g) the connection D' defined by

$$Xg(Y,Z) = g(D_XY,Z) + g(Y,D'_XZ)$$

is called the **dual connection** of *D* with respect to *g*, and the pair (D', g) the **dual Codazzi structure** of (D, g).

#### Definition

A Codazzi structure (D, g) is of constant curvature c if the curvature tensor  $R_D$  of D satisfies

$$R_D(X, Y)Z = c\{g(Y, Z)X - g(X, Z)Y\}.$$

Robert Wolak (JU)

Information Geometry

# Codazzi structures cont.

### Definition

A pair (D, g) where D is a torsion-free connection and g a Riemannian metric on a manifold M is called a **Codazzi structure** if it satisfies the Codazzi

 $(D_Xg)(Y,Z)=(D_Yg)(X,Z),$ 

For a Codazzi structure (D, g) the connection D' defined by

$$Xg(Y,Z) = g(D_XY,Z) + g(Y,D'_XZ)$$

is called the **dual connection** of *D* with respect to *g*, and the pair (D', g) the **dual Codazzi structure** of (D, g).

#### Definition

A Codazzi structure (D, g) is of constant curvature *c* if the curvature tensor  $R_D$  of *D* satisfies

$$R_D(X,Y)Z = c\{g(Y,Z)X - g(X,Z)Y\}.$$

Robert Wolak (JU)

#### Proposition

Let (D, g) be a Codazzi structure nad (D', g) its dual Codazzi structure. Then (1)

$$g(R_D(X,Y)Z,W)+g(Z,R_{D'}(X,Y)W)=0;$$

(2) if (D, g) is a Codazzi struccture of constant curvature c, then (D', g) is also of constant curvature c.

#### Proposition

A Codazzi structure (D, g) is of constant curvature 0 iff (D, g) is a Hessian structure.

#### Proposition

Let (D, g) be a Codazzi structure of constant curvature. Then locally g is of the form

$$Dd\phi + \frac{\phi}{m-1}Ric_D$$

where  $Ric_D$  is the Ricci tensor of D and  $\phi$  is a local function.

#### Proposition

Let (D, g) be a Codazzi structure nad (D', g) its dual Codazzi structure. Then (1)

$$g(R_D(X,Y)Z,W)+g(Z,R_{D'}(X,Y)W)=0;$$

(2) if (D, g) is a Codazzi struccture of constant curvature c, then (D', g) is also of constant curvature c.

#### Proposition

A Codazzi structure (D, g) is of constant curvature 0 iff (D, g) is a Hessian structure.

#### Proposition

Let (D, g) be a Codazzi structure of constant curvature. Then locally g is of the form

$$Dd\phi + \frac{\phi}{m-1}Ric_D$$

where  $Ric_D$  is the Ricci tensor of D and  $\phi$  is a local function.

#### Proposition

Let (D, g) be a Codazzi structure nad (D', g) its dual Codazzi structure. Then (1)

$$g(R_D(X,Y)Z,W)+g(Z,R_{D'}(X,Y)W)=0;$$

(2) if (D, g) is a Codazzi struccture of constant curvature c, then (D', g) is also of constant curvature c.

#### Proposition

A Codazzi structure (D, g) is of constant curvature 0 iff (D, g) is a Hessian structure.

#### Proposition

Let (D, g) be a Codazzi structure of constant curvature. Then locally g is of the form

$$Dd\phi + rac{\phi}{m-1}Ric_D$$

where  $Ric_D$  is the Ricci tensor of D and  $\phi$  is a local function.

# Example

Let S(m) be the set of real symmetric matrices of degree *m*, and let  $S(m)^+$  be the subset of S(m) consisting of of positive-definite symmetric matrices. Put

$$p(x; \mu, \sigma) = (2\pi)^{-m/2} (det\sigma)^{-1/2} exp(\{-\frac{t(x-\mu)\sigma^{-1}(x-\mu)}{2}\})$$

where  $\mu \in \mathbb{R}^m$  and  $\sigma \in S(m)^+$ . Then  $\{p(x; \mu, \sigma) : (\mu, \sigma) \in \mathbb{R}^m \times S(m)^+\}$  is a family of probability distributions on  $\mathbb{R}^m$  parametrized by  $(\mu, \sigma)$  and called a family of **m-dimensional normal distributions**.

Let  $\Omega$  be a domain in a finite dimensional real vector space *V*, and let  $\rho$  be an injective linear mapping from  $\Omega$  into  $S(m)^+$ .

#### Proposition

Let  $\{p(x; \mu, \omega) \colon (\mu, \omega) \in \mathbb{R}^m \times \Omega\}$  be a family of probability distributions induced by  $\rho$ . Then the family is an exponential family parametrized by  $\theta = \rho(\omega)\mu \in \mathbb{R}^m$  and  $\omega \in \Omega$ . The Fisher information metric is a Hessian metric on  $\mathbb{R}^m \times \Omega$  with potential function

$$\phi(\theta,\omega) = (1/2)\{{}^t\theta\rho(\omega)^{-1}\theta - \log \det\rho(\omega)\}.$$

 Amari S-I and Nagaoka H.: Methods of Information Geometry, Translations of Mathematical monographs, AMS-OXFORD, vol 191
Shima, H.: The Geometry of Hessian Structures. World Scientific (2007)
**[BW]** Nguiffo Boyom M and W R.: Transversely Hessian foliations and information geometry, Int. J. Math. 27,11 (2016)

#### Conferences

1) Geometric Science of Information - GSI (Mines ParisTech, Paris 2014 and 2017)

2) Topological and Geometrical Structure of Information (CIRM 2017)

# Foliations

Let  $\mathcal{F}$  be a foliation on an *m*-manifold *M*. Then  $\mathcal{F}$  is defined by a cocycle  $\mathcal{U} = \{U_i, f_i, k_{ij}\}_{i \in I}$  modeled on a *q*-manifold  $N_0$  (0 < q < m) such that (1)  $\{U_i\}_{i \in I}$  is an open covering of *M*, (2)  $f_i : U_i \to N_0$  are submersions with connected fibres, (3)  $k_{ij} : N_0 \to N_0$  are local diffeomorphisms of  $N_0$  with  $f_i = k_{ij}f_j$  on  $U_i \cap U_j$ .

The connected components of the trace of any leaf of  $\mathcal{F}$  on  $U_i$  are fibres of  $f_i$ , and the trace itself consists of at most a denombrable number of these fibres.

The open subsets  $N_i = f_i(U_i) \subset N_0$  form a *q*-dimensional manifold  $N_{\mathcal{U}} = \bigsqcup N_i$ , which can be considered to be a complete transverse manifold of the foliation  $\mathcal{F}$ . The pseudogroup  $\mathcal{H}_{\mathcal{U}}$  of local diffeomorphisms of N generated by  $k_{ij}$  is called the holonomy pseudogroup of the foliated manifold  $(M, \mathcal{F})$  defined by the cocycle  $\mathcal{U}$ .

A foliation on a smooth manifold M understood as an involutive subbundle of TM, or equivalently, according to the Frobenius theorem as a partition of the manifold by submanifolds of the same dimension with some regularity condition, can be defined by many different cocycles.

There is a notion of equivalent cocycles, similar to the notion of equivalent atlases of a smooth manifold, and a foliation can be understood as an equivalence class of such cocycles. The equivalence class  $\mathcal{H}$  of  $\mathcal{H}_{\mathcal{U}}$ , is called the **holonomy group** of  $\mathcal{F}$ , or of the foliated manifold  $(M, \mathcal{F})$ .

The vector bundle  $N(M, \mathcal{F}) = TM/T\mathcal{F}$  is called the **normal bundle of the foliation**  $\mathcal{F}$ . Then the tangent bundle TM is isomorphic to the direct sum  $T\mathcal{F} \oplus N(M, \mathcal{F})$ . These isomorphisms are determined by the choice of a supplementary subbundle Q in TM to the tangent bundle to the foliation  $T\mathcal{F}$ .

The cocycle  $\mathcal{U} = \{U_i, f_i, k_{ij}\}_{i \in I}$  modeled on a *q*-manifold  $N_0$  induces on the normal bundle a cocycle  $\mathcal{V} = \{V_i, \overline{f}_i, \overline{k}_{ij}\}_{i \in I}$  modeled on the 2*q*-manifold  $TN_0$ , where  $V_i = TU_i$ ,  $\overline{f}_i$  is the mapping induced by  $df_i$ , and  $\overline{k}_{ij} = dk_{ij}$ .

The foliation  $\mathcal{F}_N$  of the normal bundle is of codimension 2q, its leaves project on leaves of  $\mathcal{F}$ . They are, in fact, coverings of these leaves.

In a similar way one can foliate any bundle obtained via a point-wise process from the normal bundle, e.g., the frame bundle of the normal bundle, the dual normal bundle, any tensor product of these bundles.

イロト イヨト イヨト イヨト

## Geometric structures on foliated manifolds

In the case of a foliated manifold we can consider three types of geometrical structures related to the foliation:

*transverse* - defined on the transverse manifold, the associated holonomy pseudogroup consists of automorphisms of this geometrical structure;

*foliated* - only defined on the normal bundle, and when expressed in a local adapted chart, depending only on the transverse coordinates; a foliated structure projects to a transverse structure along submersions of the cocycle defining the foliation.;

*associated* - defined globally, on the tangent bundle but adapted to the spliting, and defining a foliated structure on the normal bundle.

Foliated and transverse structures are in one-to-one correspondence, an associated structure defines a foliated structure, but different associated structures can define the same foliated structure.