

# Wasserstein Riemannian geometry of Gaussian densities

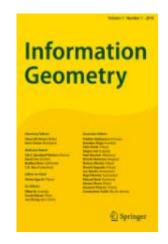
Luigi Malagò<sup>1</sup> · Luigi Montrucchio<sup>2</sup> · Giovanni Pistone<sup>3</sup>

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### **Abstract**

The Wasserstein distance on multivariate non-degenerate Gaussian densities is a Riemannian distance. After reviewing the properties of the distance and the metric geodesic, we present an explicit form of the Riemannian metrics on positive-definite matrices and compute its tensor form with respect to the trace inner product. The tensor is a matrix which is the solution to a Lyapunov equation. We compute the explicit formula for the Riemannian exponential, the normal coordinates charts and the Riemannian gradient. Finally, the Levi-Civita covariant derivative is computed in matrix form together with the differential equation for the parallel transport. While all computations are given in matrix form, nonetheless we discuss also the use of a special moving frame.

**Keywords** Information geometry  $\cdot$  Gaussian distribution  $\cdot$  Wasserstein distance  $\cdot$  Riemannian metrics  $\cdot$  Natural gradient  $\cdot$  Riemannian exponential  $\cdot$  Normal coordinates  $\cdot$  Levi-Civita covariant derivative  $\cdot$  Optimization on positive-definite symmetric matrices





# Natural gradient via optimal transport

Wuchen Li¹ (□) • Guido Montúfar<sup>2,3</sup>

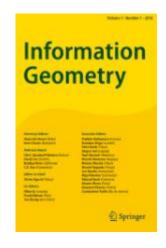
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### Abstract

We study a natural Wasserstein gradient flow on manifolds of probability distributions with discrete sample spaces. We derive the Riemannian structure for the probability simplex from the dynamical formulation of the Wasserstein distance on a weighted graph. We pull back the geometric structure to the parameter space of any given probability model, which allows us to define a natural gradient flow there. In contrast to the natural Fisher–Rao gradient, the natural Wasserstein gradient incorporates a ground metric on sample space. We illustrate the analysis of elementary exponential family examples and demonstrate an application of the Wasserstein natural gradient to maximum likelihood estimation.

**Keywords** Optimal transport  $\cdot$  Information geometry  $\cdot$  Wasserstein statistical manifold  $\cdot$  Displacement convexity  $\cdot$  Machine learning



#### RESEARCH PAPER

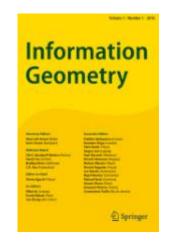
## Superharmonic priors for autoregressive models

Fuyuhiko Tanaka<sup>1</sup>

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**Abstract** Tanaka and Komaki (Sankhya Ser A Indian Stat Inst 73-A:162–184, 2011) proposed superharmonic priors in Bayesian time series analysis as alternative to the famous Jeffreys prior. By definition the existence of superharmonic priors on a specific time series model with finite-dimensional parameter is equivalent to that of positive nonconstant superharmonic functions on the corresponding Riemannian manifold endowed with the Fisher metric. In the autoregressive models, whose Fisher metric and its inverse have quite messy forms, we obtain superharmonic priors in an explicit manner. To derive this result, we developed a systematic way of dealing with symmetric polynomials, which are related to Schur functions.

**Keywords** Jeffreys prior · Superharmonic priors · Autoregressive models · Noninformative priors · Kullback–Leibler divergence · Fisher metric





# Asymptotic dependency structure of multiple signals

Asymptotic equipartition property for diagrams of probability spaces

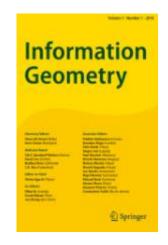
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### Abstract

We formalize the notion of the dependency structure of a collection of *multiple* signals, relevant from the perspective of information theory, artificial intelligence, neuroscience, complex systems and other related fields. We model multiple signals by commutative diagrams of probability spaces with measure-preserving maps between some of them. We introduce the asymptotic entropy (pseudo-)distance between diagrams, expressing how much two diagrams differ from an information-processing perspective. If the distance vanishes, we say that two diagrams are asymptotically equivalent. In this context, we prove an asymptotic equipartition property: any sequence of tensor powers of a diagram is asymptotically equivalent to a sequence of homogeneous diagrams. This sequence of homogeneous diagrams expresses the relevant dependency structure.





## Asymptotic dependency structure of multiple signals

Asymptotic equipartition property for diagrams of probability spaces

Rostislav Matveev<sup>1</sup> • Jacobus W. Portegies<sup>2</sup>

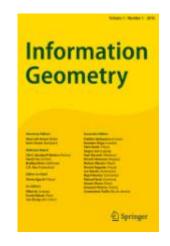
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**Keywords** Asymptotic equipartition property  $\cdot$  Entropy distance  $\cdot$  Diagrams of probability spaces  $\cdot$  Multiple signals



#### RESEARCH PAPER

## Ordering positive definite matrices

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**Abstract** We introduce new partial orders on the set  $S_n^+$  of positive definite matrices of dimension n derived from the affine-invariant geometry of  $S_n^+$ . The orders are induced by affine-invariant cone fields, which arise naturally from a local analysis of the orders that are compatible with the homogeneous geometry of  $S_n^+$  defined by the natural transitive action of the general linear group GL(n). We then take a geometric approach to the study of monotone functions on  $S_n^+$  and establish a number of relevant results, including an extension of the well-known Löwner-Heinz theorem derived using differential positivity with respect to affine-invariant cone fields.

 $\textbf{Keywords} \ \ Positive \ definite \ matrices \cdot Partial \ orders \cdot Monotone \ functions \cdot Monotone \ flows \cdot Differential \ positivity \cdot Matrix \ means$ 

