Quantum Geometry insights in deep learning <u>Noemie Combe</u>
Boltzmann machine

A connection between optimal transport theory and deep learning $\,^\circ$ How [Monge-Ampere equation] governs probability transformations in generative models $\,^\circ$

Information geometry :

A statistical manifold is a space where each point represents a probability distribution. For example, the set of all Gaussian distributions forms a statistical manifold parameterized by the mean and variance.

The parameters of the distributions serve as coordinates on this manifold.

2. Riemannian Metric:

- The statistical manifold is equipped with a Riemannian metric, which provides a way to measure distances and angles between probability distributions.
- The most commonly used metric in information geometry is the **Fisher information metric**, which measures the amount of information that a random variable carries about a parameter.

Frobenius manifold :

Frobenius manifolds were introduced by Boris Dubrov in in the early 1990s as a framework to unify various phenomena in mathematical physics and algebraic geometry • It provides a unifying framework for studying integrable systems • topological field theories • and mirror symmetry • making it a central concept in modern mathematical physics and algebraic geometry •

Quantum geometry :

Deep learning theory :