

# Statistical Inference and the Parallel Transport of Probability

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Based on 2203.08119 (+ 2204.05084, 2205.11543)

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## Maximum entropy

Consider a probability density  $p$  over some space  $X$  satisfying the diffusion process

$$\frac{\partial}{\partial t} p = -\frac{\partial}{\partial x} \left( \frac{\partial}{\partial x} Jp \right) + D \frac{\partial^2}{\partial x^2} p$$

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Discussed by Jordan, Kinderlehrer, and Otto (1998); Markowich and Villani (2000)

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Stationary solution is the desired Gibbs measure,  $\exp\{-\lambda J\}$ .

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and therefore as a section  $s$  of a line bundle  $E \xrightarrow{\pi} X$  with typical fibre  $\mathbb{R}_{>0}$ . Hence  $J$  is also a constraint on the shape of the graph of  $s$ .

## A quick word on parallel transport

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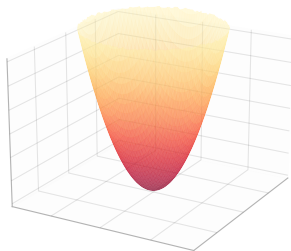
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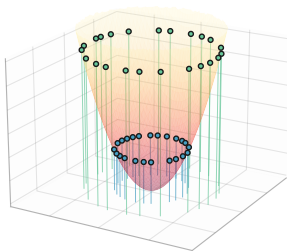
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So  $s$  should consist of parallel transport lines

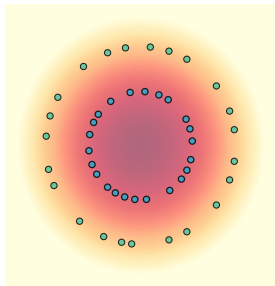
(a)  $J(x, y)$



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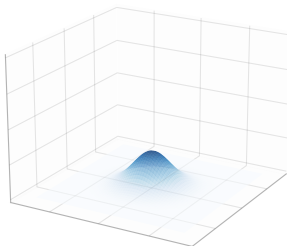


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(d)

$p(x, y)$



Adapted from  
arXiv:2205.11543.  
Credit to  
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## The stationary solution to CME is parallel transport

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$\therefore$  maximising entropy yields the solution to parallel transport with connection valued in  $\mathbb{R}$ .

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Stationary solution to constrained maximum entropy:

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Solution to covariant differentiation of a section  $s$  in connection

$$dJ(x): s(x) = \exp\left\{-\lambda \int_k^x dJ(\tilde{x})\right\} = \exp\{-\lambda J(x)\}$$



## A first idea of gauge theory

Suppose  $S$  is a functional on the space of  $p$ 's,  $\Gamma(E)$  and let  $G$  be a set. A *gauge* is a quantity  $g$  in some Lie group  $G$  such that

$$S(\rho(g)\Gamma) = 0$$

for all  $g \in G$ . A choice of gauge is a choice of one such  $g$ . Our  $\rho(g)$  is a  $G$ -valued transition function  $t$ .

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Motivation: changes of frame on  $TX$ , where

$$V = V^i \frac{\partial}{\partial x^i} \text{ or } \tilde{V}^i \frac{\partial}{\partial y^i}$$

related by the Jacobian matrix

$$\tilde{V}^j = \frac{\partial y^j}{\partial x^i} V^i$$

## A gauge-theoretic interpretation

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Under  $p \mapsto e^{-J'} p$ ,  $J \mapsto J' + J$ ,

$$\begin{aligned}\mathcal{L} &= -e^{-J} p \ln e^{-J} p - (J' + J) e^{-J} p \\ &= e^{-J} p (-\ln p - J).\end{aligned}$$

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What is the variation? Root of

$$e^{-J} (-\ln p - J)$$

is equal to the root of

$$-\ln p - J.$$

## Gibbs frames

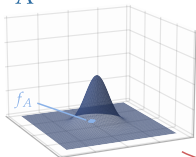
### Theorem

*Let  $P$  be the principal bundle associated to  $E$ . Since level sets of  $p$  are constant with respect to  $J$ , there exists some group element  $\exp\{J\}$  for  $p$  whose logarithmic derivative is  $dJ$ ; moreover, there exists a principal bundle of such Gibbs frames, which is  $P$ .*



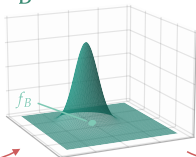
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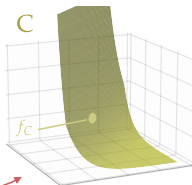
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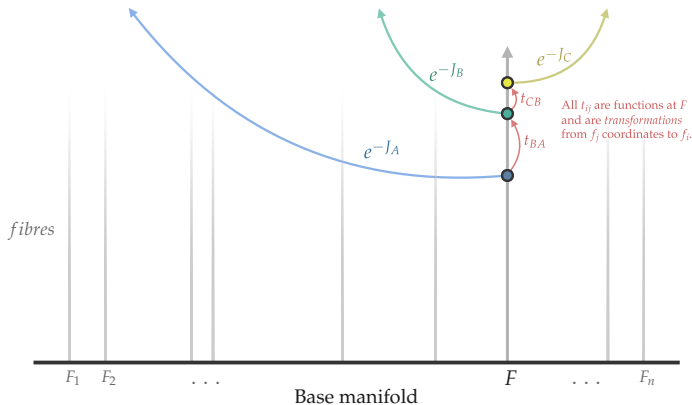


$$t_{BA}$$

$$f_B = t_{BA}f_A$$

$$t_{CB}$$

$$f_C = t_{CB}f_B$$



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- ▶ Vertical  $\oplus$  horizontal flows