# Notes on Fisher distance in statistics, thermodynamics, quantum informatics 

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(1) Ronald Fisher

- Fisher-information 1925
- I never understood, but a week ago
- Fisher's statistical distance
- I never understood
(2) From Fisher to Riemannian metric in thermodynamics
- I noticed a Feldmann-Levine-Salamon paper 1986
- and wrote "him" a letter
- Encounter 8yy later, "he" is she 1994
- What (the hell) is the thermodynamic length $\ell_{12}$ ?
(3) From Fisher to quantum informatics
- Statistical distinguishability distance of quantum states
- A best(?) answer to what (...) the statistical distance is

Sir Ronald Aylmer Fisher FRS ${ }^{[3]}$ (17 February 1890 - 29 July 1962) was a British statistician and geneticist. For his work in statistics, he has been described as "a genius who almost single-handedly created the foundations for modern statistical science" ${ }^{[4]}$ and "the single most important figure in 20th century statistics". ${ }^{[5]}$ In genetics, his work used mathematics to combine Mendelian genetics and natural selection; this contributed to the revival of Darwinism in the early 20th-century revision of the theory of evolution known as the modern synthesis. For his contributions to biology, Fisher has been called "the greatest of Darwin's successors". ${ }^{[6]}$

From 1919 onward, he worked at the Rothamsted Experimental Station for 14 years; ${ }^{[7]}$ there, he analysed its immense data from crop experiments since the 1840s, and developed the analysis of variance (ANOVA). He established his reputation there in the following years as a biostatistician.

He is known as one of the three principal founders of population genetics. He outlined Fisher's principle, the Fisherian runaway and sexy son hypothesis theories of sexual selection. His contributions to statistics include the maximum likelihood, fiducial inference, the derivation of various sampling distributions, founding principles of the design of experiments, and much more.

Fisher held strong views on race. Throughout his life, he was a prominent supporter of eugenics, an interest which led to his work on statistics and genetics. ${ }^{[8]}$ Notably, he was a dissenting voice in the 1950 UNESCO statement The Race Question, insisting on racial differences. ${ }^{[9]}$


## Fisher-information 1925

Fisher: Maximum Likelihood Method of parameter estimation Probability $p_{k}(\theta)$, sample $\bar{k}$, unknown continuous parameter $\theta$ MaxLik estimate $\theta^{\star}$ : maximize $p_{\bar{k}}(\theta) \Leftrightarrow$ maximize $L_{k}(\theta)=\ln p_{\bar{k}}(\theta)$ :

$$
\left.\frac{\partial L_{\bar{k}}(\theta)}{\partial \theta}\right|_{\theta=\theta^{\star}}=0
$$

Confidence interval $\Delta \theta$ of the estimate satisfies

$$
-\left.\frac{\partial^{2} L_{\bar{k}}(\theta)}{\partial \theta^{2}}\right|_{\theta=\theta^{\star}}(\Delta \theta)^{2}=1
$$

Expectation value

$$
-\sum_{k} p_{k}(\theta) \frac{\partial^{2} L_{k}(\theta)}{\partial \theta^{2}}
$$

is called Fisher-information.

## I never understood, but a week ago

Fisher-information 1925

$$
F(\theta)=-\sum_{k} p_{k}(\theta) \frac{\partial^{2} L_{k}(\theta)}{\partial \theta^{2}} \equiv \sum_{k} \frac{1}{p_{k}(\theta)}\left(\frac{\partial p_{k}(\theta)}{\partial \theta}\right)^{2}
$$

Shannon-information 1948:

$$
S(\theta)=-\sum_{k} p_{k}(\theta) L_{k}(\theta)=-\sum_{k} p_{k}(\theta) \ln p_{k}(\theta)
$$

I never understood why we keep calling $F(\theta)$ information?
Finally (a week ago) I looked into Fisher 1925:
He called it information, we follow him, but we never tell our students that his information has little to do with Shannon's later (and standard) concept of information.
the amount of information provided by a combination of two or more independent observations is thus merely the sum of the amounts of information in each piece separately.

It is a common case for a sample of $n$ observations to be distributed into a finite number of classes, the numbers "expected" in each class being functions of one or more unknown parameters, if $p$ is the probability of an observation falling into any one class, the amount of information in the sample is

$$
S\left\{\frac{1}{m}\left(\frac{\partial m}{\partial \theta}\right)^{2}\right\}
$$

where $m=n p$, is the expectation in any one class. The variance of an efficient statistic derived from a large sample may, of course, be calculated from this expression.

## 7. Efficiency of the maximum likelihood solution.

We shall now prove that when an efficient statistic as defined above exists one may be found by the method of maximum likelihood.

If $f$ stand for the probability that any particular type of observation should occur, and $\phi$ for the probability that any particular type of sample should occur, then

$$
\log \phi=C+S(\log f)
$$

when $C$ is a constant which does not involve the parameters, the summation extending over all observations.

## Fisher's statistical distance

Fisher-information is a metric of distiguishability rather than information:

$$
(\Delta \ell)^{2}=F(\theta)(\Delta \theta)^{2}=\sum_{k} \frac{1}{p_{k}(\theta)}\left(\frac{\partial p_{k}(\theta)}{\partial \theta}\right)^{2}(\Delta \theta)^{2}
$$

$\Delta \ell$ : Fisher statistical distance, reparametrization invariant.


$$
\Delta \ell>1
$$

Cramer-Rao bound 1945-46:
Given $p_{k}(\theta)$, the variance $\Delta \theta$ of any (unbiassed) estimate is bounded by $F(\theta)(\Delta \theta)^{2} \geq 1$. Attainable if statistics $\rightarrow \infty$.
$\Rightarrow$ Definition: neighbouring $p_{k}(\theta)$ and $p_{k}(\theta+\Delta \theta)$ are distinguishable if their distance

$$
\Delta \ell=1 .
$$

## I never understood

Fisher statistical distance

$$
\Delta \ell=\sqrt{F(\theta)} \Delta \theta
$$

is exact distinguishability measure (due to Cramer-Rao bound) between neighbouring $p_{k}(\theta)$ 's.

What about distant ones $p_{k}\left(\theta_{1}\right)$ and $p_{k}\left(\theta_{2}\right)$ ?
Their Fisher-distance:

$$
\ell_{12}=\int_{\theta_{1}}^{\theta^{2}} \sqrt{F(\theta)} d \theta
$$

What (the hell) is the meaning of $\ell_{12}$ ?

## From Fisher to Riemann metric in thermodynamics

Weinhold, Ruppeiner, Salamon et al, D. et al. 1976-
Fisher statistical distance $\ell$, multiparameter case, $\theta=\left\{\theta^{a}\right\}$ :

$$
\begin{gathered}
(d \ell)^{2}=g_{a b}(\theta) d \theta^{a} d \theta^{b} \\
g_{a b}(\theta)=\sum_{k} \frac{1}{p_{k}(\theta)} \frac{\partial p_{k}(\theta)}{\partial \theta^{a}} \frac{\partial p_{k}(\theta)}{\partial \theta^{b}}
\end{gathered}
$$

Riemannian metric (of distinguishablity) on space $\{\theta\}$ of $p_{k}(\theta)$ 's. Apply it to Gibbs-distributions $p_{E}(\theta)$ of the extensives $\left\{E_{a}\right\}$, parametrized by the intensives $\left\{\theta^{a}\right\}$, taking $V \rightarrow \infty$ :

$$
\begin{gathered}
g_{a b}(\theta)=\frac{\partial^{2} \phi(\theta)}{\partial \theta^{a} \partial \theta^{b}} \\
\phi\left(\theta_{1}, \theta_{2}, \ldots\right)=\text { thermodynamic potential }
\end{gathered}
$$

Spread (in Fisher metric) of thermodynamic fluctuations $\delta \theta^{a}$ is 1 :

$$
\left\langle(\delta \ell)^{2}\right\rangle=g_{a b}(\theta)\left\langle\delta \theta^{a} \delta \theta^{b}\right\rangle=\frac{\partial^{2} \phi(\theta)}{\partial \theta^{a} \partial \theta^{b}}\left\langle\delta \theta^{a} \delta \theta^{b}\right\rangle=1
$$

$\ell$ : thermodynamic length, $\min \ell$ : thermodynamic distance

## I noticed a Feldmann-Levine-Salamon paper 1986

# A Geometrical Measure for Entropy Changes 

Tova Feldmann, ${ }^{1}$ R. D. Levine, ${ }^{1}$ and Peter Salamon ${ }^{2}$

Received September 4, 1984; revision received September 6, 1985

The geometrical approach to statistical mechanics is used to discuss changes in entropy upon sequential displacements of the state of the system. An interpretation of the angle between two states in terms of entropy differences is thereby provided. A particular result of note is that any state can be resolved into a state of maximal entropy (both states having the same expectation values for the constraints) and an orthogonal component. A cosine law for the general case is also derived.

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Prof. Tova Feldman Dept. of Phys. Chem. The Hebrew University Jerusalem 91904
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1986. szeptember 18

## Tisztelt és Kedves Kolléga!

Nagy örömmel olvastam, hogy - munkatársaival -
a termodinamika Weinhold-Ruppeiner-... típusú geometriai elméletével foglalkozik. Szomorú vagyok viszont azért, mert errôl csak a budapesti konferencia poszterje után szereztem tudomást Lukács Béla kollégámtól, aki jelen volt de Ơnnel/Veled/ már nem tudott találkozni. En idén márciusban a Technionon voltam vendég és egy napra a Héber Egyetemet is meglátogattam /prof. Amin tanszékét/. Nagyon-nagyon sajnálom, hogy nem tudtunk egymásról, mert nekünk is több cikkünk jelent meg a Riemann-metricizációról. Ezeket most mellékelem, ha nem veszi tolakodásnak.

A Weinhold-Ruppeiner távolság és a Wootters távolság kapcsolatára mi is utaltunk, örülök hogy JSe 42, 1127 /1986/ cikkükben ez bővebben is ki van fejtve. fin elsôsorban kvantummechanikai méréselmélettel foglalkozom és annak idején

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Boldog lennék, ha - legalább irásban - kapcsolatba kerülnénk egymással. Kérem irja meg, ha angolul könnyebb.

Végezetül boldog UjÉvet kivánok!


Diósi Lajos

## Encounter 8yy later, "he" is she 1994

Conf. on Modern Developments in Thermodynamics (1994 Oct 2-7, Irsee, Germany)

Dinner time, talking to Peter Salamon in English, I don't remember why not in Hungarian, but English did matter, since in Hungarian there're no genders:

P: Didn't you talk to Tova?
I: Does he attend the conference?
P: Yes, but Tova is she, not he!
I: Really? Are you sure Tova is femail in Hebrew?
P : Be sure, it is.

## What (the hell) is the thermodynamic length $\ell_{12}$ ?

Thermodynamic distance $\Delta \ell=\sqrt{g_{a b}(\theta) \Delta \theta^{a} \Delta \theta^{b}}$ (with $g_{a b}=\partial_{a} \partial_{b} \phi$ ) is exact distinguishability measure between neighbouring equilibrium thermodynamic states $\theta$ and $\theta+\Delta \theta$, against their fluctuations $\delta \theta$.

What about distant states $\theta_{1}$ and $\theta_{2}$ ?
Thermodynamic length between them (parameter invariant):

$$
\ell_{12}=\int_{P} \sqrt{g_{a b}(\theta) d \theta^{a} d \theta^{b}}
$$

Attempts to interpret $\ell_{12}$ :
Bring state from $\theta_{1}$ to $\theta_{2}$ in finite time, under very specific conditions of Salamon, or of D. et al! For entropy production $S($ prod $)$ we got:

$$
\min S(\text { prod })=\text { const } \times\left(\min \ell_{12}\right)^{2}
$$

Under general conditions: we don't know what (...) $\ell_{12}$ is good for!

## From Fisher to quantum informatics

Fisher-distance on the full class of $p_{k}$ 's becomes simple:

$$
\begin{gathered}
(d \ell)^{2}=\sum_{k} \frac{\left(d p_{k}\right)^{2}}{p_{k}}=4 \sum\left(d x_{k}\right)^{2} \\
x_{k}=\sqrt{p_{k}}, \quad \sum_{k} x_{k}^{2}=1
\end{gathered}
$$

Euclidean metric on the unit hypersphere! Closed form(!) of distance: $2 \times$ spherical angle

$$
\ell_{12}=2 \arccos \left(\sum_{k} x_{1 k} x_{2 k}\right)=2 \arccos \left(\sum_{k} \sqrt{p_{1 k} p_{2 k}}\right) \quad \epsilon[0, \pi]
$$

With one eye on quantum, represent $p_{k}$ 's by diagonal density matrices

$$
\begin{gathered}
\hat{\rho}=\operatorname{diag}\left[p_{1}, p_{2}, \ldots, p_{k}, \ldots\right] . \\
\ell_{12}=2 \arccos \left(\operatorname{tr}\left(\sqrt{\hat{\rho}_{1} \hat{\rho}_{2}}\right) \quad \epsilon[0, \pi]\right.
\end{gathered}
$$

If quantum theory answers me what (...) Fisher-distance is good for?

## Statistical distinguishability distance of quantum

## states

Bures, Uhlmann, Braunstein \&Caves, ... 1970's-
Full Fisher-story has been extended for quantum mechanics

- quantum Fisher-"information", quantum Cramer-Rao bound
- statistical length in space of density matrices $\hat{\rho}$ : Bures-distance
- measure of undistingushability: fidelity $\mathcal{F}$

$$
\mathcal{F}\left(\left|\psi_{1}\right\rangle,\left|\psi_{2}\right\rangle\right)=\mid\left\langle\overline{\left.\psi_{1}\left|\psi_{2}\right\rangle\right|^{2}} \quad \epsilon[0,1]\right.
$$

For mixed states:

$$
\mathcal{F}\left(\hat{\rho}_{1}, \hat{\rho}_{2}\right)=\operatorname{tr} \sqrt{\sqrt{\hat{\rho}_{1}} \hat{\rho}_{2} \sqrt{\hat{\rho}_{1}}} \in[0,1]
$$

Uhlmann: If $\hat{\rho}_{1}, \hat{\rho}_{2}$ on a subsystem are reduced states of of the pure states $\left|\psi_{1}\right\rangle,\left|\psi_{2}\right\rangle$ then

$$
\mathcal{F}\left(\hat{\rho}_{1}, \hat{\rho}_{2}\right)=\min \mathcal{F}\left(\left|\psi_{1}\right\rangle,\left|\psi_{2}\right\rangle\right)
$$

where minimization is over all possible $\left|\psi_{1}\right\rangle,\left|\psi_{2}\right\rangle$, called purifications, of $\hat{\rho}_{1}, \hat{\rho}_{2}$ resp.

## Quantum informatic answer: what statistical distance is good for?

Fisher statistical distance between $\left\{p_{1 k}\right\}$ and $\left\{p_{2 k}\right\}$

$$
\ell_{12}=2 \arccos \left(\operatorname{tr}\left(\sqrt{\hat{\rho}_{1} \hat{\rho}_{2}}\right)\right.
$$

where $\hat{\rho}_{1}=\operatorname{diag}\left[\left\{p_{1 k}\right\}\right], \hat{\rho}_{2}=\operatorname{diag}\left[\left\{p_{2 k}\right\}\right]$.
Quantum informatic fidelity of $\hat{\rho}_{1}, \hat{\rho}_{2}$ :

$$
\mathcal{F}\left(\hat{\rho}_{1}, \hat{\rho}_{2}\right)=\operatorname{tr} \sqrt{\sqrt{\hat{\rho}_{1}} \hat{\rho}_{2} \sqrt{\hat{\rho}_{1}}}=\operatorname{tr} \sqrt{\hat{\rho}_{1} \hat{\rho}_{2}}
$$

Hence $\cos \left(\ell_{12} / 2\right)=\mathcal{F}\left(\hat{\rho}_{1}, \hat{\rho}_{2}\right)$. Uhlmann: $\cos \left(\ell_{12} / 2\right)=\min \left|\left\langle\psi_{1} \mid \psi_{2}\right\rangle\right|^{2}$. The best interpretation of the classical statistical length comes from quantum informatics:

## A best(?) answer to what (...) the statistical distance is

If everything is quantum, and classical statistics (e.g. our two probability distributions $p_{1}$ and $p_{2}$ ) come from famous intrinsic randomness of quantum systems, and if we have access to the full quantum system, then $p_{1}$ and $p_{2}$ are at least as distinguishable than any two statevectors satisfying

$$
\cos \left(\ell_{12} / 2\right)=\left|\left\langle\psi_{1} \mid \psi_{2}\right\rangle\right|^{2}
$$

Interpretation of a classical concept invokes quantum concepts. Strange!

Neither statistics, nor thermodynamics or quantum informatics have answered me: What (...) is statistical lenght $\ell_{12}$ is good for in its full generality?

After my 35yy of attempts, I'm still unhappy.
But, on the other hand, and above all:

# I'm glad to celebrate <br> Tova's 90th birthday <br> and <br> our 25 years in friendship and collaboration 

