The theory of statistical comparison:

from majorization to the “quantum Blackwell theorem” and beyond

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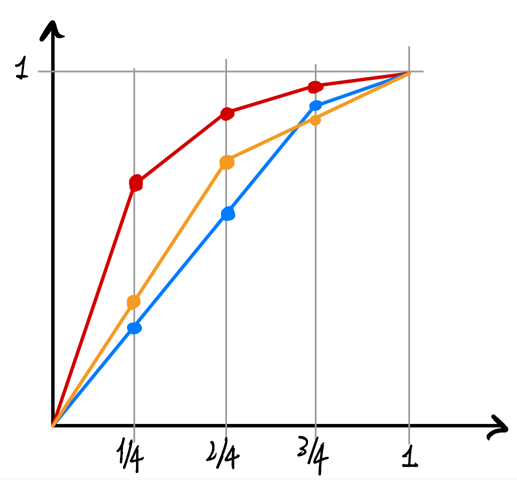
Prelude: majorization

Lorenz curves and majorization

- two probability distributions, $\boldsymbol{p} = (p_1, \ldots, p_n)$ and $\boldsymbol{q} = (q_1, \ldots, q_n)$
- truncated sums $P(k) = \sum_{i=1}^{k} p_i^{\downarrow}$ and $Q(k) = \sum_{i=1}^{k} q_i^{\downarrow}$, for all $k = 1, \ldots, n$
- $\boldsymbol{p}$ majorizes $\boldsymbol{q}$, i.e., $\boldsymbol{p} \succ_{\text{maj}} \boldsymbol{q}$, whenever $P(k) \geqslant Q(k)$, for all k
- minimal element: uniform distribution $\boldsymbol{e} = n^{-1}(1, 1, \cdots, 1)$

Hardy–Littlewood–Pólya, 1929

$\boldsymbol{p} \succ_{\text{maj}} \boldsymbol{q} \iff \boldsymbol{q} = M\boldsymbol{p}$, for some bistochastic matrix M.



$(x_k, y_k) = (k/n, P(k)), \quad 1 \leqslant k \leqslant n$

Blackwell's information preorder

Statistical experiments

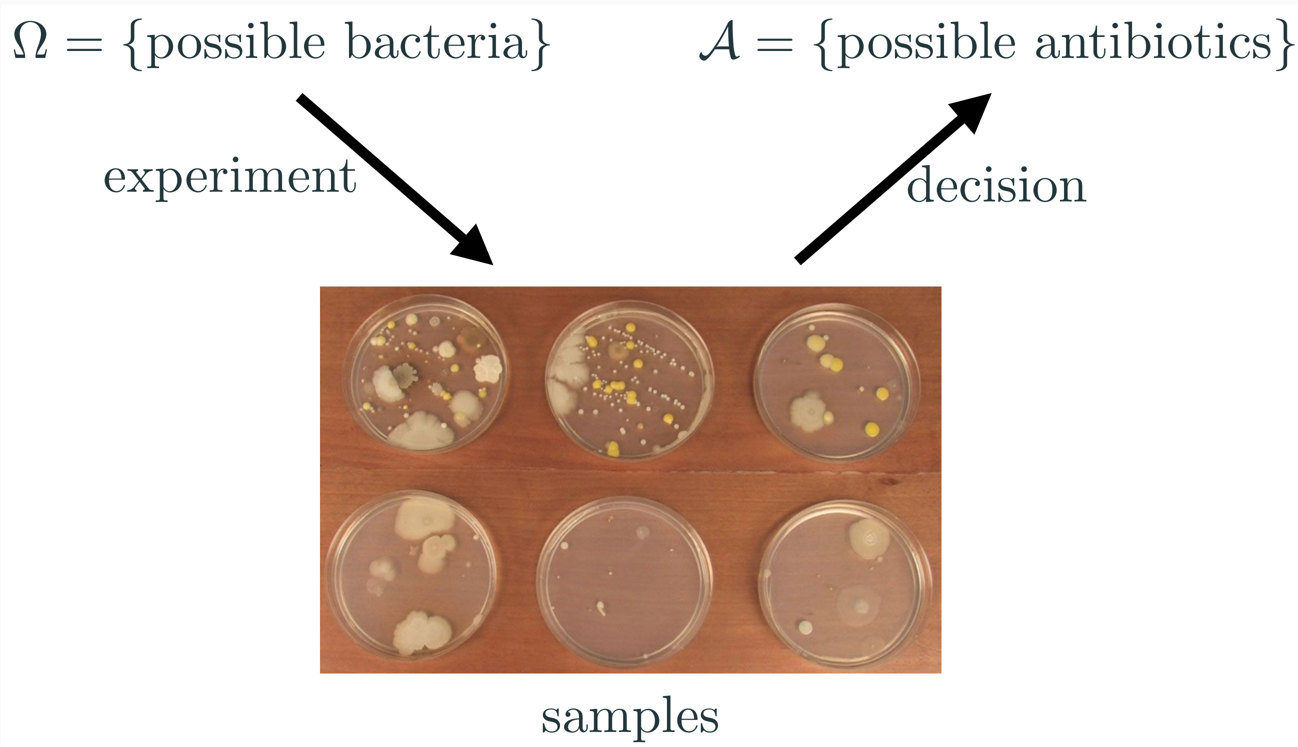
Lucien Le Cam (1924-2000)

"The basic structures in the whole affair are systems that Blackwell called experiments, and transitions between them.

An experiment is a mathematical abstraction intended to describe the basic feature of an observational process if that process is contemplated in advance of its implementation."

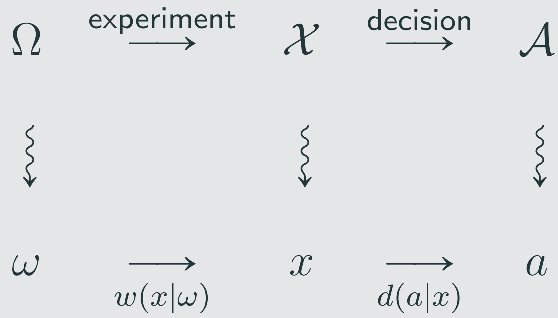
Lucien Le Cam (1984)

A concrete example...



...and its abstract formulation

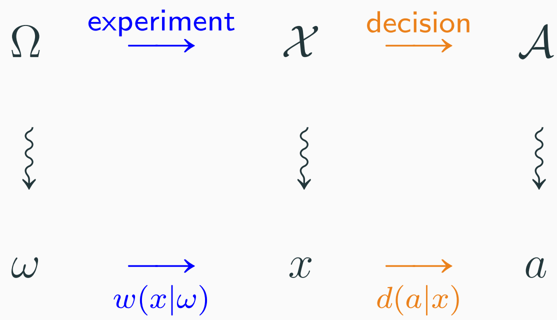
Definition (Statistical models and decision problems)



- *parameter* set $\Omega = \{\omega\}$, *sample* set $\mathcal{X} = \{x\}$, *action* set $\mathcal{A} = \{a\}$
- a **statistical model/experiment** is a triple $\mathbf{w} = \langle \Omega, \mathcal{X}, w(x|\omega) \rangle$
- a **decision** is a triple $\langle \mathcal{X}, \mathcal{A}, d(a|x) \rangle$
- a **statistical decision problem/game** is a triple $\mathbf{g} = \langle \Omega, \mathcal{A}, c \rangle$, where $c : \Omega \times \mathcal{A} \to \mathbb{R}$ is a payoff function

Playing statistical games with experiments

- the experiment/model is the **resource**: it is given
- the decision is the **transition**: it can be optimized



Definition

The **(expected) maximin payoff** of a statistical model $\mathbf{w} = \langle \Omega, \mathcal{X}, w \rangle$ w.r.t. a decision problem $\mathbf{g} = \langle \Omega, \mathcal{A}, c \rangle$ is given by

$$c^*_{\mathbf{g}}(\mathbf{w}) \stackrel{\text{def}}{=} \max_{d(a|x)} \min_{\omega} \sum_{a,x} c(\omega, a) d(a|x) w(x|\omega) \ .$$

Comparison of statistical models 1/2

$$\mathbf{w} = \langle \Omega, \mathcal{X}, w(x|\omega) \rangle$$

$$\begin{array}{ccccc} \Omega & \xrightarrow{\text{experiment}} & \mathcal{X} & \xrightarrow{\text{decision}} & \mathcal{A} \\ \rightsquigarrow & & \rightsquigarrow & & \rightsquigarrow \\ \omega & \xrightarrow[w(x|\omega)]{} & x & \xrightarrow[d(a|x)]{} & a \end{array}$$

$$\mathbf{w}' = \langle \Omega, \mathcal{Y}, w'(y|\omega) \rangle$$

$$\begin{array}{ccccc} \Omega & \xrightarrow{\text{experiment}} & \mathcal{Y} & \xrightarrow{\text{decision}} & \mathcal{A} \\ \rightsquigarrow & & \rightsquigarrow & & \rightsquigarrow \\ \omega & \xrightarrow[w'(y|\omega)]{} & y & \xrightarrow[d'(a|y)]{} & a \end{array}$$

For a fixed decision problem $\mathbf{g} = \langle \Omega, \mathcal{A}, c \rangle$, the payoffs $c^*_{\mathbf{g}}(\mathbf{w})$ and $c^*_{\mathbf{g}}(\mathbf{w}')$ can always be ordered (they are just real numbers).

Comparison of statistical models 2/2

Definition (Information preorder)

If the model $\mathbf{w} = \langle \Omega, \mathcal{X}, w \rangle$ is better than model $\mathbf{w}' = \langle \Omega, \mathcal{Y}, w' \rangle$ *for all* decision problems $\mathbf{g} = \langle \Omega, \mathcal{A}, c \rangle$, that is,

$$c^*_{\mathbf{g}}(\mathbf{w}) \geqslant c^*_{\mathbf{g}}(\mathbf{w}'), \quad \forall \mathbf{g} \, ,$$

then we say that $\mathbf{w}$ is (always) more informative than $\mathbf{w}'$, and write

$$\mathbf{w} \succ_{\text{info}} \mathbf{w}' \, .$$

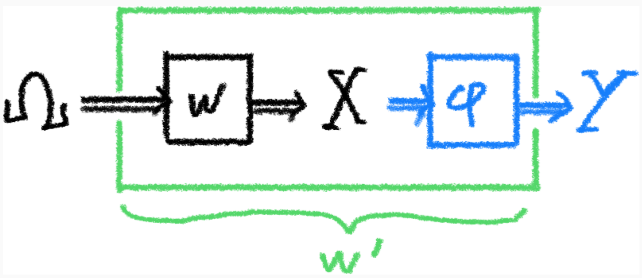
Can we visualize the information preorder more concretely?

Information preorder = statistical sufficiency

Theorem (Blackwell, 1953)

Given two statistical experiments **w** $= \langle \Omega, \mathcal{X}, w \rangle$ *and* **w**$'$ $= \langle \Omega, \mathcal{Y}, w' \rangle$, *the following are equivalent:*

1. **w** $\succ_{\text{info}}$ **w**$'$;
2. $\exists$ *cond. prob. dist.* $\varphi(y|x)$ *such that* $w'(y|\omega) = \sum_x \varphi(y|x) w(x|\omega)$ *for all* y *and* ω.



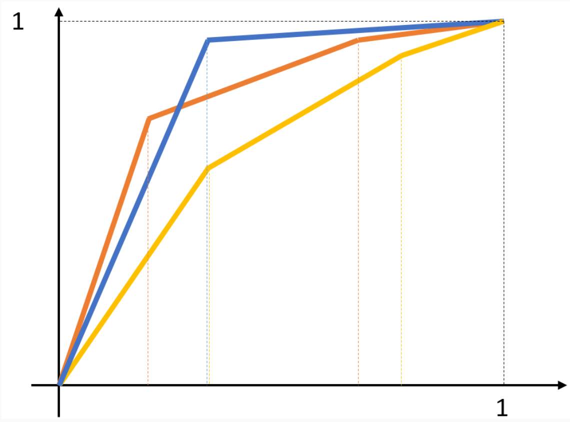
David Blackwell (1919-2010)

The case of dichotomies (a.k.a. relative majorization)

- for $\Omega = \{1, 2\}$, we compare two dichotomies, i.e., two pairs of probability distributions $(\boldsymbol{p}_1, \boldsymbol{p}_2)$ and $(\boldsymbol{q}_1, \boldsymbol{q}_2)$, of dimension m and n, respectively
- relabel entries such that ratios p_1^i/p_2^i and q_1^j/q_2^j are nonincreasing
- for $\omega \in \{1, 2\}$, let the truncated sums be $P_\omega(k) = \sum_{i=1}^k p_\omega^i$ and $Q_\omega(k) = \sum_{j=1}^k q_\omega^j$
- write $(\boldsymbol{p}_1, \boldsymbol{p}_2) \succ_{\text{maj}} (\boldsymbol{q}_1, \boldsymbol{q}_2)$ whenever the relative Lorenz curve of the former is never below that of the latter

Blackwell, 1953

For dichotomies, $\succ_{\text{maj}} \iff \succ_{\text{info}} \iff \exists$ stochastic matrix M s.t. $\boldsymbol{q}_\omega = M\boldsymbol{p}_\omega$



$(x_k, y_k) = (P_2(k), P_1(k)), \quad 1 \leqslant k \leqslant n$

Quantum versions

Quantum statistical decision theory (Holevo, 1973)

classical case	**quantum case**
• decision problems $\mathbf{g} = \langle \Omega, \mathcal{A}, c \rangle$	• decision problems $\mathbf{g} = \langle \Omega, \mathcal{A}, c \rangle$
• models $\mathbf{w} = \langle \Omega, \mathcal{X}, \{w(x\|\omega)\} \rangle$	• quantum models $\mathcal{E} = \langle \Omega, \mathcal{H}_S, \{\rho_S^\omega\} \rangle$
• decisions $d(a\|x)$	• POVMs $\{P_S^a : a \in \mathcal{A}\}$
• $c^*_{\mathbf{g}}(\mathbf{w}) = \max_{d(a\|x)} \min_{\omega} \cdots$	• $c^*_{\mathbf{g}}(\mathcal{E}) = \max_{\{P_S^a\}} \min_{\omega} \sum_a c(\omega, a) \operatorname{Tr}[\rho_S^\omega \ P_S^a]$

Quantum statistical morphisms (FB, CMP 2012)

Definition (Generalized decisions)

Given a quantum statistical model (QSM) $\mathcal{E} = \langle \Omega, \mathcal{H}_S, \{\rho_S^\omega\} \rangle$, a family of operators $\{Z_S^a\}_a$ is said to be an $\mathcal{E}$-decision if and only if $\exists$ POVM $\{P_S^a\}_a$ s.t.

$$\operatorname{Tr}[\rho_S^\omega \ Z_S^a] = \operatorname{Tr}[\rho_S^\omega \ P_S^a] \ , \quad \forall \omega, \forall a \ .$$

Definition (Statistical morphisms)

Given two QSMs $\mathcal{E} = \langle \Omega, \mathcal{H}_S, \{\rho_S^\omega\} \rangle$ and $\mathcal{E}' = \langle \Omega, \mathcal{H}_{S'}, \{\sigma_{S'}^\omega\} \rangle$, a linear map $\mathcal{M} : \mathsf{L}(\mathcal{H}_S) \to \mathsf{L}(\mathcal{H}_{S'})$ is said to be an $\mathcal{E} \to \mathcal{E}'$ quantum statistical morphism iff

1. $\mathcal{M}$ is trace-preserving;
2. $\mathcal{M}(\rho_A^\omega) = \sigma_{S'}^\omega$, for all $\omega \in \Omega$;
3. the trace-dual map $\mathcal{M}^\dagger : \mathsf{L}(\mathcal{H}_{S'}) \to \mathsf{L}(\mathcal{H}_S)$ maps $\mathcal{E}'$-decisions into $\mathcal{E}$-decisions.

Quantum statistical comparison (FB, CMP 2012)

Given two QSMs $\mathcal{E} = \langle \Omega, \mathcal{H}_S, \{\rho_S^\omega\} \rangle$ and $\mathcal{E}' = \langle \Omega, \mathcal{H}_{S'}, \{\sigma_{S'}^\omega\} \rangle$

- information ordering: $\mathcal{E} \succ_{\text{info}} \mathcal{E}' \overset{\text{def}}{\iff} c^*_{\mathbf{g}}(\mathcal{E}) \geqslant c^*_{\mathbf{g}}(\mathcal{E}')$ for all $\mathbf{g}$
- *complete* information ordering: $\mathcal{E} \succcurlyeq_{\text{info}} \mathcal{E}' \overset{\text{def}}{\iff} \mathcal{E} \otimes \mathcal{F} \succ_{\text{info}} \mathcal{E}' \otimes \mathcal{F}$ for all ancillary models $\mathcal{F} = \langle \Theta, \mathcal{H}_A, \{\tau_A^\theta\} \rangle$

Theorem 1/3: $\mathcal{E} \succ_{\text{info}} \mathcal{E}'$ iff there exists a *quantum statistical morphism* $\mathcal{M} : \mathsf{L}(\mathcal{H}_S) \to \mathsf{L}(\mathcal{H}_{S'})$ such that $\mathcal{M}(\rho_S^\omega) = \sigma_{S'}^\omega$, $\forall \omega \in \Omega$

Theorem 2/3: $\mathcal{E} \succcurlyeq_{\text{info}} \mathcal{E}'$ iff there exists a *completely positive trace-preserving linear map* $\mathcal{N} : \mathsf{L}(\mathcal{H}_S) \to \mathsf{L}(\mathcal{H}_{S'})$ such that $\mathcal{N}(\rho_S^\omega) = \sigma_{S'}^\omega$, for all $\omega \in \Omega$

Theorem 3/3: if $\mathcal{E}'$ is *commutative*, that is, if $[\sigma^{\omega_1}, \sigma^{\omega_2}] = 0$ for all $\omega_1, \omega_2 \in \Omega$, then $\mathcal{E} \succcurlyeq_{\text{info}} \mathcal{E}'$ iff $\mathcal{E} \succ_{\text{info}} \mathcal{E}'$

Quantum dichotomies and quantum majorization

Classical hypothesis testing

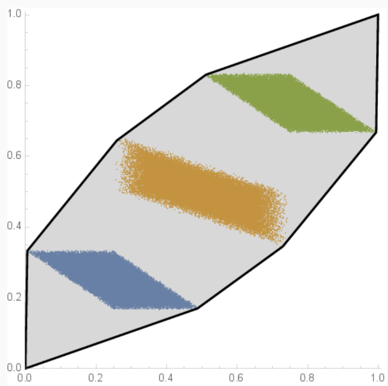
- parameter set: $\Omega = \{1, 2\}$
- sample space: $\mathcal{X} = \{1, 2, \ldots, n\}$
- two possible hypotheses to test: $\boldsymbol{p}_1$ or $\boldsymbol{p}_2$
- an effect is a vector $\boldsymbol{t} = (t_1, t_2, \cdots, t_n)$ s.t. $0 \leqslant t_i \leqslant 1$, $\forall i$
- a test is a pair $(\boldsymbol{t}, \boldsymbol{e} - \boldsymbol{t})$, i.e., $\mathcal{A} \equiv \Omega$
- expected probability of true positive: $\boldsymbol{t} \cdot \boldsymbol{p}_1$
- expected probability of false positive: $\boldsymbol{t} \cdot \boldsymbol{p}_2$

How to capture the "distinguishability" of the two hypotheses?

Relative testing region and relative Lorenz curve

The testing region of $\boldsymbol{p}_1$ relative to $\boldsymbol{p}_2$ is defined as the set

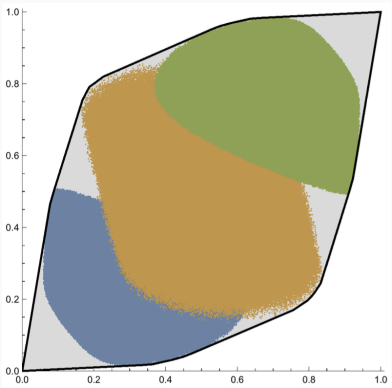
$$\mathcal{T}(\boldsymbol{p}_1 \| \boldsymbol{p}_2) \stackrel{\text{def}}{=} \{(x, y) = (\boldsymbol{t} \cdot \boldsymbol{p_2}, \boldsymbol{t} \cdot \boldsymbol{p_1}) : \boldsymbol{t} \text{ effect}\}$$



Theorem (Renes, JMP, 2016)

The relative Lorenz curve of $(\boldsymbol{p}_1, \boldsymbol{p}_2)$ coincides with the upper boundary of $\mathcal{T}(\boldsymbol{p}_1 \| \boldsymbol{p}_2)$, so that $(\boldsymbol{p}_1, \boldsymbol{p}_2) \succ_{\text{maj}} (\boldsymbol{q}_1, \boldsymbol{q}_2) \iff \mathcal{T}(\boldsymbol{p}_1 \| \boldsymbol{p}_2) \supseteq \mathcal{T}(\boldsymbol{q}_1 \| \boldsymbol{q}_2)$.

Quantum relative Lorenz curve



Definition (FB and G. Gour, 2016)

Given two density matrices ρ_1 and ρ_2 on Hilbert space $\mathcal{H}$, the quantum testing region of ρ_1 relative to ρ_2 is defined as

$$\mathcal{T}(\rho_1\|\rho_2) \stackrel{\text{def}}{=} \left\{(x,y) = (\mathrm{Tr}[E\ \rho_2]\,, \mathrm{Tr}[E\ \rho_1])\right\}\,,$$

where E can vary over all effects on $\mathcal{H}$ (i.e. $0 \leqslant E \leqslant \mathbb{1}$).

The quantum Lorenz curve of ρ_1 relative to ρ_2 is the upper boundary of $\mathcal{T}(\rho_1\|\rho_2)$ so that

$$(\rho_1,\rho_2) \succ_{\mathrm{maj}} (\sigma_1,\sigma_2) \stackrel{\text{def}}{\iff} \mathcal{T}(\rho_1\|\rho_2) \supseteq \mathcal{T}(\sigma_1\|\sigma_2)$$

Remark. A quantum Lorenz curve may have strictly convex sections.

Equivalent characterizations of $\succ_{\mathrm{maj}}$ 1/2

Definition

Given two density matrices ρ and σ, we define the hypothesis testing relative entropy (FB, Datta; 2010)

$$D_H^{\epsilon}(\rho\|\sigma) := -\log \min_{\substack{0\leqslant E\leqslant \mathbb{1}\\ \mathrm{Tr}[\rho E]\geqslant 1-\epsilon}} \mathrm{Tr}[\sigma E]\ , \quad \epsilon\in[0,1]$$

and the Hilbert α-divergence (FB, Gour; 2017)

$$H_\alpha(\rho\|\sigma) := \frac{\alpha}{\alpha-1}\log \sup_{\frac{1}{\alpha}\mathbb{1}\leqslant E\leqslant \mathbb{1}} \frac{\mathrm{Tr}[\rho E]}{\mathrm{Tr}[\sigma E]}\ , \quad \alpha>1\ ,$$

with $H_1(\rho\|\sigma) := \lim_{\alpha\to 1^+} H_\alpha(\rho\|\sigma)$ and $H_\infty(\rho\|\sigma) := \lim_{\alpha\to\infty} H_\alpha(\rho\|\sigma)$

Equivalent characterizations of $\succ_{\rm maj}$ 2/2

Theorem (FB, Gour; 2017)

Given two quantum dichotomies (ρ, σ) *and* (ρ', σ') *(possibly on different Hilbert spaces), the following are equivalent:*

1. $\mathcal{T}(\rho\|\sigma) \supseteq \mathcal{T}(\rho'\|\sigma')$, *i.e.,* $(\rho, \sigma) \succ_{\rm maj} (\rho', \sigma')$
2. $D_H^\epsilon(\rho\|\sigma) \geqslant D_H^\epsilon(\rho'\|\sigma')$, *for all* $\epsilon \in [0, 1]$
3. $H_\alpha(\rho\|\sigma) \geqslant H_\alpha(\rho'\|\sigma')$ *and* $H_\alpha(\sigma\|\rho) \geqslant H_\alpha(\sigma'\|\rho')$, *for all* $\alpha \geqslant 1$
4. $\|\rho - t\sigma\|_1 \geqslant \|\rho' - t\sigma'\|_1$, *for all* $t \geqslant 0$

The problem with quantum dichotomies

classically:
$(\boldsymbol{p}, \boldsymbol{q}) \succ_{\rm maj} (\boldsymbol{p}', \boldsymbol{q}') \overset{\Longleftarrow}{\Longrightarrow} (\boldsymbol{p}, \boldsymbol{q}) \succ_{\rm info} (\boldsymbol{p}', \boldsymbol{q}') \overset{\Longleftarrow}{\Longrightarrow} (\boldsymbol{p}, \boldsymbol{q}) \ggcurly_{\rm info} (\boldsymbol{p}', \boldsymbol{q}')$

the same equivalences hold also if both (ρ, σ) and (ρ', σ') are qubit dichotomies (Alberti and Uhlmann, 1980)

however, in general:
$(\rho, \sigma) \succ_{\rm maj} (\rho', \sigma') \overset{\Longleftarrow}{\not\Longrightarrow} (\rho, \sigma) \succ_{\rm info} (\rho', \sigma') \overset{\Longleftarrow}{\not\Longrightarrow} (\rho, \sigma) \ggcurly_{\rm info} (\rho', \sigma')$
(counterexample by Matsumoto, 2014)

Problem: can we find conditions that are weaker, but easier to work with?

Information-theoretic treatment

Theorem (Matsumoto, 2010; FB, D. Sutter, M. Tomamichel, 2019)

Given two dichotomies (ρ, σ) *and* (ρ', σ')*, if*

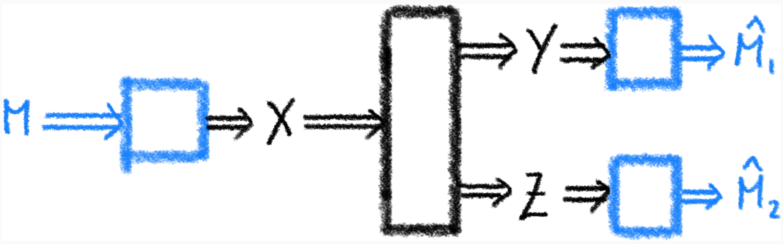
$$D(\rho\|\sigma) > D(\rho'\|\sigma') \ ,$$

then there exists $\gamma > 0$, $n_0 \in \mathbb{N}$, *and a sequence of CPTP linear maps* $\{\mathcal{E}_n\}_{n\in\mathbb{N}}$*, such that*

$$\begin{cases} \mathcal{E}_n(\sigma^{\otimes n}) = \sigma'^{\otimes n} & \forall n \in \mathbb{N} \ , \\ \|\mathcal{E}_n(\rho^{\otimes n}) - \rho'^{\otimes n}\|_1 \leqslant e^{-\gamma n} & \forall n \geqslant n_0 \ . \end{cases}$$

Applications in information theory

Classical broadcast channels



How to capture the idea that Y carries more information than Z?

(i) (stochastically) degradable: $\exists$ channel $Y \to Z$

(ii) less noisy: for all M, $H(M|Y) \leqslant H(M|Z)$

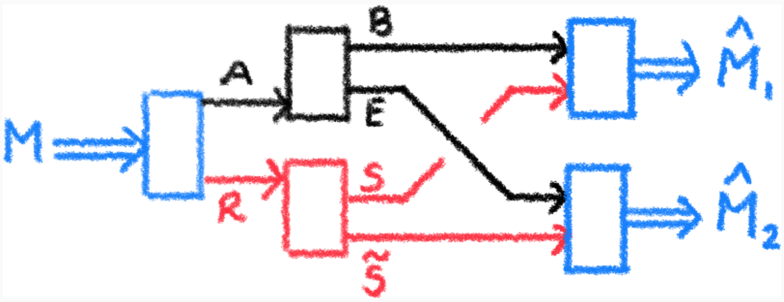
(iii) less ambiguous: for all M, $\max \mathbb{P}\{\hat{M}_1 = M\} \geqslant \max \mathbb{P}\{\hat{M}_2 = M\}$

(iv) less ambiguous (reformulation): for all M, $H_{\min}(M|Y) \leqslant H_{\min}(M|Z)$

Theorem (Körner–Marton, 1977; FB, 2016)

less noisy $\overset{\not\Longrightarrow}{\Longleftarrow}$ *degradable* $\Longleftrightarrow$ *less ambiguous*

Quantum broadcast channels



(i) (CPTP) degradable: $\exists$ channel $B \to E$

(ii) *completely* less noisy: for all M and all symmetric side-channels $R \to S\tilde{S}$, $H(M|BS) \leqslant H(M|E\tilde{S})$

(iii) *completely* less ambiguous: for all M and all symmetric side-channels $R \to S\tilde{S}$, $H_{\min}(M|BS) \leqslant H_{\min}(M|E\tilde{S})$

Theorem (FB–Datta–Strelchuk, 2014)

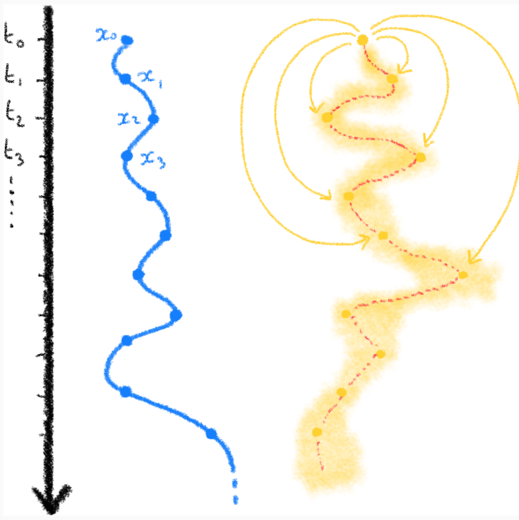
completely less noisy $\overset{\not\Longrightarrow}{\Longleftarrow}$ *degradable* $\Longleftrightarrow$ *completely less ambiguous*

Applications in open quantum systems dynamics

Discrete-time stochastic processes

Formulation of the problem:

- for $i \in \mathbb{N}$, let x_i index the state of a system at time $t = t_i$
- given the system's initial state at time $t = t_0$, the process is fully predicted by the conditional distribution $p(x_N, \ldots, x_1 | x_0)$
- if the system evolving is quantum, we only have a quantum dynamical mapping $\left\{ \mathcal{N}^{(i)}_{Q_0 \to Q_i} \right\}_{i \geqslant 1}$
- the process is divisible if there exist channels $\mathcal{D}^{(i)}$ such that $\mathcal{N}^{(i+1)} = \mathcal{D}^{(i)} \circ \mathcal{N}^{(i)}$ for all $i \geqslant 1$
- **problem**: to provide a *fully information-theoretic characterization* of divisibility

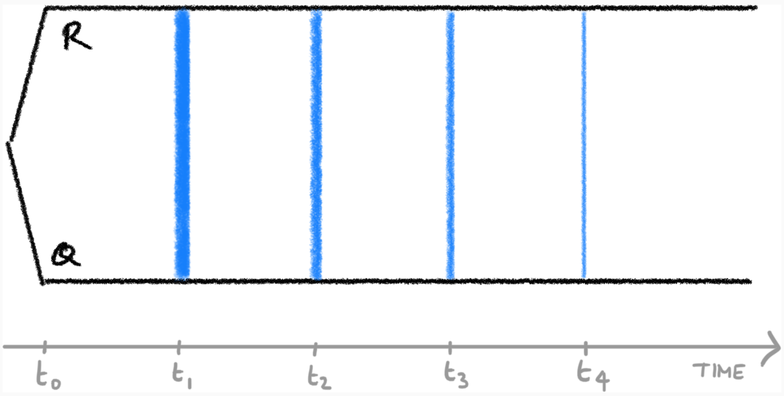


Divisibility as "information flow"

Theorem (FB–Datta, 2016; FB, 2018)

Given an initial open quantum system Q_0, a quantum dynamical mapping $\left\{\mathcal{N}^{(i)}_{Q_0\to Q_i}\right\}_{i\geqslant 1}$ is divisible if and only if*, for any initial state ω_{RQ_0},*

$$H_{\min}(R|Q_1) \leqslant H_{\min}(R|Q_2) \leqslant \cdots \leqslant H_{\min}(R|Q_N) .$$



Applications in quantum thermodynamics

Quantum thermodynamics from relative majorization

Basic idea (FB, arXiv:1505.00535)

Thermal accessibility $\rho \to \sigma$ can be characterized as the statistical comparison between quantum dichotomies (ρ, γ) and (σ, γ), for γ thermal state

Two main problems:

- for dimension larger than 2 and $[\sigma, \gamma] \neq 0$, we need a complete (i.e., extended) comparison
- moreover, Gibbs-preserving channels can create coherence between energy levels, while a truly thermal operation should not

Complete comparison of quantum dichotomies 1/2

Definition (ON/OFF channels)

Given a d-dimensional quantum dichotomy $\mathcal{E} = (\rho, \gamma)$, we define the corresponding ON/OFF channel $\mathcal{N}_{\mathcal{E}} : \mathscr{L}(\mathbb{C}^2) \to \mathscr{L}(\mathbb{C}^d)$ as

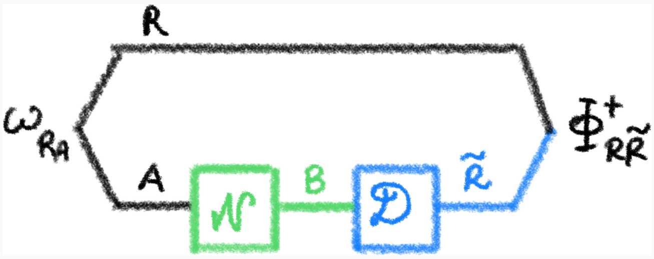
$$\mathcal{N}_{\mathcal{E}}(\cdot) := \gamma \langle 0| \cdot |0\rangle + \rho \langle 1| \cdot |1\rangle$$

Complete comparison of quantum dichotomies 2/2

For a quantum channel $\mathcal{N} : A \to B$ and a state ω_{RA}, define the singlet fraction as

$$\Phi^*_\omega(\mathcal{N}) := \max_{\mathcal{D}:B\to\tilde{R}} \langle \Phi^+_{R\tilde{R}} | (\mathrm{id}_R \otimes \mathcal{D} \circ \mathcal{N})(\omega_{RA}) | \Phi^+_{R\tilde{R}} \rangle \,,$$

where $\mathcal{D}$ is a decoding quantum channel with output system $R \cong \tilde{R}$



Theorem (FB, 2015)

Given two quantum dichotomies $\mathcal{E} = (\rho_1, \rho_2)$ and $\mathcal{F} = (\sigma_1, \sigma_2)$, let $\mathcal{N}_\mathcal{E}$ and $\mathcal{N}_\mathcal{F}$ the corresponding ON/OFF channels. Then, $\mathcal{E} \succcurlyeq \mathcal{F}$ if and only if

$$\Phi^*_\omega(\mathcal{N}_\mathcal{E}) \geqslant \Phi^*_\omega(\mathcal{N}_\mathcal{F}) \,, \quad \forall \omega$$

Dealing with quantum coherence (sketch)

For quantum dichotomies $\mathcal{E} = (\rho, \gamma)$ and $\mathcal{F} = (\sigma, \gamma)$ and group $\mathscr{T} = \{e^{-it\log\gamma}\}_{t\in\mathbb{R}}$, we write $\mathcal{E} \succcurlyeq_\mathscr{T} \mathcal{F}$ iff $\exists$ CPTP linear $\mathcal{M}$ such that:

(i) $\mathcal{M}(\rho) = \sigma$ and $\mathcal{M}(\gamma) = \gamma$;

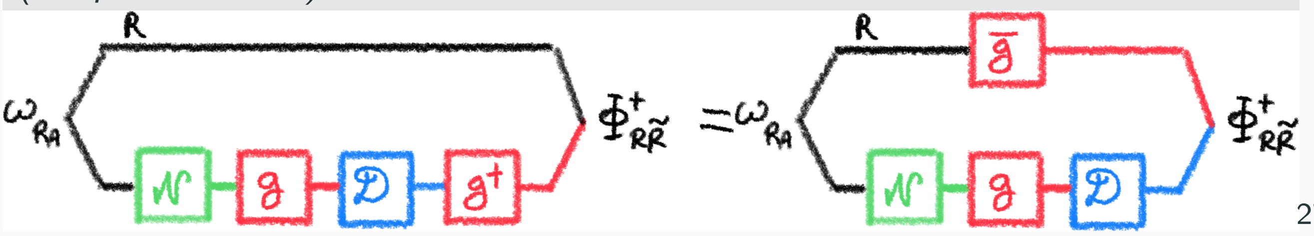
(ii) $\mathcal{M}(U_t \cdot U_t^\dagger) = U_t \mathcal{M}(\cdot) U_t^\dagger$, for all $t \in \mathbb{R}$

Theorem (Gour–Jennings–FB–Duan–Marvian, 2018)

$\mathcal{E} \succcurlyeq_\mathscr{T} \mathcal{F}$ if and only if

$$\widetilde{\Phi}^*_\omega(\mathcal{N}_\mathcal{E}) \geqslant \widetilde{\Phi}^*_\omega(\mathcal{N}_\mathcal{F}) \,, \quad \forall \omega$$

(see picture below)



Conclusions

Conclusions

- the theory of statistical comparison studies **morphisms** (preorders) of one "statistical object" X into another "statistical object" Y
- equivalent conditions are given in terms of (finitely or infinitely many) **monotones**, e.g., $f_i(X) \geqslant f_i(Y)$
- such monotones quantify the **resources** at stake in the operational framework at hand

Thank you