

# A Note on Distances between Probabilistic and Quantum distributions

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Where we are, so far

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## Distances in probability theory

- ▶ Much usage of metrics in probabilistic computation, eg.
  - measuring the behavioural similarity of states in probabilistic transition systems (Panangaden, Desharnais, Mislove, Worrell, van Breughel, König . . .)
  - evaluating performance and uncertainty of Bayesian network models
- ▶ Nice systematic description of metrics via universality in LICS'16 paper of Madare, Panangaden, Plotkin
- ▶ **Here:** two loosely connected topics, concrete and abstract:
  - “entwinedness” measure for classical and quantum probability
  - metric versions of “state-and-effect” triangles

## Discrete probability distributions

### Notation

- ▶ Fair coin:  $\frac{1}{2}|H\rangle + \frac{1}{2}|T\rangle$
- ▶ Fair dice:  $\frac{1}{6}|1\rangle + \frac{1}{6}|2\rangle + \frac{1}{6}|3\rangle + \frac{1}{6}|4\rangle + \frac{1}{6}|5\rangle + \frac{1}{6}|6\rangle$

### ket notation

- ▶  $|-\rangle$  is pure syntactic sugar — stemming from quantum
- ▶ more confusing to omit them, as in:  $\frac{1}{6}1 + \frac{1}{6}2 + \frac{1}{6}3 + \frac{1}{6}4 + \frac{1}{6}5 + \frac{1}{6}6$
- ▶ Write  $\mathcal{D}(X)$  for the set of such probability distributions  $\sum_i r_i |x_i\rangle$  where  $x_i \in X$ ,  $r_i \in [0, 1]$  with  $\sum_i r_i = 1$
- ▶ This  $\mathcal{D}$  is a monad on sets; its algebras are **convex sets**
- ▶ Distributions  $\omega \in \mathcal{D}(X)$  will often be called **states** of  $X$

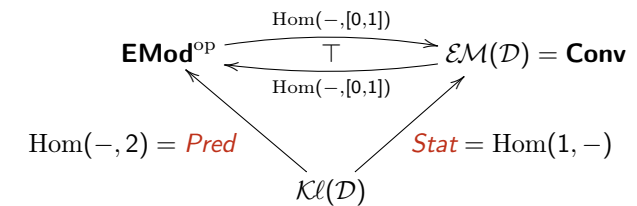


## Predicates for probabilistic logic

- ▶ A **predicate** on a set  $X$  is a function  $p: X \rightarrow [0, 1]$ 
  - It is called **sharp** (non-fuzzy) if  $p(x) \in \{0, 1\}$  for each  $x \in X$
- ▶ Basic **effect module structure** on these predicates:
  - true and false, as constant-one and constant-zero
  - orthosupplement  $p^\perp(x) = 1 - p(x)$
  - partial sum  $(p \oplus q)(x) = p(x) + q(x)$  if  $p(x) + q(x) \leq 1$  for all  $x$
  - scaling  $(r \cdot p)(x) = r \cdot p(x)$ , for  $r \in [0, 1]$
- ▶ Each Kleisli map  $f: X \rightarrow \mathcal{D}(Y)$  gives a **predicate transformation** map  $f^*: [0, 1]^Y \rightarrow [0, 1]^X$  preserving the effect module structure
- ▶ Also sequential and parallel conjunction, via multiplication

## State-and-effect triangle

Much is summarised in:



This is a much more general pattern in **state- and predicate-transformation** semantics of computation — including quantum



## Combining states and predicates

Let  $\omega \in \mathcal{D}(X)$  be state/distribution,  $p \in [0, 1]^X$  a predicate, both on  $X$ .

- ▶ **Validity**  $\omega \models p$ , in  $[0, 1]$ 
  - defined as  $\sum_x \omega(x) \cdot p(x)$
  - also known as expected value of  $p$  in state  $\omega$
- ▶ **Conditioning**  $\omega|_p$ , in  $\mathcal{D}(X)$ 
  - assuming validity  $\omega \models p$  is non-zero
  - defined as:  $\omega|_p = \sum_x \frac{\omega(x) \cdot p(x)}{\omega \models p} |x\rangle$

## Validity and conditioning example

- ▶ Take  $X = \{1, 2, 3, 4, 5, 6\}$  with state  $\text{dice} \in \mathcal{D}(X)$ 
  - recall  $\text{dice} = \frac{1}{6}|1\rangle + \frac{1}{6}|2\rangle + \frac{1}{6}|3\rangle + \frac{1}{6}|4\rangle + \frac{1}{6}|5\rangle + \frac{1}{6}|6\rangle$
- ▶ Take **even** predicate  $E \in [0, 1]^X$ ; it's sharp, given by:
  - $E(1) = E(3) = E(5) = 0$ ,  $E(2) = E(4) = E(6) = 1$
  - define **odd** via orthosupplement:  $O = E^\perp$
- ▶  $\text{dice} \models E = \frac{1}{2}$
- ▶  $\text{dice}|_E = \frac{1/6}{1/2}|2\rangle + \frac{1/6}{1/2}|4\rangle + \frac{1/6}{1/2}|6\rangle = \frac{1}{3}|2\rangle + \frac{1}{3}|4\rangle + \frac{1}{3}|6\rangle$
- ▶  $\text{dice}|_E \models O = 0$



## A distance question, asked with Fabio Zanasi (MFCS'17)

- ▶ How different are  $\omega$  and  $\omega|_p$  ?
- ▶ This can be formalised as  $d(\omega, \omega|_p)$
- ▶ This number captures the **influence** of  $p$  on  $\omega$
- ▶ Used to describe d-separation and blocking of influence in Bayesian networks



## Products and marginalisations of states

- ▶ For states  $\omega_1 \in \mathcal{D}(X_1)$  and  $\omega_2 \in \mathcal{D}(X_2)$  we can form the **product** state  $\omega_1 \otimes \omega_2 \in \mathcal{D}(X_1 \times X_2)$  by:

$$\omega_1 \otimes \omega_2 = \sum_{(x_1, x_2)} (\omega_1(x_1) \cdot \omega_2(x_2)) |x_1, x_2\rangle$$

- ▶ For a **joint** state  $\sigma \in \mathcal{D}(X_1 \otimes X_2)$  there are marginalisations  $M_i(\sigma) \in \mathcal{D}(X_i)$ , given by:

$$M_1(\sigma) = \sum_{x_1} (\sum_{x_2} \sigma(x_1, x_2)) |x_1\rangle \quad M_2(\sigma) = \sum_{x_2} (\sum_{x_1} \sigma(x_1, x_2)) |x_2\rangle$$

- ▶ It is to easy that marginalisation after product returns the originals:

$$M_1(\omega_1 \otimes \omega_2) = \omega_1 \quad M_2(\omega_1 \otimes \omega_2) = \omega_2$$

- ▶ But what about the other way around: product of marginals?



## Entwinedness (cf. entanglement), and distance

- ▶ In general:  $\sigma \neq M_1(\sigma) \otimes M_2(\sigma)$ 
  - “the whole is more than the sum of its parts”
- ▶ Example:  $\sigma = \frac{1}{2}|a, 1\rangle + \frac{1}{2}|b, 2\rangle$ , then:
  - $M_1(\sigma) = \frac{1}{2}|a\rangle + \frac{1}{2}|b\rangle$ ,  $M_2(\sigma) = \frac{1}{2}|1\rangle + \frac{1}{2}|2\rangle$
  - And:  $M_1(\sigma) \otimes M_2(\sigma) = \frac{1}{4}|a, 1\rangle + \frac{1}{4}|a, 2\rangle + \frac{1}{4}|b, 1\rangle + \frac{1}{4}|b, 2\rangle$
- ▶ Question: how **different** are  $\sigma$  and  $M_1(\sigma) \otimes M_2(\sigma)$  ?
  - What is their **distance**  $d(\sigma, M_1(\sigma) \otimes M_2(\sigma))$  ?
  - Can it be maximalised?
- ▶ How does this compare to the **quantum** case, where the folklore opinion is that joint states can be “more entangled” than in classical probability?
- ▶ The paper contains several **experiments**
  - findings: the difference is not that big, certainly not in the limit



## Broader perspective

Entwinedness of joint states is related to both:

- (1) **disintegration** (see also REPAS talk)
  - extracting a channel  $X \rightarrow Y$  from a joint state on  $X, Y$
  - (or the other way around)
  - deep questions about correlation and causality
  - ongoing joint work with Kenta Cho
- (2) **cross-over influence**
  - conditioning in one coordinate changes the other coordinate
  - in the quantum world, this is Einstein’s: “spooky interaction”
  - ongoing joint work with Fabio Zanasi (see again MFCS'17)



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## Which distance to use?

- ▶ In the **discrete classical** case: **total variation distance**
  - $\mathbf{tvd}(\omega, \rho) = \frac{1}{2} \sum_x |\omega(x) - \rho(x)|$
  - special (discrete) case of **Kantorovic** metric, for distributions on metric spaces
- ▶ In the **quantum** case: **trace distance**
  - states are density matrices on Hilbert spaces, or more generally, points of von Neumann algebras
  - $\mathbf{trd}(\omega, \rho) = \frac{1}{2} \text{tr}(|\omega - \rho|) = \frac{1}{2} \text{tr}(\sqrt{(\omega - \rho)^\dagger(\omega - \rho)})$
  - total variation is a special case, for classical (diagonal) states

Both distances have values in the unit interval  $[0, 1]$

Both distances have been implemented in **EfProb**, an embedded language of Python for probabilistic computing (discrete, continuous, quantum)

- ▶ See [efprob.cs.ru.nl](http://efprob.cs.ru.nl) and the CALCO tools talk



## Experimental results I

- ▶ For the  $2 \times 2$  Bell state  $qs$ 
  - distance  $\mathbf{trd}(qs, M_1(qs) \otimes M_2(qs)) = \frac{3}{4}$
- ▶ For a classical  $2 \times 2$  analogue  $cs$ 
  - $cs = \frac{1}{2}|0, 0\rangle + \frac{1}{2}|1, 1\rangle$
  - distance  $\mathbf{tvd}(cs, M_1(cs) \otimes M_2(cs)) = \frac{1}{2}$

## Experimental results II: $n$ -ary generalisation

- ▶ For the  $2^n \times 2^n$   $n$ -ary Bell state  $qs_n$ 
  - distance  $\mathbf{trd}(qs_n, M_1(qs_n) \otimes \cdots \otimes M_n(qs_n)) = \frac{2^n - 1}{2^n}$
- ▶ For a classical  $2^n \times 2^n$  analogue  $cs_n$ 
  - $cs_n = \frac{1}{2}|0, \dots, 0\rangle + \frac{1}{2}|1, \dots, 1\rangle$
  - distance  $\mathbf{tvd}(cs_n, M_1(cs_n) \otimes \cdots \otimes M_n(cs_n)) = \frac{2^{n-1} - 1}{2^{n-1}}$

**Conclusion:** the classical case lags one step behind the quantum case

- ▶ Aside: the same happens if you use **mutual information**
- ▶  $qs_n \mapsto n$      $cs_n \mapsto n - 1$



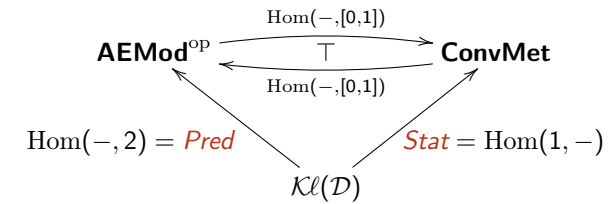
## Logical reformulation of distances

- ▶ In the **classical** case, for **Kantorovic** distance **kvd**
  - consider distributions  $\omega, \rho$  on metric space  $X$
  - $\text{kvd}(\omega, \rho) = \bigvee_{\substack{q: X \rightarrow [0,1] \\ \text{non-expansive}}} \left| \omega \models q - \rho \models q \right|$
  - (alternative formulation uses ‘couplings’)
- ▶ In **quantum** case, for trace distance **trd**
  - for states  $\omega, \rho$  on Hilbert space  $\mathcal{H}$
  - $\text{trd}(\omega, \rho) = \bigvee_{\substack{q: \mathcal{H} \rightarrow \mathcal{H} \\ 0 \leq q \leq 1}} \left| \omega \models q - \rho \models q \right|$
  - The same definition transfers to **von Neumann algebras**



## Metric state-and-effect triangle

The earlier triangle specialises to:

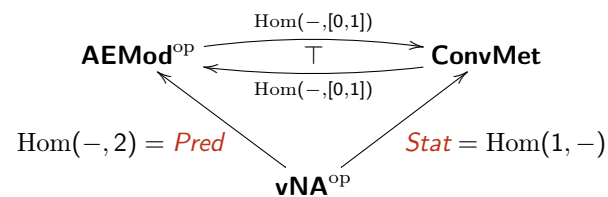


where:

- ▶ **AEMod** is the category of **Archimedean** effect modules
  - a metric can be defined on them; it's  $\vee$ -metric for predicates
  - so that all effect module maps are automatically non-expansive
- ▶ **ConvMet** is the category of convex metric spaces
  - algebra  $\mathcal{D}(X) \rightarrow X$  must be non-expansive
  - homomorphisms are both affine and non-expansive



## Quantum version



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## Final remarks

- ▶ Metrics have been used to compare classical and quantum probability, wrt. entwinedness
- ▶ Categorically, they fit in the same triangle pictures
- ▶ Open question: can we turn the adjunction  $\mathbf{AEMod}^{\text{op}} \rightleftarrows \mathbf{ConvMet}$  into a **Kadison** style duality, by additionally requiring completeness of metric spaces?

