

Categories of partial probabilistic computations

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Discrete probability distributions

Definition

A **discrete probability distribution** on a set X is a function $\varphi : X \rightarrow [0, 1]$ such that

- ❶ $\text{supp}(\varphi) = \{x \in X : \varphi(x) \neq 0\}$ is a finite set.
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We often write such distributions as formal sums:

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$$\frac{1}{3}|a, 1\rangle + \frac{1}{3}|b, 1\rangle + \frac{1}{3}|b, 3\rangle \quad \mapsto \text{renormalise}$$

$$\frac{1}{2}|b, 1\rangle + \frac{1}{2}|b, 3\rangle \quad = \omega|_{\mathbb{A}}(b)$$

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???

Discrete probability subdistributions

Definition

A **discrete probability subdistribution** on a set X is a function $\varphi : X \rightarrow [0, 1]$ such that

- 1 $\text{supp}(\varphi) = \{x \in X : \varphi(x) \neq 0\}$ is a finite set.
- 2 $\sum_{x \in X} \varphi(x) \leq 1$.

There is a monad $\mathcal{D}_{\leq} : \mathbf{Sets} \rightarrow \mathbf{Sets}$. Its Kleisli-category is symmetric monoidal.

Write $f : X \multimap Y$ for a Kleisli-map $f : X \rightarrow \mathcal{D}_{\leq}(Y)$.

The structure of $\mathcal{KI}(\mathcal{D}_{\leq})$

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- Comparator maps:

$$\nabla_X : X \times X \multimap X$$

$$\nabla_X(x, y) = \begin{cases} 1|x\rangle & \text{if } x = y \\ \mathbf{0} & \text{otherwise} \end{cases}$$

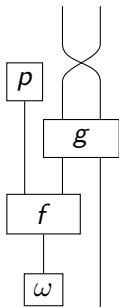
Copy-discard-compare (CDC) categories

Definition

A **copy-discard-compare category** is a symmetric monoidal category (C, \otimes, I) with **copier** $\Delta_X : X \rightarrow X \otimes X$, **discard** $d_X : X \rightarrow I$, and **comparator** $\nabla_X : X \otimes X \rightarrow X$ maps such that...

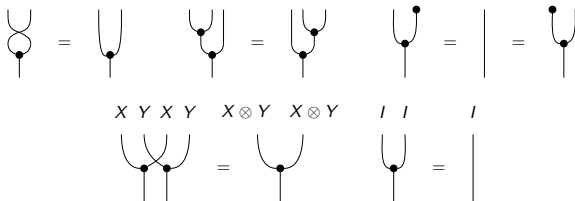
The anatomy of a string diagram

Let $(\mathbf{C}, \otimes, I, \alpha, \lambda, \rho, \sigma)$ be symmetric monoidal.



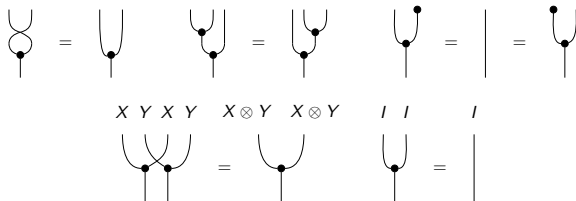
The axioms of copy-discard-compare (CDC) categories

- Copy and discard:

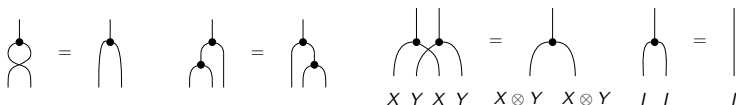


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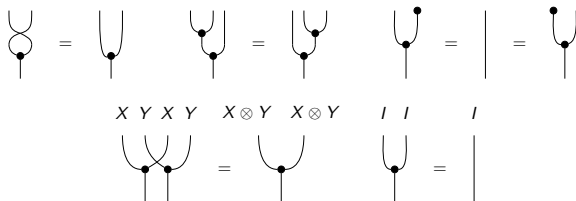


- Compare:

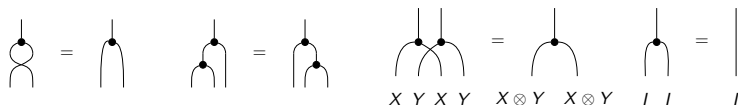


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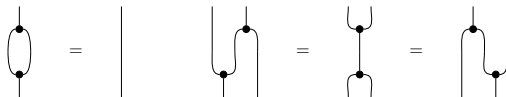
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- Compare:



- Copy-compare interaction:



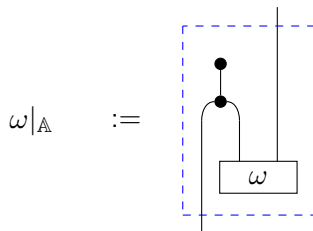
Examples of CDC-categories

- Sets and subprobability channels: $\mathcal{KL}(\mathcal{D}_{\leq})$
- Finite dimensional vector spaces and linear maps: **FinVect**
- Sets and relations: **Rel**
- Standard Borel spaces and subprobability kernels: **BorelStoch** _{\leq}
- ...

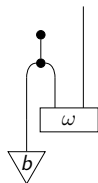
Our disintegration problem revisited

From $\omega : 1 \multimap \mathbb{A} \times \mathbb{N}$

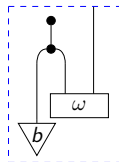
extract $\omega|_{\mathbb{A}} : \mathbb{A} \multimap \mathbb{N}$:



Normalisation

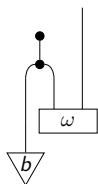


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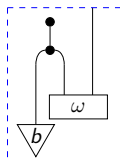


$$\frac{1}{3}|a, 1\rangle + \frac{1}{3}|b, 1\rangle + \frac{1}{3}|b, 3\rangle \mapsto \frac{1}{2}|b, 1\rangle + \frac{1}{2}|b, 3\rangle$$

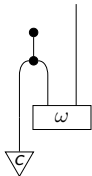
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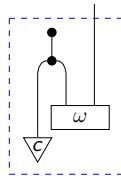
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$$\frac{1}{3}|a, 1\rangle + \frac{1}{3}|b, 1\rangle + \frac{1}{3}|b, 3\rangle \mapsto 0$$

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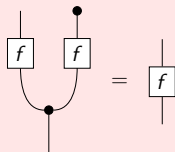
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- These axioms uniquely characterise normalisation boxes when they exist.
- The normalisation boxes select a least normalisation with respect to a partial order on normalisations.
- The dashed boxes enjoy many compositional properties.

Self-normalising maps

Definition

We call a map $f : X \rightarrow Y$ **self-normalising** if



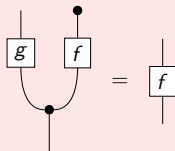
For a subchannel $f : X \multimap Y$ in $\mathcal{KL}(\mathcal{D}_{\leq})$ this translates to

$$\forall x \in X. \sum_{y \in Y} f(x)(y) \in \{0, 1\}$$

The 'normalised by' relation

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A map $g : X \rightarrow Y$ **normalises** $f : X \rightarrow Y$ if



In this case, we write $f \preceq g$.

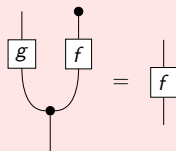
For subchannels $f, g : X \rightarrow Y$ in $\mathcal{Kl}(\mathcal{D}_{\leq})$ this translates to

$$f(x) \neq \mathbf{0} \implies \forall y \in Y. g(x)(y) = \frac{f(x)(y)}{\sum_{y'} f(x)(y')}$$

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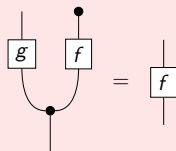
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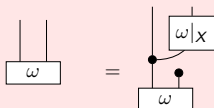
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‘How to factorise $P(X, Y) = P(X) \cdot P(Y|X)$?’.

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If $\omega : I \rightarrow X \otimes Y$, then a *disintegration* of ω is a map $\omega|_X : X \rightarrow Y$ that satisfies



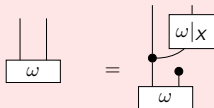
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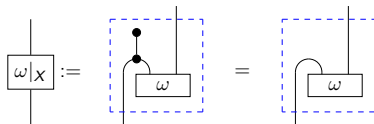
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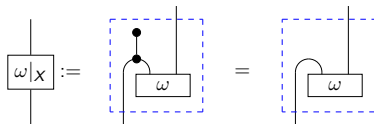
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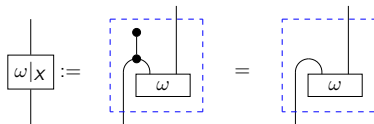
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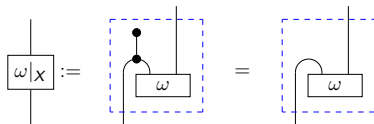


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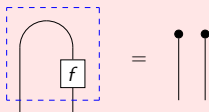
Remark

This does not recover disintegration in **BorelStoch**_≤

Outlook: notions of support

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A map $f : X \rightarrow Y$ has **everywhere full support** if

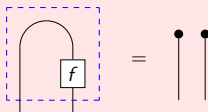


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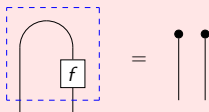


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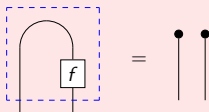


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- If f has everywhere full support, updating f never fails.
- Everywhere full support maps are closed under many operations: composition, tensor, marginalisation, normalisation, disintegration.
- We use this notion for a compositional graphical calculus of disintegrations in a future article with Bart Jacobs, and Dario Stein.

Outlook: connections to effectus theory

- Effectus theory is a category theoretic framework for quantum and probability theory, focusing on the structure of well-behaved coproducts.

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- Effectuses need not have a monoidal structure. Adding that creates a very rich setting.
- Effectuses come in a total and partial flavour:

$$\begin{array}{ccc} \text{TotEf} & \xrightleftharpoons[\mathcal{T}_{ot}]{\mathcal{K}I(-+1)} & \text{ParEf} \\ & \cong & \end{array}$$

References

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