

Bob Walters memorial

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TOWARDS MARKOV BICATEGORIES

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joint work with Mario Román

OVERVIEW

- Cartesian bicategories of relations are a syntax for relations
- Some facts hint at bicategories of stochastic relations
 - ↳ Minimal conditionals in partial Markov categories
 - ↳ convex sets of distributions monad

CARTESIAN BICATEGORIES OF RELATIONS

A cartesian bicategory is a locally posetal monoidal bicategory with a cocommutative comonoid and a commutative monoid structures satisfying

$$\begin{array}{c} \text{---} \\ \text{---} \end{array} \quad \begin{array}{c} \text{---} \\ \text{---} \end{array}$$
$$\begin{array}{c} \text{---} \\ \text{---} \end{array} \quad \begin{array}{c} \text{---} \\ \text{---} \end{array}$$

$$\text{---} \leq \text{---} \circ \text{---}$$

$$\text{---} \leq \text{---} \cdot \text{---}$$

(adjointness)

$$\begin{array}{c} \text{---} \\ \text{---} \end{array} \leq \text{---}$$

$$\text{---} \leq \text{---}$$

and

$$\begin{array}{c} \text{---} \otimes \text{---} \\ \text{---} \end{array} \leq \begin{array}{c} \text{---} \otimes \text{---} \\ \text{---} \end{array}$$

$$\begin{array}{c} \text{---} \otimes \text{---} \\ \text{---} \end{array} \leq \text{---}$$

(lax naturality)

$$\begin{array}{c} \text{---} \\ \text{---} \end{array} = \begin{array}{c} \text{---} \\ \text{---} \end{array}$$

(Grobenius)

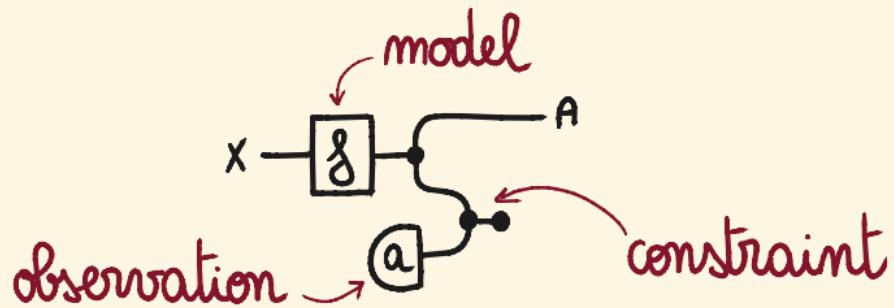
[Carboni & Walters, 1987]

OUTLINE

- Observations & updates
- Canonical conditionals and inequalities

PARTIALITY FOR OBSERVATIONS

Updating a model on an observation means restricting the model to scenarios that are compatible with this observation.



constraints \rightarrow cannot be total computations
because $\neq \equiv$.

OVERVIEW

combine Markov and cartesian restriction categories to express partial stochastic processes.

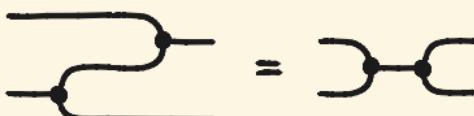
cartesian
restriction

Markov
with conditionals



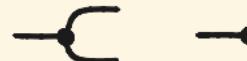
Add the discrete structure to express equality checking.

discrete cartesian
restriction



COPY-DISCARD CATEGORIES

A copy-discard category is a symmetric monoidal category where every object is a uniform cocommutative comonoid.



COCOMMUTATIVE COMONOID

$\text{---} \bullet \text{---} = \text{---} \bullet \text{---}$

$\text{---} \bullet \text{---} = \text{---}$

$\text{---} \bullet \text{---} = \text{---} \bullet \text{---}$

UNIFORMITY

$x \otimes y \text{---} \bullet \text{---} = \begin{matrix} x \\ y \end{matrix} \text{---} \bullet \text{---}$

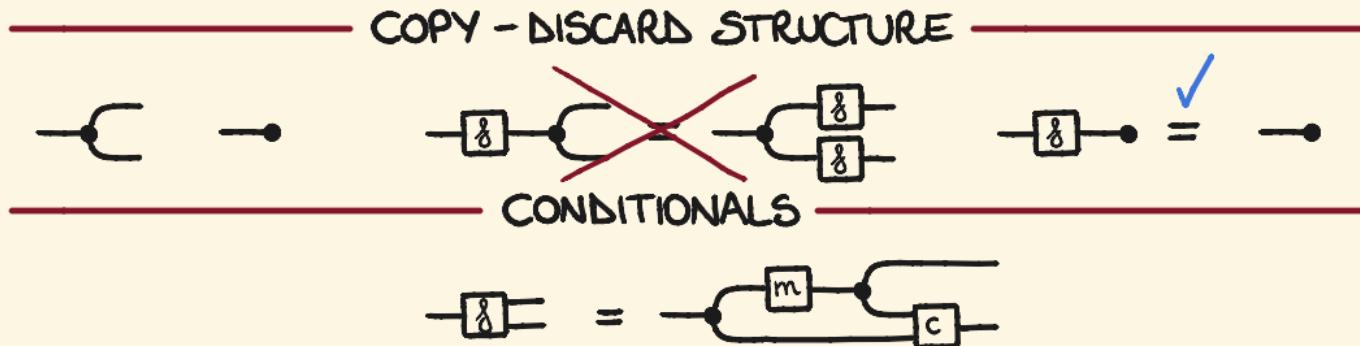
$x \otimes y \text{---} = \begin{matrix} x \\ y \end{matrix} \text{---}$

NO NATURALITY REQUIRED



MARKOV CATEGORIES & CONDITIONALS

A Markov category with conditionals is a copy-discard category with conditionals where all morphisms are total.



FINITARY DISTRIBUTIONS

A finitary distribution $\sigma \in \mathcal{D}(A)$ is a function
 $\sigma: A \rightarrow [0, 1]$ such that

- its support, $\text{supp}(\sigma) := \{a \in A \mid \sigma(a) > 0\}$, is finite, and
- its total probability mass is 1, $\sum_{a \in A} \sigma(a) = 1$.

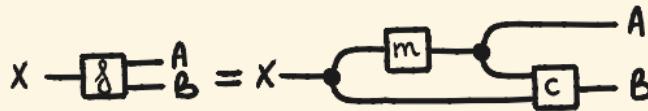
A morphism $x \dashv \vdash A$ in $\text{Kl}\mathcal{D}$ is a function $X \rightarrow \mathcal{D}(A)$
 $g(a|x)$ = "probability of a given x "

composition is

$$x \dashv \vdash g \dashv \vdash B \quad (b|x) := \sum_{a \in A} g(a|x) \cdot g(b|a)$$

CONDITIONALS

KlD has conditionals.



$$m(a|x) := \sum_{b \in B} \delta(a, b|x)$$

$$x - \boxed{m} - A := x - \boxed{\delta} - A$$

$$c(b|a,x) := \begin{cases} \frac{\delta(a,b|x)}{m(a|x)} & \text{if } m(a|x) \neq 0 \\ \sigma(b) & \text{if } m(a|x) = 0 \end{cases}$$

any distribution on B

~ conditionals are not unique and they cannot be

MARGINALS IN MARKOV CATEGORIES

Marginals in Markov categories are as expected :

$$x - \boxed{m} - A = x - \boxed{\delta} - \begin{matrix} A \\ B \end{matrix}$$

PROOF

$$\begin{aligned} & \text{---} \boxed{\delta} \text{ ---} \\ = & \text{---} \bullet \text{---} \boxed{m} \text{---} \bullet \text{---} \boxed{c} \text{---} \bullet \text{---} \\ = & \text{---} \bullet \text{---} \boxed{m} \text{---} \bullet \text{---} \end{aligned}$$

$$\begin{aligned} & \text{conditionals :} \\ \rightsquigarrow & \text{---} \boxed{\delta} \text{ ---} = \text{---} \bullet \text{---} \boxed{m} \text{---} \bullet \text{---} \boxed{c} \text{---} \bullet \text{---} \\ & \rightsquigarrow \text{totality} \end{aligned}$$

□

DROPPING TOTALITY

We want to keep the nice marginals of Markov categories.

$$x - \boxed{\delta} = x - \text{m} \quad x - \boxed{m} = x - \boxed{\delta}$$

Should we ask conditionals to be total ? X NO

→ too strong: total conditionals fail to exist in $\text{Kl}(\mathcal{D}_{\leq 1})$.

Can we ask conditionals to be quasi-total ? ✓ YES

→ sweet spot: quasi-total conditionals usually exist
and give nice marginals.

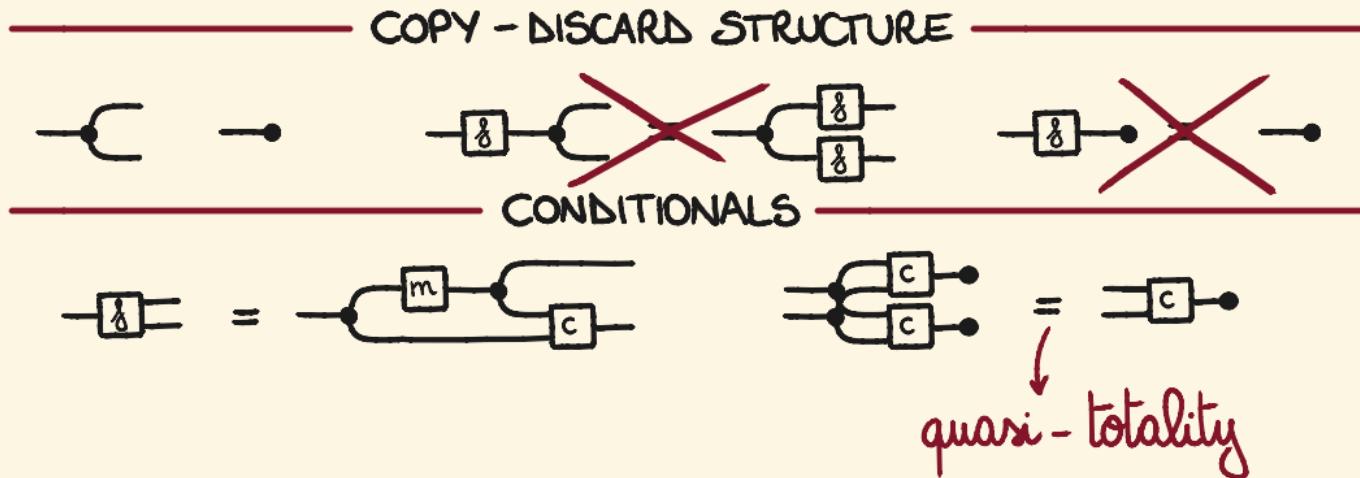
QUASI-TOTAL MORPHISM (in a copy-discard category)

$$\begin{array}{c} \text{---} \\ | \\ \boxed{\delta} \\ | \\ \boxed{\delta} \end{array} = \text{---} \quad \rightsquigarrow \text{failure is deterministic}$$

↑
domain of definition

PARTIAL MARKOV CATEGORIES

A partial Markov category is a copy-discard category with quasi-total conditionals.



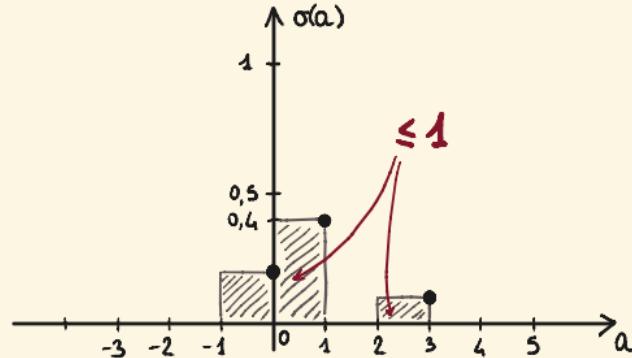
SUBDISTRIBUTIONS

A subdistribution σ on A is a distribution on $A+1$:

$\sigma \in \mathcal{D}_{\leq 1}(A)$ is a function $\sigma: A \rightarrow [0, 1]$ such that

- its support, $\text{supp}(\sigma) := \{a \in A \mid \sigma(a) > 0\}$, is finite, and
- its total probability mass is at most 1, $\sum_{a \in A} \sigma(a) \leq 1$.

ex $A = \mathbb{N}, \sigma =$



Subdistributions give a monad $\mathcal{D}_{\leq 1}: \text{Set} \rightarrow \text{Set}$
(there's a distributive law $\mathcal{D}(-)+1 \rightarrow \mathcal{D}(-+1)$)

SUBDISTRIBUTIONS

A morphism $x \dashv \delta \dashv A$ in $\text{Kl}\mathcal{D}_{\leq 1}$ is a function $X \rightarrow \mathcal{D}_{\leq 1}(A)$

$\delta(a|x) = \text{"probability of a given } x\text{"}$

$\delta(\perp|x) = \text{"probability of failure"}$

composition is:

$$x \dashv \delta \dashv g \dashv B \quad (b|x) := \sum_{a \in A} \delta(a|x) \cdot g(b|a)$$

$$x \dashv \delta \dashv g \dashv B \quad (\perp|x) := \delta(\perp|x) + \sum_{a \in A} \delta(a|x) \cdot g(\perp|a)$$

CONDITIONALS IN SUBDISTRIBUTIONS

A quasi-total morphism $g: X \rightarrow B$ is a function $g: X \rightarrow \mathcal{D}B + 1$.

$$x - \boxed{g} \xrightarrow{A} B = x - \begin{array}{c} m \\ \text{---} \\ c \end{array} \xrightarrow{A} B$$

The marginal of \mathfrak{f} is:

$$x - \boxed{m} \xrightarrow{A} (a|x) = x - \boxed{g} \xrightarrow{A} (a|x) = \sum_{b \in B} \mathfrak{f}(a, b|x)$$

$$x - \boxed{m} \xrightarrow{A} (\perp|x) = x - \boxed{g} \xrightarrow{A} (\perp|x) = \mathfrak{f}(\perp|x)$$

A conditional of \mathfrak{f} is:

$$\begin{array}{l} x - \boxed{c} \xrightarrow{A} B (b|a,x) = \begin{cases} \frac{\mathfrak{f}(a,b|x)}{m(a|x)} & m(a|x) \neq 0 \\ 0 & m(a|x) = 0 \end{cases} \end{array}$$

$$\begin{array}{l} x - \boxed{c} \xrightarrow{A} B (\perp|a,x) = \begin{cases} 0 & m(a|x) \neq 0 \\ 1 & m(a|x) = 0 \end{cases} \end{array}$$

EXAMPLES : PARTIAL STOCHASTIC PROCESSES

A partial stochastic process is a stochastic process
that may fail.

↳ Maybe monad

a Markov category with conditionals

Partial stochastic processes form a partial Markov category.

PROPOSITION

{ of Markov category with conditionals and coproducts
some ugly technical conditions
 $\Rightarrow \text{Kl}(\cdot + 1)$ is a partial Markov category.

EXAMPLES

- $\text{Kl}(\mathcal{D}(\cdot + 1))$ \rightsquigarrow finitary subdistributions
- $\text{Kl}(\mathcal{C}\mathcal{I}\mathcal{R}\mathcal{Y}_S(\cdot + 1))$ \rightsquigarrow subdistributions on standard Borel spaces

BAYES INVERSION & NORMALISATION

BAYES INVERSION

$$\text{original DAG: } \sigma \rightarrow g \rightarrow A = \text{transformed DAG: } \sigma \rightarrow \text{marginal} \rightarrow g^+ \rightarrow A$$
$$g_\sigma^+(b|a) := \frac{g(a|b)\sigma(b)}{\sum_{b' \in B} g(a|b')\sigma(b')}$$

g_σ^+ is a Bayes inversion of g w.r.t. σ

NORMALISATION

$$\text{original DAG: } x \rightarrow f \rightarrow A = \text{transformed DAG: } x \rightarrow \text{marginal} \rightarrow f^- \rightarrow A$$
$$\bar{f}(a|x) := \frac{f(x|a)}{1 - f(\perp|a)}$$

\bar{f} is a normalisation of f

→ Both are particular cases of quasi-total conditionals.

EQUALITY CHECK

$\text{KlD}_{\leq 1}$ has equality checks.

$$\overset{A}{\underset{A}{\textstyle \bigtriangleright}} (a, a') := \begin{cases} \delta_a & \text{if } a = a' \\ \delta_{\perp} & \text{if } a \neq a' \end{cases}$$

Equality checks interact with the comonoid structure.

$$A - \text{---} \circ \text{---} A = A - \text{---} \quad \text{and}$$

$$A - \overset{A}{\underset{A}{\textstyle \bigtriangleright}} \text{---} \underset{A}{\textstyle \bigtriangleright} A = \begin{array}{c} A \\ \text{---} \\ A - \circ \text{---} A \end{array}$$

DISCRETE PARTIAL MARKOV CATEGORIES

A discrete partial Markov category is a copy-discard category with quasi-total conditionals and comparators.

COPY - DISCARD STRUCTURE

$$\text{---} \bullet \text{---} = \text{---} \boxed{\delta} \text{---} \bullet \text{---} \neq \text{---} \bullet \text{---} \boxed{\delta} \text{---} \bullet \text{---} \neq \text{---} \bullet \text{---}$$

CONDITIONALS

$$\boxed{\delta} \text{---} = \text{---} \bullet \text{---} \boxed{m} \text{---} \bullet \text{---} \boxed{c} \text{---} \quad \text{---} \bullet \text{---} \boxed{c} \text{---} \bullet \text{---} = \text{---} \bullet \text{---} \boxed{c} \text{---} \bullet \text{---}$$

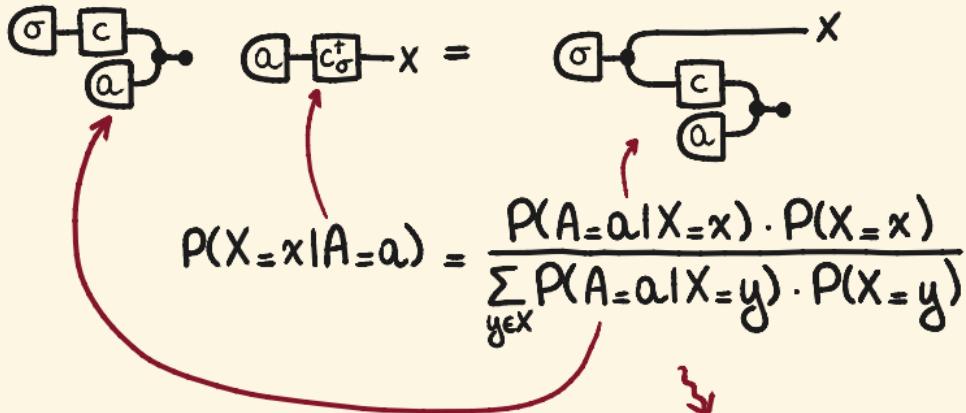
PARTIAL FROBENIUS STRUCTURE

$$\text{---} \bullet \text{---} = \text{---} \quad \text{---} \bullet \text{---} = \text{---} \quad \text{---} \bullet \text{---} = \text{---} \quad \text{---} \bullet \text{---} = \text{---}$$

↑ COMPARATOR

SYNTHETIC BAYES THEOREM

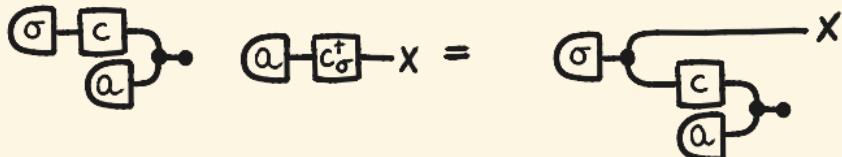
A deterministic observation $a: I \rightarrow A$ from a prior $\sigma: I \rightarrow X$ through a channel $c: X \rightarrow A$ determines an update proportional to the Bayes inversion c_σ^+ evaluated on a .



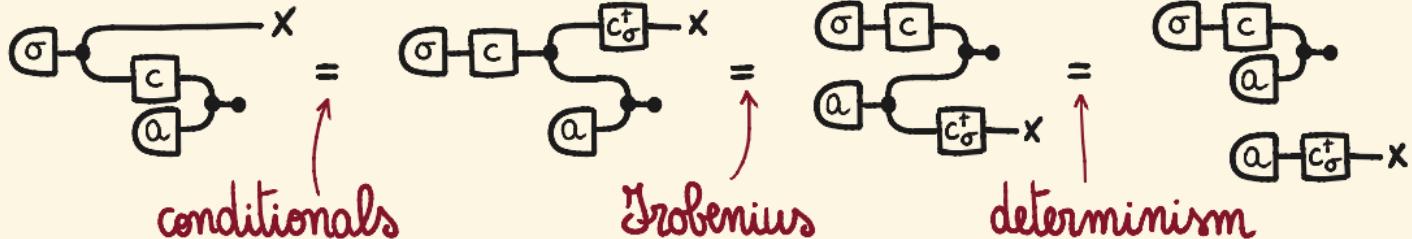
classical formula
for Bayes theorem

SYNTHETIC BAYES THEOREM

A deterministic observation $a: I \rightarrow A$ from a prior $\sigma: I \rightarrow X$ through a channel $c: X \rightarrow A$ determines an update proportional to the Bayes inversion c_σ^+ evaluated on a .



PROOF

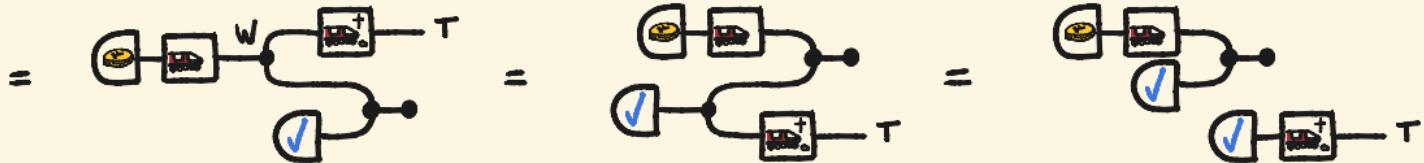


□

EXAMPLE : THE SLEEPING BEAUTY

$\omega = \begin{cases} H & \Rightarrow \text{to Rome} \rightarrow \text{wake up once} \\ T & \Rightarrow \text{to Naples} \rightarrow \text{wake up 3 times} \end{cases}$

If Sleeping Beauty wakes up, how should she update her belief?



OUTLINE

- Observations & updates
- Canonical conditionals and inequalities

MINIMAL CONDITIONALS

• partial Markov category

RESTRICTION ORDER

$$\boxed{\delta} \leq \boxed{g} \Leftrightarrow \boxed{\delta} = \begin{array}{c} g \\ \downarrow \\ \delta \end{array}$$

→ this is a partial order on quasi-total morphisms

$$\boxed{\delta} \leq \boxed{\delta} \Leftrightarrow \boxed{\delta} = \begin{array}{c} \delta \\ \downarrow \\ \delta \end{array}$$

(reflexivity)

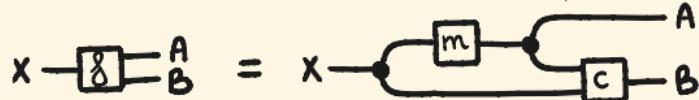
(quasi-totality)

MINIMAL CONDITIONALS

• has minimal conditionals if, for every $\boxed{\delta}$,
the partial order on its quasi-total conditionals has
a minimal element.

MINIMAL CONDITIONALS IN SUBDISTRIBUTIONS

A quasi-total morphism $g: X \rightarrow B$ is a function $g: X \rightarrow \mathcal{D}B + 1$.



The conditionals of g are:

$$x \begin{smallmatrix} A \\ \square \\ B \end{smallmatrix} (b|a, x) = \begin{cases} \frac{g(a, b|x)}{m(a|x)} & m(a|x) \neq 0 \\ \sigma(b) & m(a|x) = 0 \end{cases}$$

$$x \begin{smallmatrix} A \\ \square \\ B \end{smallmatrix} (\perp|a, x) = \begin{cases} 0 & m(a|x) \neq 0 \\ \sigma(\perp) & m(a|x) = 0 \end{cases}$$

for some $\sigma \in \mathcal{D}(B) + 1$.

\Rightarrow The minimal choice for σ is $\begin{cases} \sigma(b) = 0 & \text{for } b \in B \\ \sigma(\perp) = 1 & \end{cases}$ for $b \in B$.

\rightsquigarrow fail on unexpected observations

CONVEX SETS OF DISTRIBUTIONS

There's a monad $\mathcal{C} : \text{Set} \rightarrow \text{Set}$:

$$\mathcal{C}(X) := \{ A \subseteq \mathcal{D}(X) \mid A \neq \emptyset, A \text{ convex} \}$$

composition in $\text{Kl}(\mathcal{C})$ is

$$\begin{aligned} A - \boxed{\delta}^B - g - C (a) &:= \bigcup_{\sigma \in g(a)} \sum_{b \in \text{supp } \sigma} \sigma(b) \cdot g(b) \\ &= \bigcup_{\sigma \in g(a)} \left\{ \sum_{b \in \text{supp } \sigma} \sigma(b) \tau_b \mid \tau_b \in g(b) \right\} \end{aligned}$$

There's a partial order on morphisms

$$x - \boxed{\delta} - A \leq x - \boxed{g} - A \iff \forall x \in X \quad f(x) \subseteq g(x)$$

CONVEX SETS OF DISTRIBUTIONS

COPY, DISCARD, UNIFORM

$$A \xrightarrow{ } \bullet_A^A \quad (a) := \{(\delta_a, \delta_a)\}$$

$$A \xrightarrow{ } \bullet \quad (a) := \{\delta_a\}$$

$$\bullet_A \xrightarrow{ } (*) := \mathcal{D}(A)$$

$$\begin{array}{c} \text{---} \leq \text{---} \bullet \\ \bullet \text{---} \leq \end{array}$$

LAX CONDITIONALS



CONVEX SETS OF SUBDISTRIBUTIONS

There's a monad $\mathcal{L}: \text{Set} \rightarrow \text{Set}$:

$$\mathcal{L}_{\leq 1}(X) := \{A \subseteq \mathcal{D}(X+1) \mid A \neq \emptyset, A \text{ convex}\}$$

composition in $\text{Kl}(\mathcal{L}_{\leq 1})$ is

$$\begin{aligned} A - \boxed{\delta}^B \boxed{g} - C (a) &:= \bigcup_{\sigma \in g(a)} \sum_{b \in \text{supp } \sigma} \sigma(b) \cdot g(b) \\ &= \bigcup_{\sigma \in g(a)} \left\{ \sum_{b \in \text{supp } \sigma} \sigma(b) \tau_b \mid \tau_b \in g(b) \right\} \end{aligned}$$

There's a partial order on morphisms

$$x - \boxed{\delta} - A \leq x - \boxed{g} - A \iff \forall x \in X \quad f(x) \subseteq g(x)$$

CONVEX SETS OF SUBDISTRIBUTIONS

COPY, DISCARD, UNIFORM

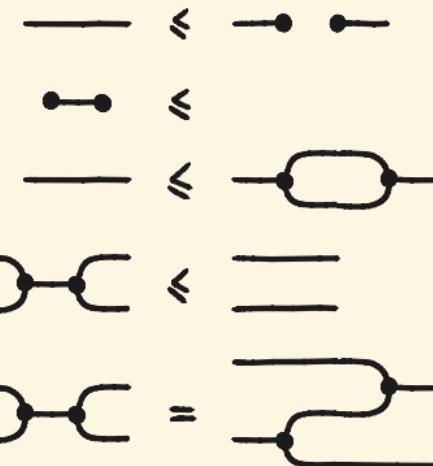
$$A \xrightarrow{=} A \quad (a) := \{\delta_{(a,a)}\}$$

$$A \xrightarrow{=} \bullet \quad (a) := \{\delta_{\perp}\}$$

$$\bullet \xrightarrow{=} A \quad (*) := \mathcal{D}(A)$$

$$A \xrightarrow{=} A \quad (a, a') := \begin{cases} \{\delta_a\} & a = a' \\ \{\delta_{\perp}\} & a \neq a' \end{cases}$$

LAX CONDITIONALS



MARKOV BICATEGORIES (TENTATIVE)

A Markov bicategory is a locally posetal monoidal bicategory with a cocommutative comonoid and a commutative monoid structures satisfying

$$\begin{array}{c} \text{---} \\ \text{---} \end{array} \quad \begin{array}{c} \text{---} \\ \text{---} \end{array}$$
$$\begin{array}{c} \text{---} \\ \text{---} \end{array} \quad \begin{array}{c} \text{---} \\ \text{---} \end{array}$$

$$\text{---} \leq \text{---} \circ \text{---}$$

$$\text{---} \leq \text{---} \cdot \text{---}$$

(adjointness)

$$\begin{array}{c} \text{---} \\ \text{---} \end{array} \leq \text{---}$$

$$\text{---} \leq \text{---}$$

and

$$\begin{array}{c} \text{---} \\ \text{---} \end{array} \leq \text{---} \circ \begin{array}{c} \text{---} \\ \text{---} \end{array} \circ \begin{array}{c} \text{---} \\ \text{---} \end{array}$$

$$\begin{array}{c} \text{---} \\ \text{---} \end{array} \leq \begin{array}{c} \text{---} \\ \text{---} \end{array} \circ \begin{array}{c} \text{---} \\ \text{---} \end{array}$$

$$\begin{array}{c} \text{---} \\ \text{---} \end{array} \leq \text{---}$$

$$\begin{array}{c} \text{---} \\ \text{---} \end{array} = \begin{array}{c} \text{---} \\ \text{---} \end{array}$$

CONCLUSIONS & FUTURE WORK

- Partial Markov categories express updates
- Canonical conditionals via the restriction order
- The Kleisli of convex sets of (sub)distributions has a partial order

⇒ What is the 'right' structure to capture $\text{Kl}(\ell)$ and $\text{Kl}(\ell_{\leq 1})$?

⇒ Would this give some interesting promonoidal structure?

$$S(X, A \diamond B) = \ell(X, A) \times \ell(A \otimes X, B)$$

