

Higher-order building blocks for statistical modelling

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Theorem (Aumann)

$$S = \mathbb{2}, \mathbb{N}, \mathbb{R}, \dots$$

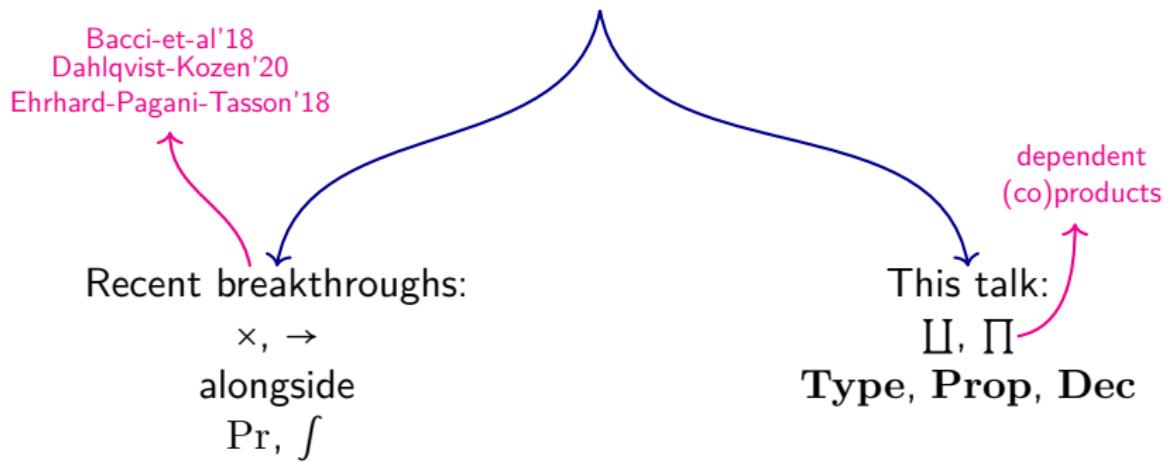
No σ -algebra on $\text{Meas}(\mathbb{R}, S)$ makes evaluation measurable:

$$\text{eval} : \text{Meas}(\mathbb{R}, S) \times \mathbb{R} \rightarrow S$$

⇒ bad fit for higher-order programming semantics:

Meas is not Cartesian closed

higher-order measure theory



higher-order measure theory with **quasi-Borel spaces**

Bacci-et-al'18
Dahlqvist-Kozen'20
Ehrhard-Pagani-Tasson'18

Recent breakthroughs:

\times, \rightarrow

alongside

\Pr, \int

dependent
(co)products

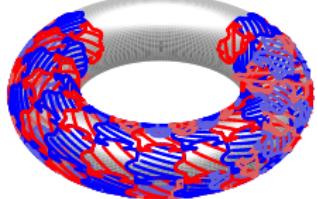
This talk:

\coprod, \prod

Type, Prop, Dec

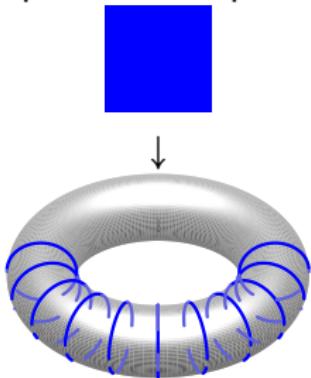
Intuition

measurable space



axiomatise
measurable events

quasi Borel space

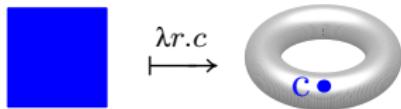


axiomatise
random elements

quasi-Borel space $X = (\lfloor X \rfloor, \mathcal{R}_X)$

random element $\alpha \in \mathcal{R}_X \subseteq \lfloor X \rfloor^{\mathbb{R}}$ axioms:

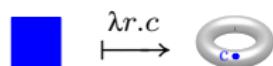
determinism. elements are random elements:



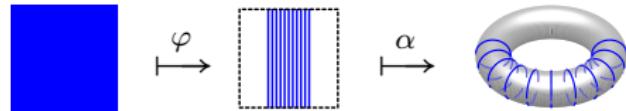
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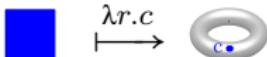
precomposition. ($\varphi \in \text{Meas}(\mathbb{R}, \mathbb{R})$)



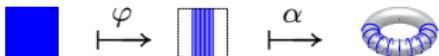
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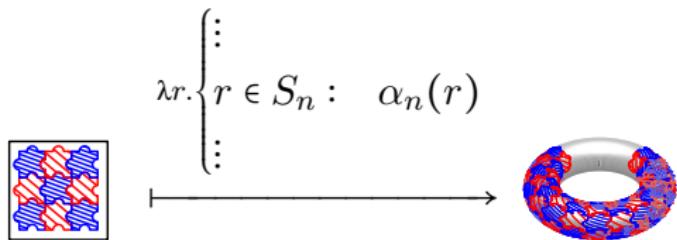
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recombination ($\mathbb{R} = \bigcup_{i=0}^{\infty} S_n$)



quasi-Borel space $X = (\lfloor X \rfloor, \mathcal{R}_X)$

random element $\alpha \in \mathcal{R}_X \subseteq \lfloor X \rfloor^{\mathbb{R}}$ axioms:

determinism.

$$\text{blue square} \xrightarrow{\lambda r.c} \text{grey torus with blue dot}$$

precomposition. ($\varphi \in \mathbf{Meas}(\mathbb{R}, \mathbb{R})$)

$$\text{blue square} \xrightarrow{\varphi} \text{vertical stripes} \xrightarrow{\alpha} \text{grey torus with blue spiral}$$

recombination ($\mathbb{R} = \bigcup_{i=0}^{\infty} S_n$)

$$\text{red and blue squares} \mapsto \text{grey torus with red and blue dots}$$

Morphisms $f : X \rightarrow Y$

functions $\lfloor f \rfloor : \lfloor X \rfloor \rightarrow \lfloor Y \rfloor$:

$$\alpha \in \mathcal{R}_X \implies f \circ \alpha \in \mathcal{R}_Y$$

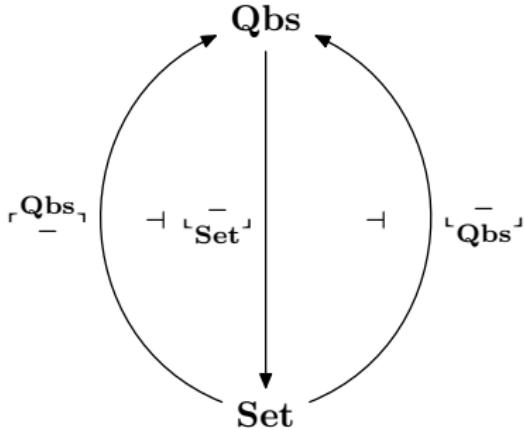
Free and cofree qbs

set A :

- free: $\mathcal{R}_{\langle \text{Qbs} \rangle_A} = \sigma\text{-simple functions:}$

$$\lambda r. \begin{cases} \vdots \\ r \in S_n : a_n \\ \vdots \end{cases}$$

- cofree: $\mathcal{R}_{\langle A \rangle_{\text{Qbs}}} = \text{all functions}$



qbses for:

$\mathbb{Z}, \mathbb{N}, \mathbb{Q}$

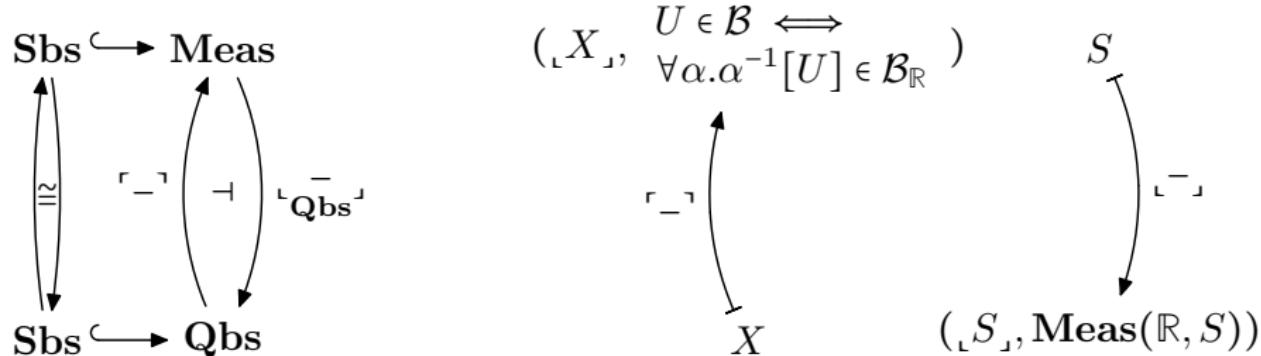
set-theoretic escape hatch

$\langle - \rangle_{\text{Set}} : \text{Qbs} \rightarrow \text{Set}$ generates limits and colimits

preserves
lifts
reflects

Using measure theory

Measurable space are carried by qbses:



Recover qbses for:

$$\mathbb{R}, \mathbb{W} := [0, \infty], \mathbb{I} := [0, 1]$$

Conservative extension for standard Borel spaces
Measure-theoretic escape hatch

Simple types

Simple products

Correlated random elements:

$$\mathcal{R}_{X \times Y} \xleftarrow[\cong]{(-,-)} \mathcal{R}_X \times \mathcal{R}_Y$$

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Simple coproducts

Recombinations:

$$\alpha \in \mathcal{R}_{\coprod_{i \in \mathcal{I}} X_i} \iff \alpha = \lambda r. \begin{cases} \vdots \\ r \in S_n : (i_n, \alpha_n r) \\ \vdots \end{cases} \quad (\mathbb{R} = \bigcup_{n=0}^{\infty}, \alpha_n \in X_{i_n})$$

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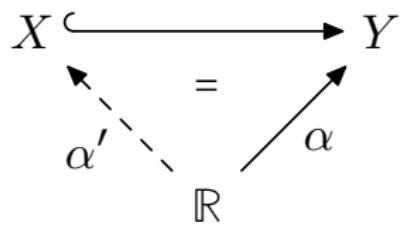
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Simple function spaces

$$[Y^X] = \mathbf{Qbs}(X, Y) \quad \mathcal{R}_{Y^X} \xrightarrow[\cong]{\text{uncurry}} \mathbf{Qbs}(\mathbb{R} \times X, Y)$$

Random element space: $\mathcal{R}_X := X^{\mathbb{R}}$

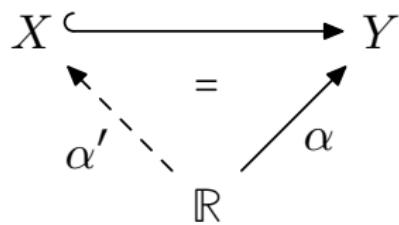
Subspaces



m injective and $\mathcal{R}_X = (m \circ)^{-1}[\mathcal{R}_Y]$

$\Omega := \mathcal{P}_{\text{Qbs}}$ subspace classifier

Subspaces



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Example

- ▶ Use $\llbracket \text{Prop} \rrbracket := \Omega$ for reasoning/axiomatics [Sato et al.'19].

Subspaces

$$\begin{array}{ccc} X & \xrightarrow{\quad} & Y \\ \alpha \swarrow & = & \searrow \alpha \\ \mathbb{R} & & \end{array}$$

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Example

- ▶ Use $\llbracket \text{Prop} \rrbracket := \Omega$ for reasoning/axiomatics [Sato et al.'19].
- ▶ Differentiation:

$$D_1 \mathbb{R} := \{f : \mathbb{R} \rightarrow \mathbb{R} \mid f \text{ differentiable everywhere}\} \hookrightarrow \mathbb{R}^{\mathbb{R}}$$

$$\frac{d}{d} : D_1 \rightarrow \mathbb{R}^{\mathbb{R}}$$

Borel subspaces

$m : X \hookrightarrow Y$ when $m : X \hookrightarrow Y$ and $[- \in X] \in \Omega^Y$
 factors through $\mathcal{Z}^Y \rightarrow \Omega^Y$

Borel subspaces

$$m : X \leftrightarrow Y \quad \text{when} \quad m : X \hookrightarrow Y \text{ and } [- \in X] \in \Omega^Y \\ \text{factors through } \mathcal{Z}^Y \rightarrow \Omega^Y$$

Example

- ▶ higher-order Qbs-internal σ -algebra:

$$-^{\mathbb{C}} : \mathcal{B}_X \rightarrow \mathcal{B}_X \quad \bigcap_{n=0}^{\infty} : \mathcal{B}_X^{\mathbb{N}} \rightarrow \mathcal{B}_X \quad -^{-1}[-] : Y^X \times \mathcal{B}_Y \rightarrow \mathcal{B}_X$$

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Example

- ▶ higher-order Qbs-internal σ -algebra:

$$-^{\complement} : \mathcal{B}_X \rightarrow \mathcal{B}_X \quad \bigcap_{n=0}^{\infty} : \mathcal{B}_X^{\mathbb{N}} \rightarrow \mathcal{B}_X \quad -^{-1}[-] : Y^X \times \mathcal{B}_Y \rightarrow \mathcal{B}_X$$

- #### ► Non-Qbs-morphisms:

$$\exists : \mathcal{B}_{X \times Y} \rightarrow \mathcal{B}_X \quad \quad [- = \emptyset] : \mathcal{B}_X \rightarrow \mathcal{P} \quad \quad [- \subseteq -] : \mathcal{B}_X^2 \rightarrow \mathcal{P}$$

Borel subspaces

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Example

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- ▶ Non-**Qbs**-morphisms:

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- ▶ $\mathcal{B}_{\mathcal{B}_X}$: Borel-on-Borel sets [Sabok-Staton-Stein-Wolman'21]

Measures

- Unrestricted Giry:

$$\langle G_1 X \rangle := \left\{ \mu : \langle \mathcal{B}_X^{\text{Qbs}} \rangle \rightarrow \mathbb{W} \middle| \text{measure on } \langle X \rangle^{\text{Meas}} \right\}$$

$$\mathcal{R}_{G_1 X} := \left\{ k : \mathbb{R} \times \langle \mathcal{B}_X^{\text{Qbs}} \rangle \rightarrow \mathbb{W} \middle| k \text{ is a kernel} \right\}$$

Analogous to the measure-theoretic Giry

But: careful to evaluate only σ -simple random Borel sets

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Analogous to the measure-theoretic Giry

But: careful to evaluate only σ -simple random Borel sets

- ▶ Following can evaluate any random Borel set:

$$\langle G_2 X \rangle := \left\{ \mu : \mathcal{B}_X \rightarrow \mathbb{W} \middle| \text{measure on } \langle X \rangle^{\text{Meas}} \right\}$$

$$\mathcal{R}_{G_2 X} := \{ k : \mathbb{R} \times \mathcal{B}_X \rightarrow \mathbb{W} | k \text{ is a kernel} \}$$

$\llbracket \mathbf{M} X \rrbracket := \{v_* \lambda_\Omega | v : \Omega \rightarrow X, \Omega \hookrightarrow \mathbb{R} \text{ a } \sigma\text{-finite standard measure space}\}$

$\mathcal{R}_{\mathbf{M} X} := \{v_* \circ k | k : \mathbb{R} \times \Omega \rightarrow X, \Omega \hookrightarrow \mathbb{R} \dots \text{ ditto } \dots\}$

$\llbracket MX \rrbracket := \{v_* \lambda_\Omega | v : \Omega \rightarrow X, \Omega \hookrightarrow \mathbb{R} \text{ a } \sigma\text{-finite standard measure space}\}$

$\mathcal{R}_{MX} := \{v_* \circ k | k : \mathbb{R} \times \Omega \rightarrow X, \Omega \hookrightarrow \mathbb{R} \dots \text{ ditto } \dots\}$

- ▶ For standard Borel spaces S, T :
 - ▶ MS are the s-finite measures
 - ▶ $(MS)^T$ are the s-finite kernels

Fully-definable semantic domain for first-order Monte Carlo models [Staton'17]

$\text{MX} := \{v_* \lambda_\Omega | v : \Omega \rightarrow X, \Omega \hookrightarrow \mathbb{R} \text{ a } \sigma\text{-finite standard measure space}\}$

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Fully-definable semantic domain for first-order Monte Carlo models [Staton'17]

- ▶ Integration is commutative
- ▶ Models synthetic measure theory [Kock'12]
- ▶ Probabilistic fragment $\mathbf{P}X := \{\mu | \mu X = 1\} \leftrightarrow \text{MX}$ with de Finetti's theorem [Heunen et al'17]
- ▶ $\mathbf{P}X$ also models name generation [Sabok et al.'21]

Syntactic spaces and recursive domains

Qbs is locally presentable

⇒ initial algebra semantics for inductive types

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Example

- ▶ Syntactic spaces for operational semantics
- ▶ Meta-programming data structures for Monte Carlo inference
[Ścibior et al.'18, Lew et al.'20]
- ▶ Extends to **recursive** types with domain theory
[Vákár-Kammar-Staton'19]
- ▶ Opportunity: abstract syntax with binding
[Fiore-Plotkin-Turi'99]

Monadic operational semantics

[Dal Lago et al.'17, Gavazzo'19, Vákár et al.'19]

$$\frac{k_1(t) w_1 \quad k_2(t, w_1) w_2 \quad \dots \quad k_n(t, w_1, \dots, w_n) v}{l(t) f(t, w_1, \dots, w_n, v)}$$

means

$$\begin{aligned} l(t) &:= k_1(t) && \rangle\!\! = \lambda w_1. \\ &\qquad\qquad\qquad k_2(t, w_1) && \rangle\!\! = \lambda w_2. \dots \\ &\qquad\qquad\qquad k_n(t, w_1, \dots, w_{n-1}) && \rangle\!\! = \lambda v. \\ &\delta_{f(t, w_1, \dots, w_n, v)} \end{aligned}$$

and $l(t) := k_1(t)$ $l(t) := k_2(t)$ means $l(t) := k_1(t) + k_2(t)$

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Example

$$\frac{t \Downarrow_n \underline{0} \quad s_1 \Downarrow_n v}{\mathbf{match } t \mathbf{ with } \{\underline{0} \rightarrow s_1 \mid \underline{_} \rightarrow s_2\} \Downarrow_n v}$$

$$\frac{t \Downarrow_n \underline{r} \quad s_2 \Downarrow_n v}{\mathbf{match } t \mathbf{ with } \{\underline{0} \rightarrow s_1 \mid \underline{_} \rightarrow s_2\} \Downarrow_n v} (r \neq 0)$$



Refinement types:

$$\frac{B : X \rightarrow \Omega^Y}{\prod_{x:X} Bx, \coprod_{x:X} Bx} \quad \prod_{x:X} Bx := \{f : X \rightarrow Y \mid \forall x \in X. fx \in Bx\} \hookrightarrow Y^X$$
$$\coprod_{x:X} Bx := \{(x, y) \in X \times Y \mid y \in Bx\} \hookrightarrow X \times Y$$

Example

Lebesgue spaces and modes of convergence:

$$\mathcal{L}_-^- \in \prod_{\substack{\lambda \in \mathbf{P}\Omega \\ p \in \mathbb{R}}} \left\{ f : \Omega \rightarrow [-\infty, \infty] \middle| \int d\omega |f \omega|^p < \infty \right\}$$

- ▶ Observation: **Sbs** closed under dependent pairs
- ▶ To get dependent types, want “good” universe **Type**

Monte Carlo inference

$$\text{model} = \rho \odot \lambda := \lambda \varphi. \int \lambda(d\omega) \rho(\omega) \cdot \varphi(\omega)$$

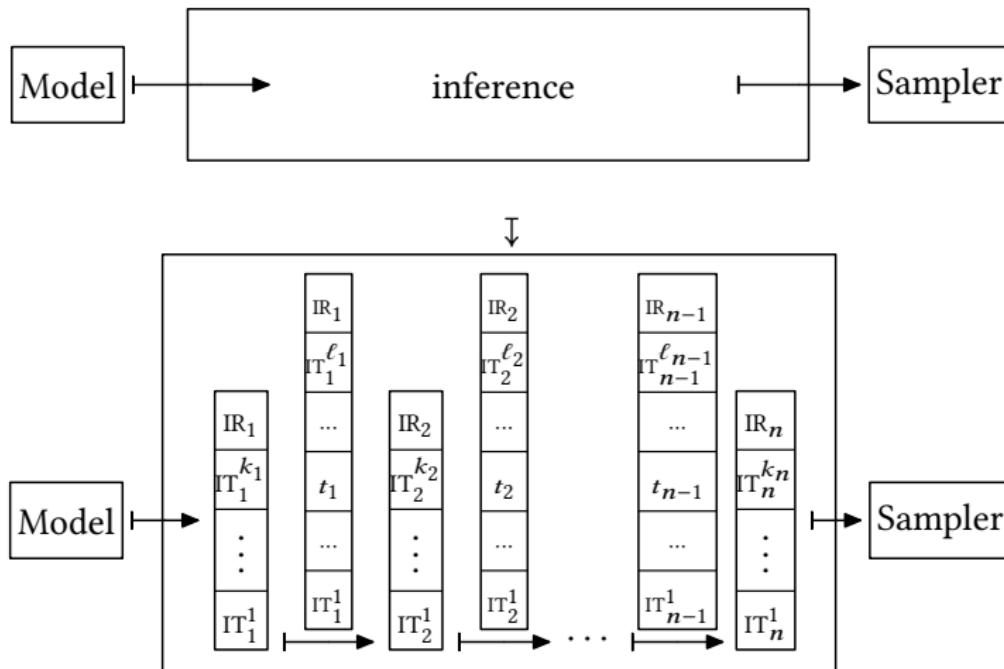
↑ likelihood ↑ prior

Defined programmatically:

sample : $\mathbf{M}\mathbb{R}$ #Uniform distribution on $[0, 1]$

score : $\mathbb{R} \rightarrow \mathbf{M}\mathbb{1}$

Modular inference [Ścibior et al.'18a+b, Lew et al.'20]



cf. inference with handlers [Bingham et al.'19] and ongoing



Thank you!

Exact conditioning

