

A domain theory for statistical probabilistic programming

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Statistical probabilistic programming

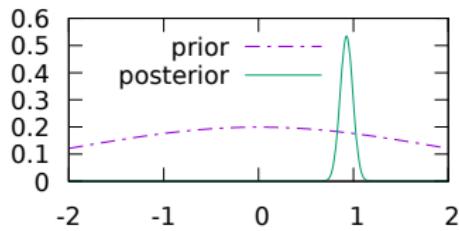
$\llbracket - \rrbracket$: programs \rightarrow unnormalised distributions

- ▶ Bayesian inference: compiler computes normalisation
- ▶ Continuous types: $\mathbb{R}, [0, \infty]$
- ▶ Probabilistic effects:

normally distributed sample

$\text{sample}(\mu, \sigma) : \mathbb{R}$

$\llbracket \text{sample}(0, 2) \rrbracket$



scale distribution by r

$$r : [0, \infty]$$

$\text{score}(r) : 1$

prior

let $x = \text{sample}(0, 2)$

in $\text{score}(\text{normalPdf}(1.1 | x, \frac{1}{4}))$;
 $\text{score}(\text{normalPdf}(1.9 | 2x, \frac{1}{4}))$;
 $\text{score}(\text{normalPdf}(2.7 | 3x, \frac{1}{4}))$;

x

posterior

conditioning/fitting to observed data with likelihood

Statistical probabilistic programming

Exact Bayesian inference
using disintegration
[Shan-Ramsey'17]

- ▶ Commutativity/exchangability/Fubini

$$\left[\begin{array}{l} \text{let } x = K \text{ in} \\ \text{let } y = L \text{ in} \\ f(x, y) \end{array} \right] = \left[\begin{array}{l} \text{let } y = L \text{ in} \\ \text{let } x = K \text{ in} \\ f(x, y) \end{array} \right]$$

$$\int \llbracket K \rrbracket (dx) \int \llbracket L \rrbracket (dy) f(x, y) = \int \llbracket L \rrbracket (dy) \int \llbracket K \rrbracket (dx) f(x, y)$$

probability
distributions

σ -finite
distributions

arbitrary
distributions

✓
not closed under
push-forward

✓
s-finite
distributions

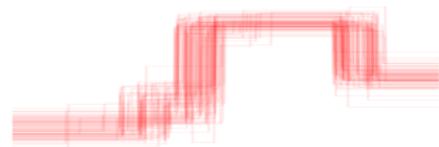
X
full definability
[Staton'17]

Statistical probabilistic programming

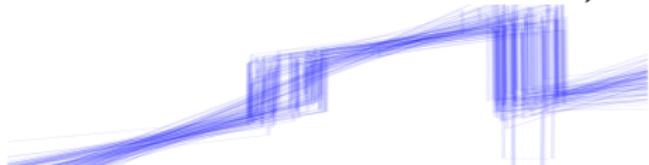
Express continuous distributions using:

- ▶ Higher-order functions:

(e.g. generative random function models)



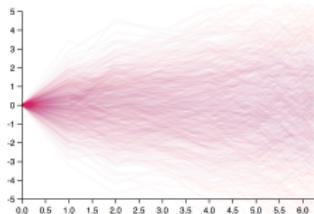
piecewise(random-constant)



piecewise(random-linear)

(e.g. Gaussian random walk)

▶ Term recursion:
 $rw(x, \sigma) = \lambda(). \quad // \text{thunk}$
let $y = \text{sample}(x, \sigma)$
in $(x, rw(y, \sigma))$



- ▶ Type recursion (à la FPC)

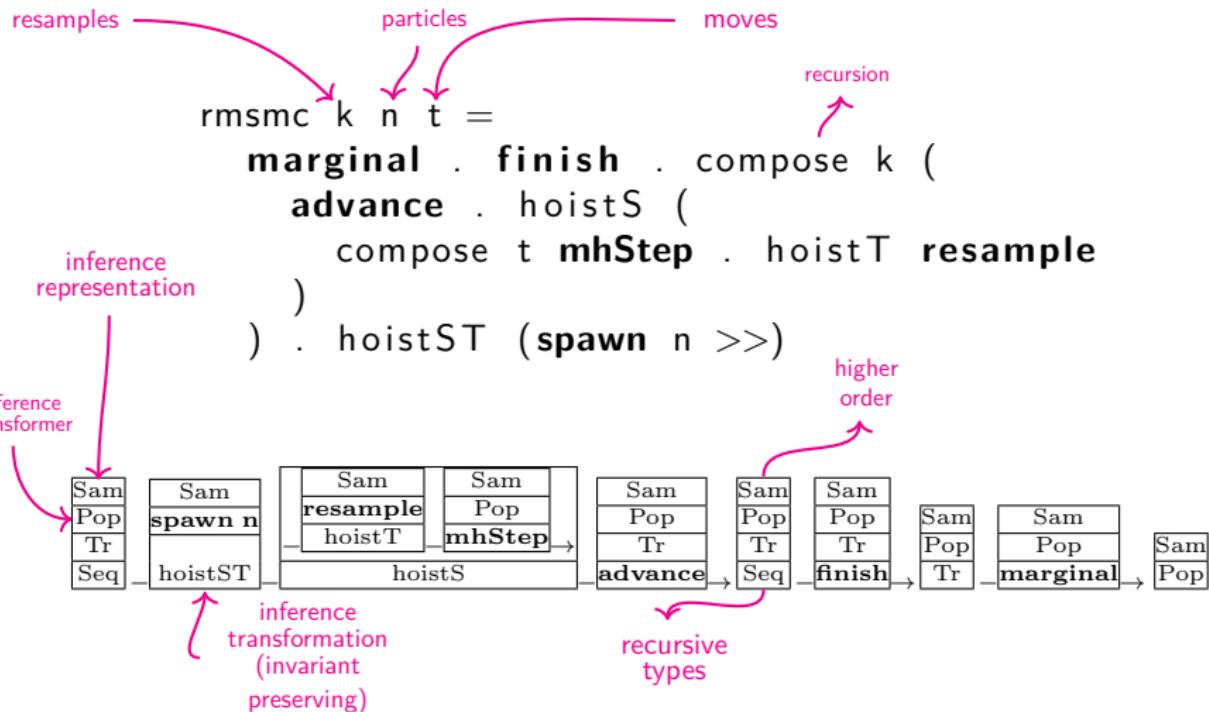
(e.g. dynamic types, IRs)

$$Dynamic = \mu\alpha.\{\text{Val}(\mathbb{R}) \mid \text{Fun}(\alpha \rightarrow \alpha)\}$$

Application: modular Bayesian inference

Resample-Move Sequential Monte Carlo

[Ścibior et al.'18a+b]



ProbProg: Important Language Features



Church Θ Venture	sample	\mathbb{R}	score	higher term	type	Fubini
	order	rec	rec	(commute)		
sets + probability	✓	✗	✗	✓	✗	✗
meas space + subprobability	✓	✓	✗	✗	1 st	✗
CPO + subprobability	✓	✓	✗	✓	✓	?
cont domain + subprobability	✓	✓	✗	✗	1 st	✗
[Jones-Plotkin'89]						
⋮ [Jung-Tix'98]	⋮	⋮	⋮	⋮	⋮	⋮
meas + s-finite distributions	✓	✓	✓	✗	1 st	✗
[Staton'17]						
qbs + s-finite distributions	✓	✓	✓	✓	1 st	✗
[Heunen et al'17, Ścibor et al'18]						
coh/meas cone + probability	✓	✓	✗	✓	?	?
[Ehrhard-Pagani-Tasson'18, Ehrhard-Tasson'15-'19]		✗			✓	✓
ωqbs + s-finite distributions	✓	✓	✓	✓	✓	✓
[This work]						

Summary

Contribution

- ▶ $\omega\mathbf{Qbs}$: a category of pre-domain quasi-Borel spaces
- ▶ M : commutative probabilistic powerdomain over $\omega\mathbf{Qbs}$
- ▶ Axiomatic treatment of measure and domain theory in $\omega\mathbf{Qbs}$
- ▶ Adequacy: $(\omega\mathbf{Qbs}, M)$ adequately interprets:
 - ▶ Statistical FPC
 - ▶ Untyped Statistical λ -calculus

This talk

- ▶ $\omega\mathbf{Qbs}$
- ▶ A probabilistic powerdomain
- ▶ Axiomatic treatment
- ▶ Characterising $\omega\mathbf{Qbs}$

Iso-recursive types: FPC

type variable contexts

$$\Delta = \{\alpha_1, \dots, \alpha_n\}$$

[Fiore-Plotkin'94]

$$\frac{\Delta, \alpha \vdash_k \tau : \text{type}}{\Delta \vdash_k \mu\alpha.\tau : \text{type}}$$

$$Lam = \mu\alpha.\{\text{Bool}\{\text{True} \mid \text{False}\}$$

$$| \text{App}(\alpha * \alpha)$$

$$| \text{Abs}(\alpha \rightarrow \alpha)\}$$

type recursion

$$\frac{\Gamma \vdash t : \sigma[\alpha \mapsto \tau]}{\Gamma \vdash \tau.\text{roll}(t) : \tau}$$

$$\frac{\begin{array}{c} \tau = \mu\alpha.\sigma \\ \Gamma \vdash t : \tau \\ \Gamma, x : \sigma[\alpha \mapsto \tau] \vdash s : \rho \end{array}}{\Gamma \vdash \text{match } t \text{ with roll } x \Rightarrow s : \rho}$$

Iso-recursive types: FPC

[Fiore-Plotkin'94]

type variable contexts

$$\Delta = \{\alpha_1, \dots, \alpha_n\}$$

$$\frac{\Delta, \alpha \vdash_k \tau : \text{type}}{\Delta \vdash_k \mu\alpha.\tau : \text{type}}$$

ω Cpo-enriched
category of
domains

$$[\![\Delta \vdash_k \tau : \text{type}]\!] : (\mathcal{C}^{\text{op}})^n \times \mathcal{C}^n \rightarrow \mathcal{C}$$

$$[\![\Delta \vdash_k \mu\alpha.\tau : \text{type}]\!] = \text{minimal invariants}$$

[Freyd'91,92, Pitts'96]

locally continuous
functor

Challenge

- ▶ probabilistic powerdomain

- ▶ commutativity/Fubini

- ▶ domain theory

- ▶ higher-order functions

traditional approach:

domain \mapsto Scott-open sets \mapsto Borel sets \mapsto distributions/valuations

our approach:

as in
[Ehrhard-Pagani-Tasson'18]

(domain, quasi-Borel space) \mapsto distributions

separate
but compatible

continuous domains
[Jones-Plotkin'89]

open problem
[Jung-Tix'98]

Rudimentary measure theory

Borel sets

- ▶ $[a, b]$ Borel
- ▶ A Borel $\implies A^c$ Borel
- ▶ $(A_n)_{n \in \mathbb{N}}$ Borel $\implies \bigcup_{n \in \mathbb{N}} A_n$ Borel

Measurable functions $f : \mathbb{R} \rightarrow \mathbb{R}$

$$f^{-1}[A] \text{ Borel} \iff A \text{ Borel}$$

Measures $\mu : \text{Borel} \rightarrow [0, \infty]$

- ▶ monotone:
 $A \subseteq B \implies \mu(A) \leq \mu(B)$
- ▶ Scott-continuous:
 $A_0 \subseteq A_1 \subseteq \dots \implies \mu(\bigcup_n A_n) = \bigvee_n \mu(A_n)$
- ▶ strict ($\mu\emptyset = 0$) and additive ($\mu(A \uplus B) = \mu A + \mu B$)

Rudimentary measure theory

Borel sets

- ▶ $[a, b]$ Borel
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- ▶ $(A_n)_{n \in \mathbb{N}}$ Borel $\implies \bigcup_{n \in \mathbb{N}} A_n$ Borel

1 dimensional

Example (Lebesgue measures)

$$\begin{aligned}\lambda[a, b] &= b - a \text{ on } \mathbb{R} \\ (\lambda \otimes \lambda)([a, b] \times [c, d]) &= \\ (b - a)(d - c) &\quad \text{on } \mathbb{R}^2\end{aligned}$$

Measurable functions $f: \mathbb{R} \rightarrow \mathbb{R}$

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▶ strict ($\mu\emptyset = 0$) and additive ($\mu(A \uplus B) = \mu A + \mu B$)

2 dimensional

Example (Push-forward measure)

$$f_*\mu(A) := \mu(f^{-1}[A])$$

Borel set

measure

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

Quasi-Borel pre-domains

ω -qbs:

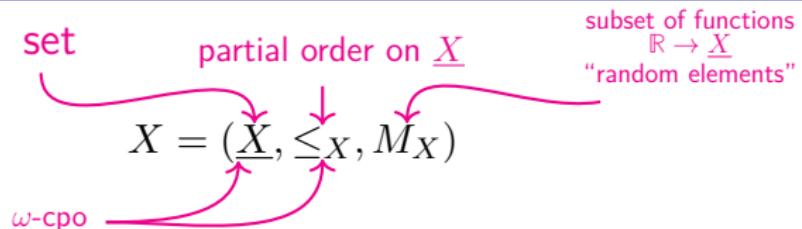
$$X = (\underline{X}, \leq_X, M_X)$$

partial order on \underline{X}

subset of functions
 $\mathbb{R} \rightarrow \underline{X}$
 "random elements"

Quasi-Borel pre-domains

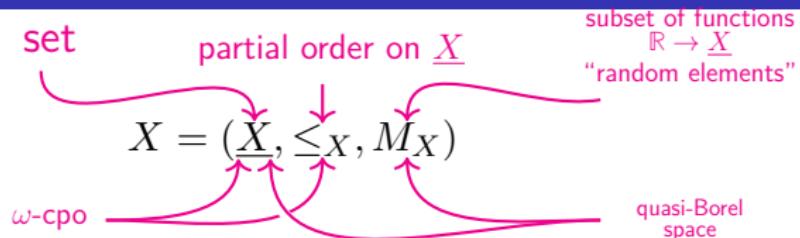
ω -qbs:



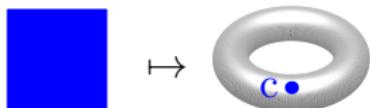
$$\blacksquare x_0 \leq x_1 \leq x_2 \leq \dots \implies \exists \bigvee_n x_n$$

Quasi-Borel pre-domains

ω -qbs:

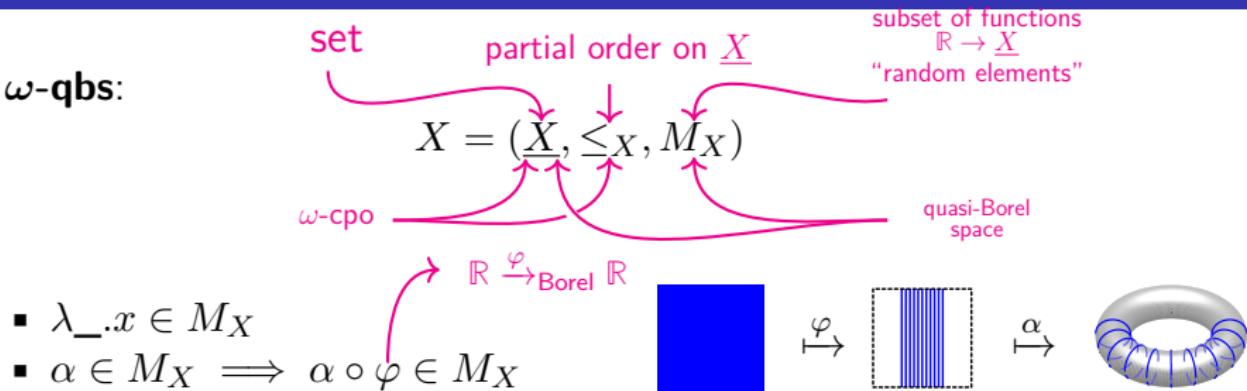


- $\lambda _. x \in M_X$



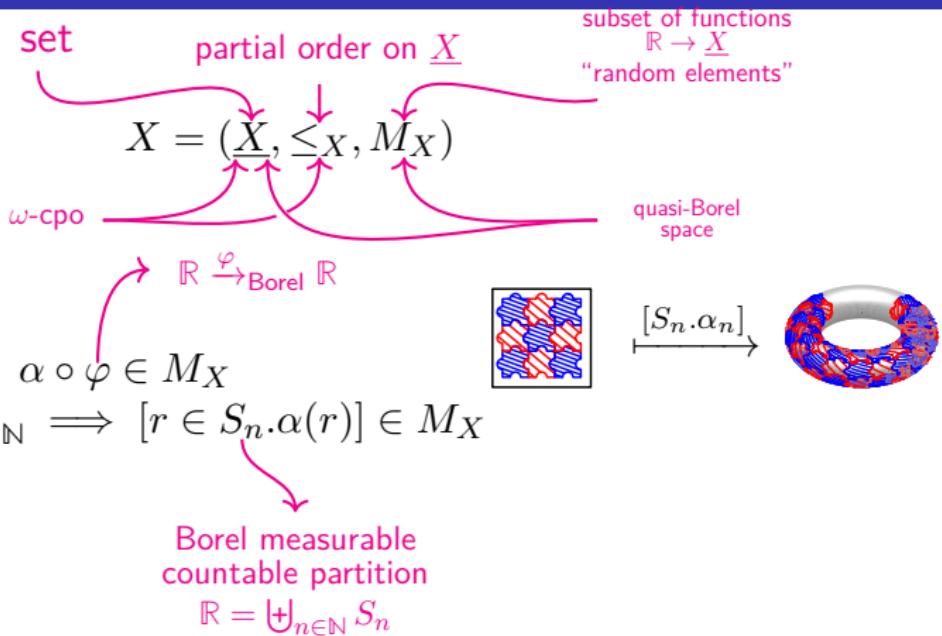
Quasi-Borel pre-domains

$\omega\text{-qbs}$:



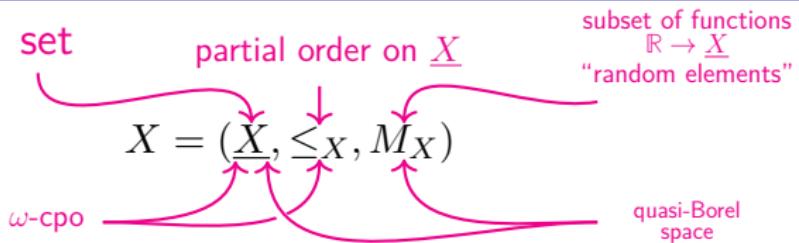
Quasi-Borel pre-domains

ω-qbs:



Quasi-Borel pre-domains

$\omega\text{-qbs}$:



- $\lambda _. x \in M_X$
- $\alpha \in M_X \implies \alpha \circ \varphi \in M_X$
- $(\alpha_n \in M_X)_{n \in \mathbb{N}} \implies [r \in S_n. \alpha(r)] \in M_X$

s.t.:

pointwise
 ω -chain

$$(\alpha_n) \in M_X^\omega$$

Borel measurable
countable partition

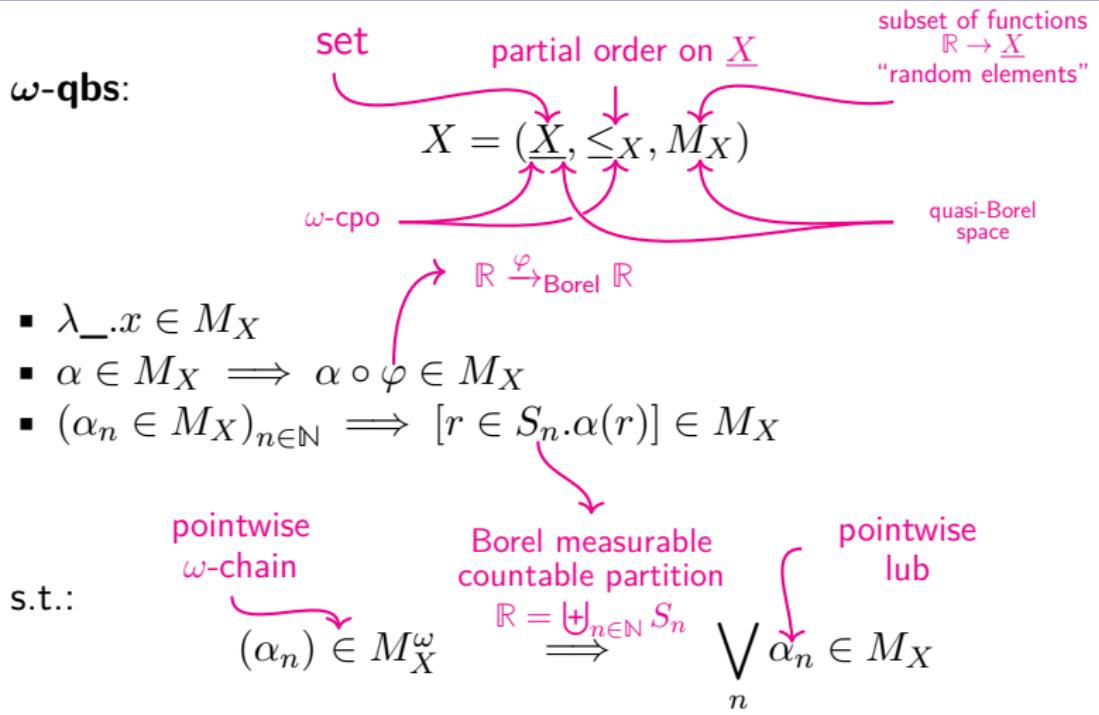
$$\mathbb{R} = \biguplus_{n \in \mathbb{N}} S_n$$

pointwise
lub

$$\bigvee_n \alpha_n \in M_X$$

Quasi-Borel pre-domains

$\omega\text{-qbs}:$



Morphisms $f : X \rightarrow Y$: Scott continuous qbs maps

monotone and
 $f \bigvee_n x_n = \bigvee_n f x_n$

$\forall \alpha \in M_X.$
 $f \circ \alpha \in M_Y$

Quasi-Borel pre-domains

Example

$S = (\underline{S}, \Sigma_S)$ measurable space



$$(\underline{S}, =, \{\alpha : \mathbb{R} \rightarrow \underline{S} \mid \alpha \text{ Borel measurable}\})$$

so $\mathbb{R} \in \omega\mathbf{Qbs}$

Reminder

wqbs: $X = (\underline{X}, \leq_X, M_X)$

- $\lambda _. x \in M_X$
- $\alpha \in M_X \implies \alpha \circ \varphi \in M_X$
- $(\alpha_n \in M_X)_{n \in \mathbb{N}} \implies [r \in S_n. \alpha(r)] \in M_X$

s.t.:

$$(\alpha_n) \in M_X^\omega \implies \bigvee_n \alpha_n \in M_X$$

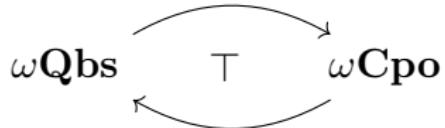
Quasi-Borel pre-domains

Example

$$P = (\underline{P}, \leq_P) \text{ } \omega\text{-cpo}$$

lubs of
step functions

$$\left(\underline{P}, \leq_P, \left\{ \bigvee_k [_ \in S_n^k . a_n^k] \middle| \forall k. \mathbb{R} = \biguplus_n S_n^k \right\} \right)$$



so $\mathbb{L} = ([0, \infty], \leq, \{\alpha : \mathbb{R} \rightarrow [0, \infty] | \alpha \text{ Borel measurable}\}) \in \omega\text{Qbs}$

Reminder

$$\text{wqbs: } X = (\underline{X}, \leq_X, M_X)$$

- $\lambda _. x \in M_X$
- $\alpha \in M_X \implies \alpha \circ \varphi \in M_X$
- $(\alpha_n \in M_X)_{n \in \mathbb{N}} \implies [r \in S_n . \alpha(r)] \in M_X$

s.t.:

$$(\alpha_n) \in M_X^\omega \implies \bigvee_n \alpha_n \in M_X$$

Quasi-Borel pre-domains

Example

X ω -qbs

$\omega\mathbf{Qbs}$

$$X_{\perp} := \left(\{\perp\} + \underline{X}, \perp \leq \underline{X}, \left\{ [S.\perp, S^{\complement}.\alpha] \middle| \alpha \in M_X, S \text{ Borel} \right\} \right)$$

Reminder

wqbs: $X = (\underline{X}, \leq_X, M_X)$

- $\lambda _. x \in M_X$
- $\alpha \in M_X \implies \alpha \circ \varphi \in M_X$
- $(\alpha_n \in M_X)_{n \in \mathbb{N}} \implies [r \in S_n.\alpha(r)] \in M_X$

s.t.:

$$(\alpha_n) \in M_X^\omega \implies \bigvee_n \alpha_n \in M_X$$

Quasi-Borel pre-domains

Products

$$\underline{X_1 \times X_2} = \underline{X}_1 \times \underline{X}_2 \quad x \leq y \iff \forall i.x_i \leq y_i$$

$$M_{X_1 \times X_2} = \{(\alpha_1, \alpha_2) : \mathbb{R} \rightarrow \underline{X}_1 \times \underline{X}_2 \mid \forall i.\alpha_i \in M_{X_i}\}$$



correlated
random elements

Quasi-Borel pre-domains

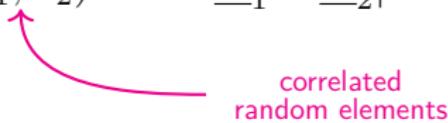
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Theorem

$\omega\mathbf{Qbs} \rightarrow \omega\mathbf{Cpo} \times \mathbf{Qbs}$ creates limits



Quasi-Borel pre-domains

Products

$$\underline{X_1 \times X_2} = \underline{X}_1 \times \underline{X}_2 \quad x \leq y \iff \forall i.x_i \leq y_i$$

$$M_{X_1 \times X_2} = \{(\alpha_1, \alpha_2) : \mathbb{R} \rightarrow \underline{X}_1 \times \underline{X}_2 \mid \forall i.\alpha_i \in M_{X_i}\}$$

↑
correlated
random elements

Exponentials

- ▶ $\underline{Y^X} = \{f : \underline{X} \rightarrow \underline{Y} \mid f \text{ Scott continuous qbs morphism}\}$
 $= \mathbf{Qbs}(X, Y)$
- ▶ $f \leq g \iff \forall x \in \underline{X}. f(x) \leq g(x)$
- ▶ $M_{Y^X} = \left\{ \alpha : \mathbb{R} \rightarrow \underline{Y^X} \middle| \begin{array}{l} \text{uncurry } \alpha : \mathbb{R} \times X \rightarrow Y \\ \text{Scott continuous qbs morphism} \end{array} \right\}$
so $\underline{Y^\mathbb{R}} = M_Y$

Fundamentals of measure theory

s-finite measures

- ▶ μ_n **bounded**: $\mu_n(\mathbb{R}) < \infty$
- ▶ μ **s-finite**: $\mu = \sum_n \mu_n$, μ_n bounded

Randomisation Theorem

Every s-finite measure is a push-forward of Lebesgue:

$$\mu \text{ s-finite} \implies \mu = f_*\lambda \text{ for some } f : \mathbb{R} \rightarrow \mathbb{R}_\perp$$

Transfer principle

$$\tau_*\lambda = \lambda \otimes \lambda \text{ for some measurable } \tau : \mathbb{R} \rightarrow (\mathbb{R} \times \mathbb{R})_\perp$$

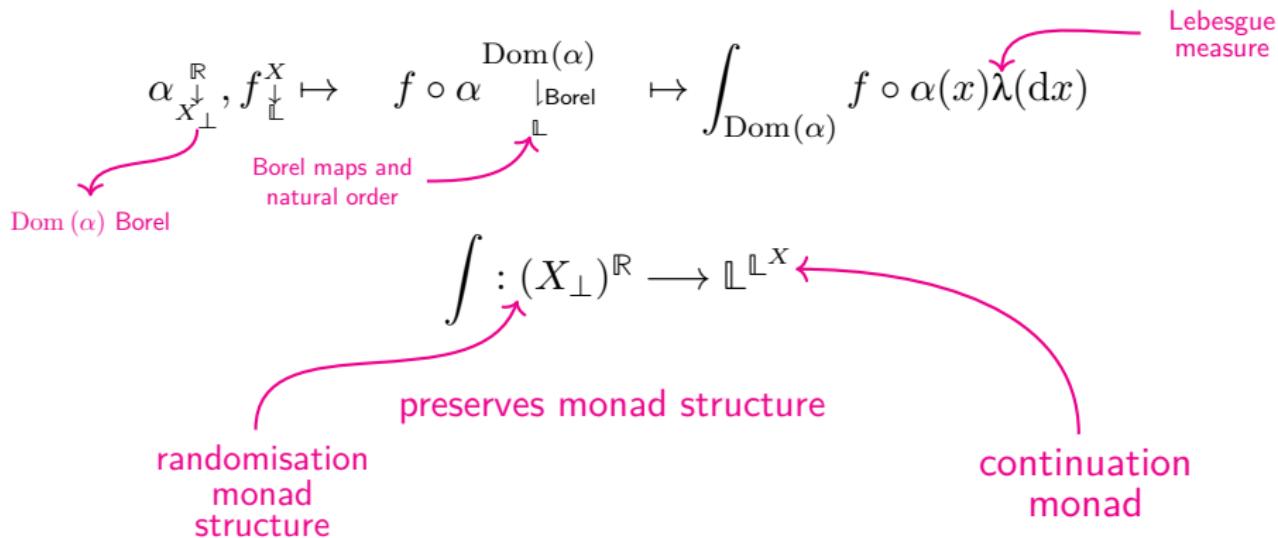
Randomisation monad structure

- ▶ $(X_\perp)^\mathbb{R}$
- ▶ $\text{return}_X(x) : r \in [0, 1] \mapsto x$
- ▶ $(\alpha \gg= f) : \mathbb{R} \xrightarrow{\tau} \mathbb{R} \times \mathbb{R} \xrightarrow{\mathbb{R} \times \alpha} \mathbb{R} \times X \xrightarrow{\mathbb{R} \times f} \mathbb{R} \times (Y_\perp)^\mathbb{R} \xrightarrow{\text{eval}} Y$

$\mathbb{R} \rightarrow X_\perp$ $X \rightarrow (Y_\perp)^\mathbb{R}$
- ▶ sample from randomisation of normal distribution
- ▶ $\text{score}(r) : r' \in [0, |r|] \mapsto ()$

monad laws fail
(associativity)

Lebesgue integration

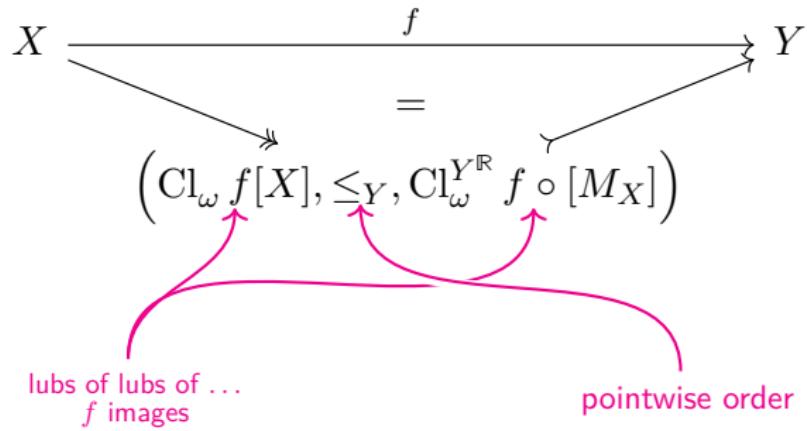


A probabilistic powerdomain

$$(X_\perp)^\mathbb{R} \xrightarrow{\int} \mathbb{L}^{\mathbb{L}^X}$$
$$MX$$

MX : randomisable integration operators

A probabilistic powerdomain



$(\mathcal{E}, \mathcal{M}) := (\text{densely strong epi, full mono}) \text{ factorisation system}$

A probabilistic powerdomain

\mathcal{E} = densely strong epis closed under:

▶ products:

$$e_1, e_2 \in \mathcal{E} \implies e_1 \times e_2 \in \mathcal{E}$$

▶ lifting:

$$e \in \mathcal{E} \implies e_{\perp} \in \mathcal{E}$$

▶ random elements:

$$e \in \mathcal{E} \implies e^{\mathbb{R}} \in \mathcal{E}$$

$\implies M$ strong monad for sampling + conditioning

[Kammar-McDermott'18]

A probabilistic powerdomain

$$(X_{\perp})^{\mathbb{R}} \xrightarrow{\quad} \mathbb{L}^{\mathbb{L}^X}$$

=

$$MX$$

- ▶ M locally continuous \implies may appear in domain equations

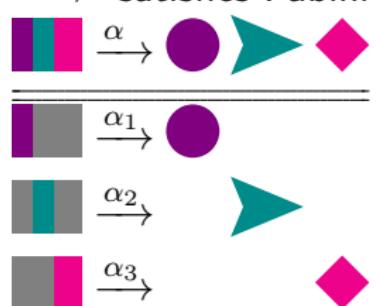
- ▶ M commutative

\implies satisfies Fubini

- ▶ M models synthetic measure theory

$$M \sum_{n \in \mathbb{N}} X_n \cong \prod_{n \in \mathbb{N}} MX_n$$

[Kock'12,
Ścibior et al.'18]



- ▶ $MX \cong \left\{ \mu \middle| \text{Scott opens} \middle| \mu \text{ is s-finite} \right\}$ generalises valuations

standard Borel space

Axiomatic domain theory

[Fiore-Plotkin'94, Fiore'96]

Structure

- ▶ Total map category: $\omega\mathbf{Qbs}$

- ▶ Admissible monos: **Borel-open** map $m : X \rightarrowtail Y$:

$$\forall \beta \in M_Y. \quad \beta^{-1}[m[X]] \in \mathcal{B}(\mathbb{R})$$



take Borel-Scott open maps as admissible monos

- ▶ Pos-enrichment: pointwise order
- ▶ Pointed monad on total maps: the powerdomain

⇒ model axiomatic domain theory

⇒ solve recursive domain equations

Axiomatic domain theory

Structure

- \mathfrak{D} total map category
- $\omega\mathbf{Qbs}$
- $f \leq g$ Pos-enrichment pointwise order
- $\mathcal{M}_{\mathfrak{D}}$ admissible monos Borel-Scott opens
- T monad for effects power-domain
- m partiality encoding $m : -_{\perp} \rightarrow T, \perp \mapsto \underline{0}$

Derived axioms/structure

- $p\mathfrak{D}$ partial map category
- $-_{\perp}$ partiality monad
- (\dashv_V) the adjunction $J \dashv L$ is locally continuous
- (p_V) $p\mathfrak{D}$ is $\omega\mathbf{Cpo}$ -enriched
- $(\mathbb{1}_{\leq})$ $p\mathfrak{D}$ has a partial terminal

Axioms

- (\dashv) every object has a partial map classifier $\downarrow_X : X \rightarrow X_{\perp}$
- (fup) every admissible mono is full (+) and upper-closed
- (\dashv_{\leq}) $\lfloor - \rfloor$ is locally monotone
- (V) \mathfrak{D} is $\omega\mathbf{Cpo}$ -enriched
- (U) ω -colimits behave uniformly
- $(\mathbb{1})$ \mathfrak{D} has a terminal object
- (\rightarrow_{\leq}) \mathfrak{D} has locally monotone exponentials
- $(+)$ locally continuous total coproducts
- $(?!)$ $\emptyset \rightarrow \mathbb{1}$ is admissible
- (\times_V) \mathfrak{D} has a locally continuous products
- (CL) \mathfrak{D} is cocomplete
- (T_V) T is locally continuous
- (\otimes) $p\mathfrak{D}$ has partial products
- (\otimes_V) (\otimes) is locally continuous
- (\rightarrow_V) \mathfrak{D} has locally continuous exponentials
- (\Rightarrow_V) $p\mathfrak{D}$ has locally continuous partial exponentials
- (pCL) $p\mathfrak{D}$ is cocomplete
- $(p+V)$ $p\mathfrak{D}$ has locally continuous partial coproducts
- (BC) $J : \mathfrak{D} \hookrightarrow p\mathfrak{D}$ is a bilimit compact expansion

Characterising ωQbs

$$\begin{array}{ccc} \mathbf{Sbs} & \xleftarrow{\text{Yoneda}} & [\mathbf{Sbs}^{\text{op}}, \mathbf{Set}]_{\text{cpp}} \\ \text{Yoneda} \swarrow & = & \searrow \\ & \mathbf{SepSh} & \end{array}$$

countable
product preserving
[Staton et al.'16]

$F : \mathbf{Sbs}^{\text{op}} \rightarrow \mathbf{Set}$ separated: $F\mathbb{R} \xrightarrow{\left(F(\mathbb{R} \xleftarrow{r} \mathbb{1})\right)_{r \in \mathbb{R}}} (F\mathbb{1})^{\mathbb{R}}$ injective

Thm: $\mathbf{Qbs} \simeq \mathbf{SepSh}$

[Heunen et al.'17]

$$\begin{array}{ccc} \mathbf{Sbs} & \xleftarrow{\text{Yoneda}} & [\mathbf{Sbs}^{\text{op}}, \omega\mathbf{Cpo}]_{\text{cpp}} \\ \text{Yoneda} \swarrow & = & \searrow \\ & \omega\mathbf{SepSh} & \end{array}$$

$$f(x) \leq f(y) \implies x \leq y$$

$F : \mathbf{Sbs}^{\text{op}} \rightarrow \omega\mathbf{Cpo}$ ω -separated: $F\mathbb{R} \xrightarrow{\left(F(\mathbb{R} \xleftarrow{r} \mathbb{1})\right)_{r \in \mathbb{R}}} (F\mathbb{1})^{\mathbb{R}}$

Thm: $\omega\mathbf{Qbs} \simeq_{\omega\mathbf{Cpo}} \omega\mathbf{SepSh}$

full

Characterising ωQbs

Grothendieck quasi-topos **Qbs**
strong subobject classifier:

$$\begin{array}{ccc} \Omega = \mathcal{Q} & & M_\Omega = 2^{\mathbb{R}} \\ \text{strong monos:} & & \\ X \xrightarrow{f} Y & P^2 \xrightarrow{\leq_P} \Omega & \omega\text{-chain}(P) \xrightarrow{\vee} P \\ (f \circ)^{-1}[M_Y] = M_X & \text{qbs} & \\ \text{Internal } \omega\text{-cpo } P: & (\underline{P}, \leq_P, \vee) & \end{array}$$

+ internal quasi-topos logic ω -cpo axioms

Theorem

$$\omega\text{Qbs} \simeq \omega\text{Cpo}(\text{Qbs})$$

Characterising $\omega\mathbf{Qbs}$

By local presentability:

$$\begin{array}{ccc} \omega\mathbf{Cpo} \simeq \mathrm{Mod}(\omega\mathbf{cpo}, \mathbf{Set}) & & \mathbf{Qbs} \simeq \mathrm{Mod}(\mathbf{qbs}, \mathbf{Set}) \\ \text{essentially algebraic theories} \\ \omega\mathbf{qbs}: \quad \omega\mathbf{cpo} \cup \mathbf{qbs} \cup \text{compatibility axiom} \end{array}$$

Theorem

$$\omega\mathbf{Qbs} \simeq \mathrm{Mod}(\omega\mathbf{qbs}, \mathbf{Set})$$

so $\omega\mathbf{Qbs}$ locally presentable, hence cocomplete

Summary

Contribution

- ▶ $\omega\mathbf{Qbs}$: a category of pre-domain quasi-Borel spaces
- ▶ M : commutative probabilistic powerdomain over $\omega\mathbf{Qbs}$
- ▶ Axiomatic treatment of measure and domain theory in $\omega\mathbf{Qbs}$
- ▶ Adequacy: $(\omega\mathbf{Qbs}, M)$ adequately interprets:
 - ▶ Statistical FPC
 - ▶ Untyped Statistical λ -calculus

[Fiore-Plotkin'94, Fiore'96]

This talk

- ▶ $\omega\mathbf{Qbs}$
- ▶ A probabilistic powerdomain
- ▶ Axiomatic treatment
- ▶ Characterising $\omega\mathbf{Qbs}$

Also in the paper

- ▶ Axiomatic domain theory
- ▶ Operational semantics
à la [Borgström et al.'16]

ProbProg: Important Language Features



Church Θ Venture	sample	\mathbb{R}	score	higher term	type	Fubini
	order	rec	rec	(commute)		
sets + probability	✓	✗	✗	✓	✗	✗
meas space + subprobability	✓	✓	✗	✗	1 st	✗
CPO + subprobability	✓	✓	✗	✓	✓	?
cont domain + subprobability	✓	✓	✗	✗	1 st	✗
[Jones-Plotkin'89]						
⋮ [Jung-Tix'98]	⋮	⋮	⋮	⋮	⋮	⋮
meas + s-finite distributions	✓	✓	✓	✗	1 st	✗
[Staton'17]						
qbs + s-finite distributions	✓	✓	✓	✓	1 st	✗
[Heunen et al'17, Ścibor et al'18]						
coh/meas cone + probability	✓	✓	✗	✓	?	?
[Ehrhard-Pagani-Tasson'18, Ehrhard-Tasson'15-'19]		✗			✓	✓
ωqbs + s-finite distributions	✓	✓	✓	✓	✓	✓
[This work]						

Presenting Pos

[cf. Adámek-Rosicky'94
Examples 3.35(1),(4)]

Sorts/arities

elem

ineq

Operations

lower : ineq → elem

upper : ineq → elem

refl : elem → ineq

irrel : ineq × ineq → ineq

Def(irrel(e_1, e_2)):
lower(e_1) = lower(e_2)
upper(e_1) = upper(e_2)

antisym : ineq × ineq → elem

Def(antisym(e, e^{op})):
lower(e) = upper(e^{op})
upper(e) = lower(e^{op})

trans : ineq × ineq → ineq

Def(trans(e_1, e_2)):
upper(e_1) = lower(e_2)

Axioms

$e_1 = \text{irrel}(e_1, e_2) = e_2$

lower(refl(x)) = x = upper(refl(x))

lower(e_1) = antisym(e_1, e_2) = lower(e_2)

lower(trans(e_1, e_2)) = lower(e_1)

upper(trans(e_1, e_2)) = upper(e_2)

Presenting ωCpo

Add to **pos**:

Operations

$$\bigvee : \prod_{n \in \mathbb{N}} \text{ineq} \multimap \text{elem}$$

$$\mathbf{ub}_k : \prod_{n \in \mathbb{N}} \text{ineq} \multimap \text{ineq}$$

$$\mathbf{lst} : \text{elem} \times \prod_{n \in \mathbb{N}} \text{ineq} \times \prod_{n \in \mathbb{N}} \text{ineq} \multimap \text{ineq}$$

Axioms

$$\begin{aligned}\mathbf{lower}(\mathbf{ub}_k(e_n)) &= \mathbf{lower}(e_k) \\ \mathbf{upper}(\mathbf{ub}_k(e_n)) &= \bigvee(e_n)\end{aligned}$$

$$\mathbf{lower}(\mathbf{lst}(x, (e_n), (b_n))) = \bigvee(e_n) \quad \mathbf{upper}(\mathbf{lst}(x, (e_n), (b_n)_n)) = x$$

Def($\bigvee_{n \in \mathbb{N}} e_n$):

$\mathbf{upper}(e_n) = \mathbf{lower}(e_{n+1})$
for each $n \in \mathbb{N}$

Def($(\mathbf{ub}_k(e_n))_{n \in \mathbb{N}}$):

$\mathbf{upper}(e_n) = \mathbf{lower}(e_{n+1})$
for each $n \in \mathbb{N}$

Def($\mathbf{lst}(x, (e_n), (b_n))$):

$\mathbf{upper}(e_n) = \mathbf{lower}(e_{n+1})$

$\mathbf{upper}(b_n) = x$

$\mathbf{lower}(e_n) = \mathbf{lower}(b_n)$

for each $n \in \mathbb{N}$

Presenting Qbs

Sorts/arities

elem

rand

Operations

$\text{ev}_r : \text{rand} \rightarrow \text{elem}$

$\text{const} : \text{elem} \rightarrow \text{rand}$

$\text{rearrange}_\varphi : \text{rand} \rightarrow \text{rand}$

$\text{match}_{(S_i)_{i \in I}} : \prod_{i \in I} \text{rand} \rightarrow \text{rand}$

$\text{ext} : \text{rand} \times \text{rand} \multimap \text{rand}$

Def($\text{ext}(\alpha, \beta)$):

$\text{ev}_r(\alpha) = \text{ev}_r(\beta)$

for each $r \in \mathbb{R}$

Axioms

$$\alpha = \text{ext}(\alpha, \beta) = \beta$$

$$\text{ev}_r(\text{const}(x)) = x$$

$$\text{ev}_r(\text{rearrange}_\varphi \alpha) = \text{ev}_{\varphi(r)} \alpha$$

$$\text{ev}_r \left(\text{match}_{(S_i)_{i \in I}} (\alpha_i)_{i \in I} \right) = \text{ev}_r(\alpha_i)$$

Presenting ω Qbs

Sorts/arities

elem

ineq

rand

Operations

Add to ω cpo and qbs:

$\sqcup : \prod_{n \in \mathbb{N}} \text{rand} \times \prod_{n \in \mathbb{N}, r \in \mathbb{R}} \text{ineq} \rightarrow \text{rand}$

Def $\sqcup((\alpha_n)_{n \in \mathbb{N}}, (e_n^r)_{n \in \mathbb{N}, r \in \mathbb{R}})$:

$$\text{lower}(e_n^r) = \text{ev}_r(\alpha_n) \quad \text{upper}(e_n^r) = \text{ev}_r(\alpha_{n+1})$$

for each $n \in \mathbb{N}, r \in \mathbb{R}$

Axioms

Add:

$$\text{ev}_r \left(\sqcup \left((\alpha_n)_{n \in \mathbb{N}}, (e_n^r)_{n \in \mathbb{N}, r \in \mathbb{R}} \right) \right) = \bigvee (e_n^r)_{n \in \mathbb{N}}$$