

# A domain theory for statistical probabilistic programming

Matthijs Vákár, Ohad Kammar, and Sam Staton



Second Workshop on  
Probabilistic Interactive and Higher-Order Computation  
6-7 February 2019



# Statistical probabilistic programming

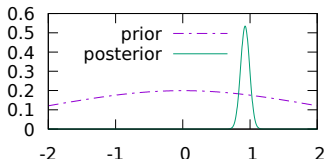
$\llbracket - \rrbracket$  : programs  $\rightarrow$  unnormalised distributions

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

normally  
distributed  
sample

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

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



scale  
distribution  
by  $r$

$r : [0, \infty]$   
**score**( $r$ ) : 1

conditioning/fitting  
to observed data  
with likelihood

prior

```
let  $x = \text{sample}(0, 2)$   
in score( $\text{normalPdf}(1.1 \mid x, \frac{1}{4})$ );  
    score( $\text{normalPdf}(1.9 \mid 2x, \frac{1}{4})$ );  
    score( $\text{normalPdf}(2.7 \mid 3x, \frac{1}{4})$ );  
 $x$ 
```

posterior

# Statistical probabilistic programming

## ► Commutativity/exchangability/Fubini

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

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

probability  
distributions



$\sigma$ -finite  
distributions



arbitrary  
distributions



s-finite  
distributions



not closed under  
push-forward

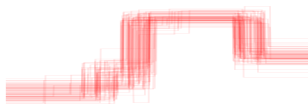
full definability  
[Staton'17]

# Statistical probabilistic programming

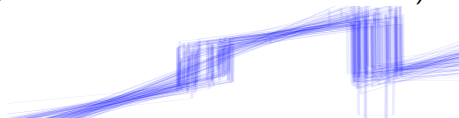
Express continuous distributions using:

- Higher-order functions:

(e.g. generative random function models)



*piecewise(random-constant)*

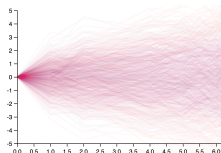


*piecewise(random-linear)*

- Term recursion:

```
rw(x, σ) = λ().    // thunk  
  let y = sample(x, σ)  
  in (x, rw(y, σ))
```

(e.g. Gaussian random walk)



- 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]

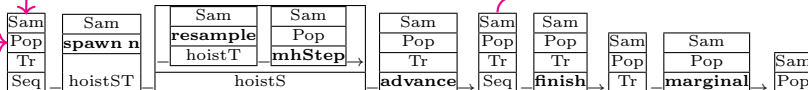
resamples  $k$  particles  $n$  moves  $t$  recursion

```
rmsmc k n t =  
  marginal . finish . compose k (  
    advance . hoistS (  
      compose t mhStep . hoistT resample  
    )  
  ) . hoistST (spawn n >>)
```

inference representation

inference transformer


higher order



inference transformation (invariant preserving)

recursive types

# ProbProg: Important Language Features

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

## This talk

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

# Iso-recursive types: FPC

type variable contexts

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

[Fiore-Plotkin'94]

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

$$\begin{aligned} \text{Lam} = \mu\alpha. \{ & \text{Bool}\{\text{True} \mid \text{False}\} \\ & \mid \text{App}(\alpha * \alpha) \\ & \mid \text{Abs}(\alpha \rightarrow \alpha) \} \end{aligned}$$

type recursion

$$\frac{\Gamma \vdash t : \sigma[\alpha \mapsto \tau]}{\Gamma \vdash \tau.\text{roll}(t) : \tau} \quad \frac{\Gamma \vdash t : \tau \quad \Gamma, x : \sigma[\alpha \mapsto \tau] \vdash s : \rho}{\Gamma \vdash \text{match } t \text{ with roll } x \Rightarrow s : \rho}$$



# Iso-recursive types: FPC

[Fiore-Plotkin'94]

type variable contexts

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

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

$\omega$ Cpo-enriched  
category of  
domains

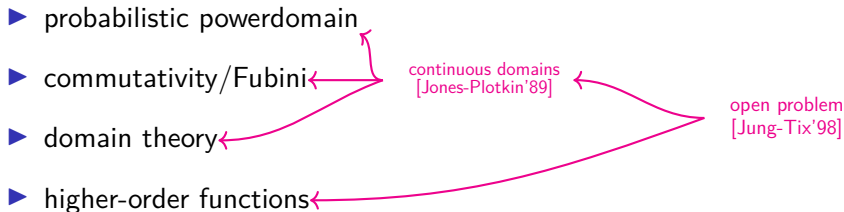
$$\llbracket \Delta \vdash_k \tau : \text{type} \rrbracket : (\mathcal{C}^{\text{op}})^n \times \mathcal{C}^n \rightarrow \mathcal{C}$$

$$\llbracket \Delta \vdash_k \mu\alpha.\tau : \text{type} \rrbracket = \text{minimal invariants}$$

[Freyd'91,92, Pitts'96]

locally continuous  
functor

# Challenge

- ▶ probabilistic powerdomain
  - ▶ commutativity/Fubini
  - ▶ domain theory
  - ▶ higher-order functions
- continuous domains [Jones-Plotkin'89]
- open problem [Jung-Tix'98]
- 

traditional approach:

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

our approach: <sup>as in</sup> [Ehrhard-Pagani-Tasson'18]



(domain, quasi-Borel space)  $\mapsto$  distributions

separate  
but compatible



# 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
- ▶  $A$  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 ( $\mu(A \uplus B) = \mu A + \mu B$ )

1 dimensional

## Example ( Lebesgue measures)

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

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

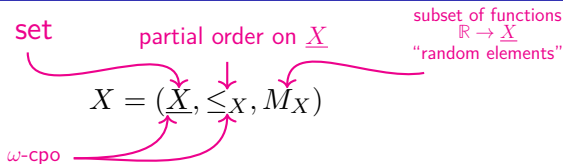
$\omega$ -qbs:

set      partial order on  $\underline{X}$       subset of functions  $\mathbb{R} \rightarrow \underline{X}$   
"random elements"

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

# Quasi-Borel pre-domains

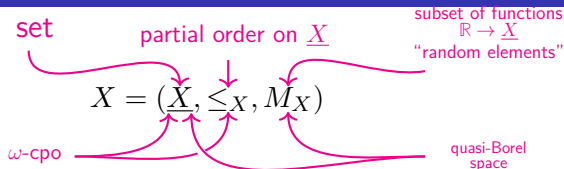
$\omega$ -qbs:



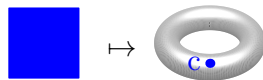
$$\blacksquare x_0 \leq x_1 \leq x_2 \leq \dots \quad \implies \quad \exists \bigvee_n x_n$$

# Quasi-Borel pre-domains

$\omega$ -qbs:

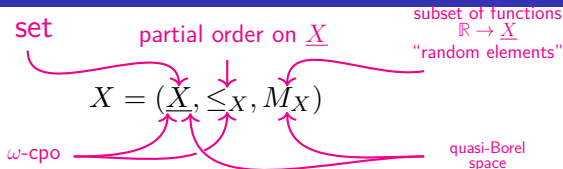


- $\lambda_.x \in M_X$



# Quasi-Borel pre-domains

$\omega$ -qbs:



- $\lambda_.x \in M_X$
- $\alpha \in M_X \implies \alpha \circ \varphi \in M_X$

$$\mathbb{R} \xrightarrow{\varphi}_{\text{Borel}} \mathbb{R}$$



$$\models \varphi$$



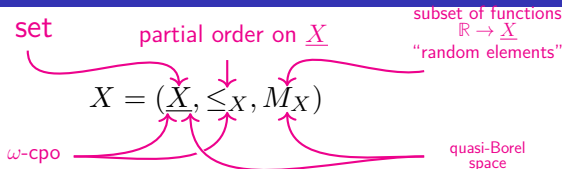
$$\models \alpha$$





# Quasi-Borel pre-domains

$\omega$ -qbs:

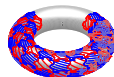


- $\lambda_.x \in M_X$
- $\alpha \in M_X \implies \alpha \circ \varphi \in M_X$
- $(\alpha_n \in M_X)_{n \in \mathbb{N}} \implies [r \in S_n. \alpha(r)] \in M_X$

$$\mathbb{R} \xrightarrow{\varphi} \text{Borel } \mathbb{R}$$



$$[S_n. \alpha_n]$$

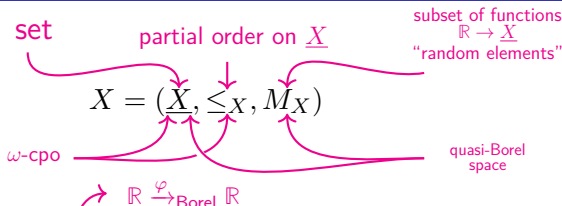


Borel measurable  
countable partition

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

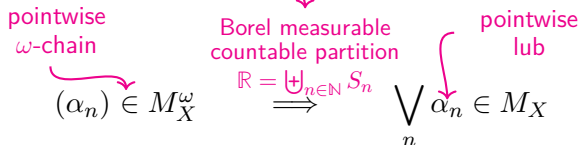
# Quasi-Borel pre-domains

$\omega$ -qbs:



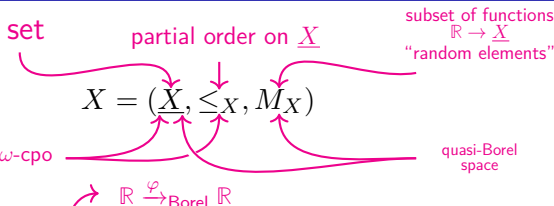
- $\lambda_.x \in M_X$
- $\alpha \in M_X \implies \alpha \circ \varphi \in M_X$
- $(\alpha_n \in M_X)_{n \in \mathbb{N}} \implies [r \in S_n. \alpha(r)] \in M_X$

s.t.:



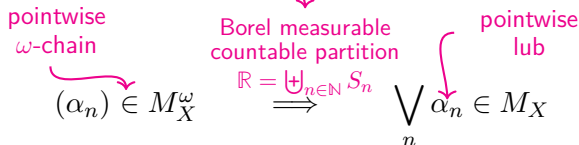
# Quasi-Borel pre-domains

$\omega$ -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.:



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

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

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

s.t.:

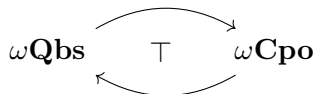
$$(\alpha_n) \in M_X^\omega \implies \bigvee_n \alpha_n \in M_X$$

# Quasi-Borel pre-domains

## Example

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

lubs of  
step functions



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

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

## Reminder

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

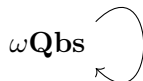
- $\lambda \_ . x \in M_X$
- $\alpha \in M_X \implies \alpha \circ \varphi \in M_X$
- $(\alpha_n \in M_X)_{n \in \mathbb{N}} \implies [r \in S_n \cdot \alpha(r)] \in M_X$

s.t.:

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

## Example

$X$   $\omega$ -qbs



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

## Reminder

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

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

s.t.:

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

# Quasi-Borel pre-domains

## Products

$$\underline{X_1} \times \underline{X_2} = \underline{X_1} \times \underline{X_2} \qquad x \leq y \iff \forall i. x_i \leq y_i$$

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



correlated  
random elements

# Quasi-Borel pre-domains

## Products

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

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

## Theorem

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

correlated  
random elements






# Quasi-Borel pre-domains

## Products

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

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



correlated  
random elements

## Exponentials

►  $\underline{Y^X} = \{f : \underline{X} \rightarrow \underline{Y} \mid f \text{ Scott continuous qbs morphism}\}$   
 $= \mathbf{Qbs}(X, Y)$

►  $f \leq g \iff \forall x \in \underline{X}. f(x) \leq g(x)$

►  $M_{Y^X} = \left\{ \alpha : \mathbb{R} \rightarrow \underline{Y^X} \mid \begin{array}{l} \text{uncurry } \alpha : \mathbb{R} \times X \rightarrow Y \\ \text{Scott continuous qbs morphism} \end{array} \right\}$   
so  $\underline{Y^{\mathbb{R}}} = M_Y$

# Fundamentals of measure theory

## s-finite measures

►  $\mu_n$  **bounded**:

$$\mu_n(\mathbb{R}) < \infty$$

►  $\mu$  **s-finite**:

$$\mu = \sum_n \mu_n, \mu_n \text{ bounded}$$

## 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

▶  $(X_{\perp})^{\mathbb{R}}$

▶  $\text{return}_X(x) : r \in [0, 1] \mapsto x$

▶  $(\alpha \gg= f) : \mathbb{R} \xrightarrow{\tau} \mathbb{R} \times \mathbb{R} \xrightarrow{\mathbb{R} \times \alpha} \mathbb{R} \times X \xrightarrow{\mathbb{R} \times f} \mathbb{R} \times (Y_{\perp})^{\mathbb{R}} \xrightarrow{\text{eval}} Y$

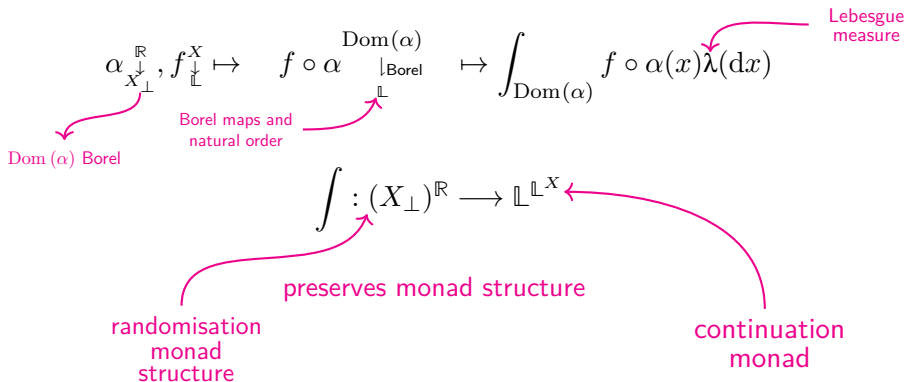
$\mathbb{R} \rightarrow X_{\perp}$        $X \rightarrow (Y_{\perp})^{\mathbb{R}}$

▶ sample from randomisation of normal distribution

▶  $\text{score}(r) : r' \in [0, |r|] \mapsto ()$

monad laws fail  
(associativity)

# Lebesgue integration

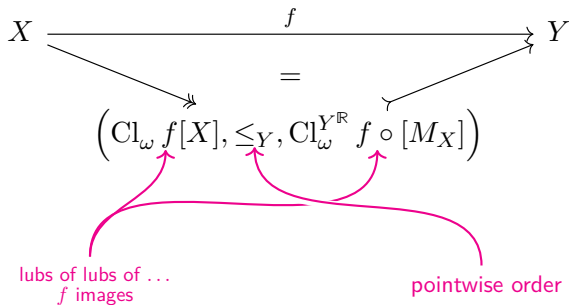


# A probabilistic powerdomain

$$\begin{array}{ccc} (X_{\perp})^{\mathbb{R}} & \xrightarrow{\int} & \mathbb{L}^X \\ & \searrow \quad \swarrow & \\ & = & \\ & MX & \end{array}$$

$MX$ : randomisable integration operators

# A probabilistic powerdomain



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

$$(X_{\perp})^{\mathbb{R}} \begin{array}{c} \xrightarrow{\quad} \\ \searrow \quad \nearrow \\ \quad \quad \quad \end{array} \begin{array}{c} \mathbb{L}^{\mathbb{L}^X} \\ = \\ MX \end{array}$$

►  $M$  locally continuous  $\implies$  may appear in domain equations

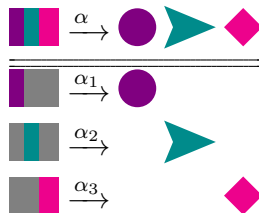
►  $M$  commutative

►  $M$  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]

$\implies$  satisfies Fubini



►  $MX \cong \left\{ \mu \mid \text{Scott opens} \mid \mu \text{ is s-finite} \right\}$  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  $m : X \rightrightarrows Y$ :

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

take Borel-Scott open maps as admissible monos

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

$\Rightarrow$  model axiomatic domain theory

$\Rightarrow$  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  
 $m : -_{\perp} \rightarrow T, \perp \mapsto 0$

## Derived axioms/structure

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

## Axioms

$(\dashv)$  every object has a partial  
 map classifier  $\downarrow_X : X \rightarrow X_{\perp}$   
 $(\text{fup})$  every admissible mono is full  
 and upper-closed  
 $(\dashv_{\leq})$   $\lfloor - \rfloor$  is locally monotone  
 $(\bigvee)$   $\mathfrak{D}$  is  $\omega\mathbf{Cpo}$ -enriched  
 $(U)$   $\omega$ -colimits behave uniformly  
 $(1)$   $\mathfrak{D}$  has a terminal object  
 $(\rightarrow_{\leq})$   $\mathfrak{D}$  has locally monotone  
 exponentials  
 $(+)$  locally continuous total  
 coproducts  
 $(?!)$   $0 \rightarrow 1$  is admissible  
 $(\times_{\vee})$   $\mathfrak{D}$  has a locally  
 continuous products  
 $(CL)$   $\mathfrak{D}$  is cocomplete  
 $(T_{\vee})$   $T$  is locally continuous

$(\otimes)$   $\mathbf{p}\mathfrak{D}$  has partial products  
 $(\otimes_{\vee})$   $(\otimes)$  is locally continuous  
 $(\rightarrow_{\vee})$   $\mathfrak{D}$  has locally continuous  
 exponentials  
 $(\Rightarrow_{\vee})$   $\mathbf{p}\mathfrak{D}$  has locally continuous  
 partial exponentials  
 $(\mathbf{p}CL)$   $\mathbf{p}\mathfrak{D}$  is cocomplete  
 $(\mathbf{p}+\vee)$   $\mathbf{p}\mathfrak{D}$  has locally continuous  
 partial coproducts  
 $(BC)$   $J : \hookrightarrow \mathbf{p}\mathfrak{D}$  is a bilimit  
 compact expansion

# 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  $\lambda$ -calculus

[Fiore-Plotkin'94, Fiore'96]


## This talk

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

## Also in the paper

- ▶ Axiomatic domain theory
- ▶ Operational semantics  
à la [Borgström et al.'16]
- ▶ Characterising  $\omega\mathbf{Qbs}$

# ProbProg: Important Language Features

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

