

Probabilistic Programming and a Domain-theoretic Approach to Skorohod's Theorem

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Workshop on Probabilistic Programming Semantics
Paris, Jussieu
January 17, 2017

Supported by US AFOSR

I. Stochastic Processes and Skorohod's Theorem

II. Domains and Probability Measures

III. Proving Skorohod's Theorem

Along the way, compare with the computational Polish space approach

Stochastic Processes and Skorohod's Theorem

A *stochastic process* is a time-indexed family $\{X_t \mid t \in T \subseteq \mathbb{R}_+\}$ of random variables / elements $X_t: \Omega \rightarrow S$, where $(\Omega, \Sigma_\Omega, \mu)$ is a probability space, and S is a Polish space.

Fact: If S is Polish, then so is $(\text{Prob}(S), d_p)$, where d_p is the *Prokhorov metric*.

Skorohod's Theorem

Let S be a Polish space, let $\nu \in \text{Prob } S$, and let λ denote Lebesgue measure on $[0, 1]$. Then there is a random variable $X: [0, 1] \rightarrow S$ satisfying $X_* \lambda = \nu$.

Moreover, if $\nu_n, \nu \in \text{Prob } S$ satisfy $\nu_n \rightarrow_w \nu$, then the random variables $X_n, X: [0, 1] \rightarrow S$ with $X_* \lambda = \nu, X_{n*} \lambda = \nu_n$ satisfy $X_n \rightarrow X$ λ -a.e.

Fact: We could use any *standard probability space* (S, Σ_S, m) instead of $([0, 1], \lambda)$.

Domains

Domains are partially ordered sets with additional properties;

Informatic partial order

$p \sqsubseteq q$ if q contains more information than p .

Example: Upper Space $U(X) = (\{K \subseteq X \mid \emptyset \neq K \text{ compact}\} \cup \{X\}, \supseteq)$.

Directed completeness

$\emptyset \neq D \subseteq P$ directed if $x, y \in D \Rightarrow (\exists z \in D) x, y \leq z$.

P directed complete: D directed $\Rightarrow \sup D$ exists.

$$\mathcal{F} \subseteq U(X) \text{ directed} \Rightarrow \sup \mathcal{F} = \bigcap \mathcal{F}.$$

Approximation

$x \ll y$ iff $y \leq \sup D \Rightarrow (\exists d \in D) x \leq d$.

Domain: $\downarrow y = \{x \mid x \ll y\}$ directed and $y = \sup \downarrow y$

$$K \ll L \text{ iff } L \subseteq K^\circ; \quad L = \bigcap \{K \mid L \subseteq K^\circ\} = \sup \{K \mid K \ll L\}.$$

Scott Topology

U Scott open if:

- $U = \uparrow U = \{x \in P \mid (\exists u \in U) u \leq x\}$ and
- D directed, $\sup D \in U \Rightarrow D \cap U \neq \emptyset$.

Morphisms

$f: P \rightarrow Q$ is Scott continuous if:

- f is monotone, and
- D directed $\Rightarrow f(\sup D) = \sup f(D)$.

$f: X \rightarrow Y \Rightarrow U(f): U(X) \rightarrow U(Y)$ by $U(f)(K) = f(K)$

is monotone and $f(\bigcap \mathcal{F}) = \bigcap f(\mathcal{F})$.

Lawson Topology

Basis: $\{\uparrow x \setminus \uparrow F \mid F \in \mathcal{P}_{<\omega} D\}$

Hausdorff refinement of Scott topology.

All the domains we discuss are Lawson compact.

Domain Environments

Embedding X in $U(\bar{X})$

X Polish $\Rightarrow X \hookrightarrow [0, 1]^\omega \Rightarrow X \subseteq_{G_\delta} \bar{X}$ compact Polish

$\{K_n \mid K_n \in U(\bar{X}), n > 0\}$: neighborhood basis of compact subsets of \bar{X} .

Then:

1° $X \subseteq \bar{X} \hookrightarrow \text{Max } U(\bar{X}) \subseteq U(\bar{X})$ by $x \mapsto \{x\} = \bigcap_n \{K_n \mid x \in K_n^\circ\}$

2° X inherits the Scott topology = Lawson topology on $\text{Max } U(\bar{X})$.

3° Each family $\{K_1, \dots, K_{n_m}\}$ with $\bar{X} \subseteq \bigcup_{n_m} K_i^\circ$ defines a

Scott-continuous projection $\psi_m: U(\bar{X}) \rightarrow U(\bar{X})$ with finite image.

4° Ordering the covers by refinement yields $\mathbf{1}_{U(\bar{X})} = \sup_m \psi_m$.

Domains and Probability Measures

Prob(D) is a Domain

D (Lawson compact) domain \Rightarrow Prob(D) (Lawson compact) domain:

- 1° $\mu \leq \nu$ iff $\int f d\mu \leq \int f d\nu$ ($\forall f: D \rightarrow \mathbb{R}_+$ Scott continuous)
- 2° D Lawson compact \Rightarrow (Prob(D), weak) = (Prob(D), Lawson).
- 3° $D = U(\overline{X})$, X Polish $\Rightarrow \mathbf{1}_{\text{Prob}(D)} = \sup_n \psi_{n*}$, so
 $\mu = \sup_n \psi_{n*} \mu$, with $\psi_{n*} \mu = \sum_{i \leq m_n} \mu(K_i) \delta_{K_i}$ ($\forall \mu$).
- 4° Choosing $r_i < \mu(K_i)$ & $L_i \lll K_i$ and $\epsilon = 1 - (\sum_i r_i)$ produces
 $\nu_n \stackrel{\text{def}}{=} \epsilon \delta_X + \sum_{i \leq m_n} r_i \delta_{L_i} \lll \psi_{n*} \mu$.
- 5° Construct $\nu_1 \lll \dots \lll \nu_n \lll \nu_{n+1} \lll \dots$ with $\mu = \sup_n \nu_n$.

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By comparison:

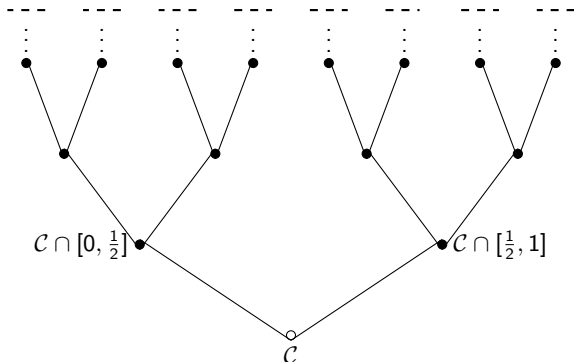
(X, S, d) a computational Polish space implies

$$\mu = \lim_n \sum_{i \leq m_n} d_i \delta_{x_i} \in \text{Max } U(\bar{X}) \text{ with } x_i \in S.$$

Then $\nu_n \ll \psi_{n*} \mu \leq \mu \Rightarrow (\exists N) n \geq N \Rightarrow \nu_n \ll \sum_{i \leq m_n} d_i \delta_{x_i}$.

Defining the Random Variables

A Domain Environment for \mathcal{C}



$$\mathcal{C} \hookrightarrow \text{Max CT} \subseteq \text{CT} ::= \{0, 1\}^* \cup \{0, 1\}^\omega$$

$$\mathcal{C}_n = \{0, 1\}^n \Rightarrow \exists \pi_n: \mathcal{C} \rightarrow \mathcal{C}_n \text{ retraction}$$

$$\mathbf{1}_{\mathcal{C}} = \sup_n \iota_n \circ \pi_n|_{\mathcal{C}}, \text{ where } \iota_n: \mathcal{C}_n \hookrightarrow \mathcal{C} \text{ lower semicontinuous.}$$

Defining the Random Variables

A Domain Environment for \mathcal{C}

- Given $\mu \in \text{Prob}U(\bar{X})$ and $\psi_{n*}\mu = \sum_{i \leq m_n} \mu(K_i)\delta_{K_i}$, fix p_n and choose $r_i \in D_{p_n} = \{\frac{s}{2^{p_n}} \mid 0 \leq s \leq 2^{p_n}\}$ and $L_i \ll K_i$.

- Then $\nu_n \stackrel{\text{def}}{=} (1 - \sum_i r_i)\delta_X + \sum_{i \leq m_n} r_i\delta_{L_i} \ll \sum_{i \leq m_n} \mu(K_i)\delta_{K_i}$.

- Define $f_n: \mathcal{C}_{p_n} \rightarrow U(\bar{X})$ by $f_n(j) = \begin{cases} L_1 & \text{if } 1 \leq j \leq r_1 \\ L_2 & \text{if } r_1 < j \leq r_1 + r_2 \\ \vdots & \\ X & \text{if } \sum_i r_i < j \end{cases}$

Then $f_{n*}(\frac{1}{2^{p_n}} \sum_{i \leq 2^{p_n}} \delta_{\frac{i}{2^{p_n}}}) = (1 - \sum_i r_i)\delta_X + \sum_{i \leq m_n} r_i\delta_{L_i} = \nu_n$.

- So, $f_n \circ \pi_n: \mathcal{C} \rightarrow U(\bar{X})$ is Lawson continuous and

$(f_n \circ \pi_n)_*(\mu_{\mathcal{C}}) = \nu_n = (1 - \sum_i r_i)\delta_X + \sum_{i \leq m_n} r_i\delta_{L_i}$.

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- By construction $f_n: \downarrow\mathcal{C}_{p_n} \rightarrow U(\overline{X})$ satisfies $f_n \leq f_{n+1}$, and then $X = (\sup_n f_n \circ \pi_n)|_{\mathcal{C}}: \mathcal{C} \rightarrow U(\overline{X})$ measurable with $X_*\mu_{\mathcal{C}} = \mu$.

Defining the Random Variables

A Domain Environment for \mathcal{C}

- By construction $f_n: \downarrow \mathcal{C}_{P_n} \rightarrow U(\overline{X})$ satisfies $f_n \leq f_{n+1}$, and then $X = (\sup_n f_n \circ \pi_n)|_{\mathcal{C}}: \mathcal{C} \rightarrow U(\overline{X})$ measurable with $X_* \mu_{\mathcal{C}} = \mu$.
- If $\mu_m \rightarrow_w \mu \in \text{Prob}(X)$, define $\rho_{m,n} \ll \psi_{n*} \mu_m$ and $f_{m,n}: \downarrow \mathcal{C}_{\rho_{m,n}} \rightarrow U(\overline{X})$ as above. Then $X_m = (\sup_n f_{m,n})|_{\mathcal{C}}$ satisfies $X_m: \mathcal{C} \rightarrow U(\overline{X})$ measurable with $X_{m*} \lambda = \mu_m$.
- Argue directly that $X_m \rightarrow X$ a.s. λ .

Questions?