

A probabilistic theory of designs based on distributions

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It's all about probability!

- Reliability of programs
- Programs run on fallible hardware
- Probabilistic specifications
- Security
- ...

Chen and Sanders @ FM09

They talk of a “so-far-unachieved goal of unifying probabilism with other programming constructs in the style of Unifying Theories of Programming” .

Main references

- Kozen's framework
- pGCL
- UTP — ... but I believe that on the second day of UTP 2012 there is no need to give a general overview of UTP. 😊

We got a paper into *iFM2012*, with the title *A UTP semantics of pGCL as a homogeneous relation*.

- Problem with the relational theory:

$$true; \mathcal{P} \neq true$$

- Solution:

$$ok \wedge Pre \Rightarrow ok' \wedge Post$$

or, for short,

$$Pre \vdash Post$$

Main elements

- States: $\sigma : \mathcal{V} \rightarrow \mathcal{W}$
- Distributions: $\chi : \mathcal{S} \rightarrow \mathbb{R}$
- Weighting distributions: $\pi : \mathcal{S} \rightarrow [0..1]$
- Probability distributions: $\delta : \mathcal{S} \rightarrow [0..1] \wedge \|\delta\| \leq 1$

where $\|\chi\| \triangleq \sum_{\sigma \in \text{dom}(\chi)} \chi(\sigma)$

Notable distributions:

- Point distribution: $\|\eta_\sigma\| = 1 \wedge \eta_\sigma(\sigma) = 1$
- Empty distribution: $\forall \sigma \bullet \epsilon(\sigma) = 0$
- Unit distribution: $\forall \sigma \bullet \iota(\sigma) = 1$

Operating on distributions

- Pointwise addition and multiplication
- Scaling
- Restriction
- Remap

Restriction

- Given a boolean condition c , the distribution $\chi\langle c \rangle$ is equal to the distribution χ purged of all states where c does not hold true.
- Given a distribution ξ , the distribution $\chi\langle \xi \rangle$ is equal to the pointwise multiplication $\chi \circ \xi$.

Sometimes it is more convenient to see a multiplication as a restriction and viceversa:

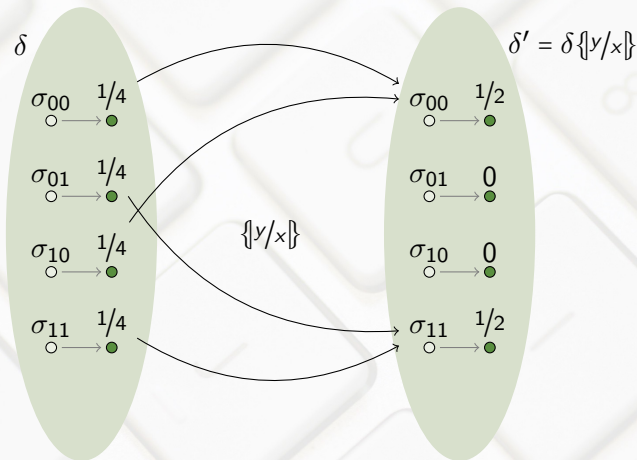
- $\chi\langle c \rangle = \chi \circ \iota\langle c \rangle$
- $\delta\langle \pi \rangle = \delta \circ \pi$

Remap

Given an expression \underline{e} and a probability distribution δ , the remap operation $\delta\{\underline{e}/\underline{v}\}$ is defined as follows:

$$(\delta\{\underline{e}/\underline{v}\})(\sigma') \triangleq (\sum \delta(\sigma) \mid \sigma' = \sigma \uparrow \{\underline{v} \mapsto \text{eval}_\sigma(\underline{e})\})$$

This is what happens when we perform the assignment $\underline{v} := \underline{e}$.



$$(\delta\{y/x\})(\sigma') \hat{=} (\sum \delta(\sigma) \mid \sigma' = \sigma \dagger \{x \mapsto \text{eval}_\sigma(y)\})$$

UTP semantics

<i>abort</i>	$\hat{=}$	$\ \delta'\ \leq \ \delta\ $
<i>miracle</i>	$\hat{=}$	$(\delta = \epsilon) \wedge (\delta' = \epsilon)$
<i>skip</i>	$\hat{=}$	$\delta' = \delta$
$\underline{v} := \underline{e}$	$\hat{=}$	$\delta' = \delta \{e/v\}$
$A; B$	$\hat{=}$	$\exists \delta_m \bullet A(\delta, \delta_m) \wedge B(\delta_m, \delta')$
<i>choice</i> (A, B, X)	$\hat{=}$	$\exists \pi, \delta_A, \delta_B \bullet \pi \in X \wedge A(\delta\{\pi\}, \delta_A) \wedge$ $B(\delta\{\bar{\pi}\}, \delta_B) \wedge \delta' = \delta_A + \delta_B$
$c * A$	$\hat{=}$	$\mu X \bullet \text{choice}(A; X, \text{skip}, \{\iota(c)\})$

The benefits of homogeneity

- we hide (some) complexity under the hood — we deal with objects which can be manipulated more easily at a higher level;
- easier to address theories based on homogeneous relations (before- and after-observations of the same type);
- sequential composition is straightforward.

We pay this with an increased complexity in some definitions and proofs, but we are rewarded with a more easily usable framework.

$$\mathit{choice}(A, B, \mathcal{X}) \triangleq \exists \pi, \delta_A, \delta_B \bullet \pi \in \mathcal{X} \wedge A(\delta\langle\pi\rangle, \delta_A) \wedge B(\delta\langle\bar{\pi}\rangle, \delta_B) \wedge \delta' = \delta_A + \delta_B$$

- for $\mathcal{X} = \{\iota\langle c \rangle\}$ we have conditional choice:

$$A \triangleleft c \triangleright B = \mathit{choice}(A, B, \{\iota\langle c \rangle\})$$

- for $\mathcal{X} = \{p \cdot \iota\}$ we have probabilistic choice:

$$A \text{ }_p\oplus\text{ } B = \mathit{choice}(A, B, \{p \cdot \iota\})$$

- for $\mathcal{X} = \mathcal{D}_w$ we have non-deterministic choice:

$$A \sqcap B = \mathit{choice}(A, B, \mathcal{D}_w)$$

- for $\mathcal{X} = \{\epsilon, \iota\}$ we have ordinary disjunction:

$$A \vee B = \mathit{choice}(A, B, \{\epsilon, \iota\})$$

Dist1 *feasibility*: $\|\delta'\| \leq \|\delta\|$

Dist2 *monotonicity*:

$$\mathcal{P}(\delta_1, \delta'_1) \wedge \mathcal{P}(\delta_2, \delta'_2) \wedge \delta_2 > \delta_1 \Rightarrow \delta'_2 \geq \delta'_1$$

Dist3 *scaling*:

$$\forall a \in \mathbb{R}^+ \wedge \|a \cdot \delta\| \leq 1 \bullet \mathcal{P}(\delta, \delta') \Leftrightarrow \mathcal{P}(a \cdot \delta, a \cdot \delta')$$

Dist4 *convexity*, consequence of the purely random non-deterministic model:

$$(\mathcal{P}_1 \sqcap \mathcal{P}_2)(\delta, \delta') \Rightarrow \delta' \geq \min(\mathcal{P}_1(\delta) \cup \mathcal{P}_2(\delta))$$

This poses restrictions on the space of possible program images, which is strictly a subset of $\wp \mathcal{D}$.

Given a standard design $Pre \vdash Post$ we can easily derive the corresponding probabilistic design with the observation that $ok \equiv (\|\delta\| = 1)$:

$$\begin{aligned}
 Pre \vdash Post &\equiv ok \wedge Pre \Rightarrow ok' \wedge Post \\
 &\equiv \|\delta\| = 1 \wedge Pre \Rightarrow \|\delta'\| = 1 \wedge Post \\
 &\equiv \|\delta\langle Pre \rangle\| = 1 \Rightarrow \|\delta'\langle Post \rangle\| = 1
 \end{aligned}$$

Assignment

$$\begin{aligned}
 \underline{v} := \underline{e} \hat{=} \text{true} &\vdash \delta' = \delta\{\underline{e}/\underline{v}\} \\
 &\equiv \text{ok} \wedge \text{true} \Rightarrow \text{ok}' \wedge \delta\{\underline{e}/\underline{v}\} \\
 &\equiv \|\delta\| = 1 \Rightarrow \|\delta'\| = 1 \wedge \delta' = \delta\{\underline{e}/\underline{v}\}
 \end{aligned}$$

This states that an assignment is a valid design only if the expression e is defined everywhere in the state space: in fact undefinedness of e causes $\delta\{\underline{e}/\underline{v}\}$ to be a sub-distribution and therefore $\underline{v} := e$ reduces to *false*.

Skip

$$\begin{aligned}
 \text{skip} &\hat{=} \text{true} \vdash \delta' = \delta \\
 &\equiv \text{ok} \wedge \text{true} \Rightarrow \text{ok}' \wedge \delta \\
 &\equiv \|\delta\| = 1 \Rightarrow \|\delta'\| = 1 \wedge \delta' = \delta \\
 &\equiv \|\delta\| = 1 \Rightarrow \delta' = \delta
 \end{aligned}$$

Abort

$$\begin{aligned} \text{abort} &\hat{=} \text{false} \vdash \text{false} \\ &\equiv \text{ok} \wedge \text{false} \Rightarrow \text{ok}' \wedge \text{false} \\ &\equiv \text{false} \Rightarrow \text{false} \\ &\equiv \text{true} \\ &\equiv \text{false} \Rightarrow \text{ok}' \wedge \text{true} \\ &\equiv \text{ok} \wedge \text{false} \Rightarrow \text{ok}' \wedge \text{true} \\ &\equiv \text{false} \vdash \text{true} \end{aligned}$$

Chaos

$$\begin{aligned}
 \text{chaos} &\equiv \text{true} \vdash \text{abort}_R \\
 &\equiv \text{ok} \wedge \text{true} \Rightarrow \text{ok}' \wedge \|\delta\| \leq \|\delta'\| \\
 &\equiv \|\delta\| = 1 \Rightarrow \|\delta'\| = 1 \wedge \|\delta\| \leq \|\delta'\| \\
 &\equiv \|\delta\| = 1 \Rightarrow \|\delta'\| = 1 \\
 &\equiv \text{ok} \Rightarrow \text{ok}' \\
 &\equiv \text{true} \vdash \text{true} \\
 &\triangleq \text{chaos}
 \end{aligned}$$

Miracle (I)

$$\textit{miracle} \triangleq \textit{true} \vdash \textit{false}$$

$$\equiv \textit{ok} \wedge \textit{true} \Rightarrow \textit{ok}' \wedge \textit{false}$$

$$\equiv \textit{ok} \Rightarrow \textit{false}$$

$$\equiv \neg \textit{ok}$$

$$\equiv \neg(\|\delta\| = 1)$$

$$\equiv \|\delta\| < 1$$

Miracle (II)

$$\begin{aligned}
 \text{miracle} &\equiv \text{true} \vdash \text{miracle}_R \\
 &\equiv \text{true} \vdash (\delta = \epsilon) \wedge (\delta' = \epsilon) \\
 &\equiv \text{ok} \wedge \text{true} \Rightarrow \text{ok}' \wedge (\delta = \epsilon) \wedge (\delta' = \epsilon) \\
 &\equiv \text{ok} \Rightarrow \text{ok} \wedge \text{ok}' \wedge (\delta = \epsilon) \wedge (\delta' = \epsilon) \\
 &\equiv \text{ok} \Rightarrow (\|\delta\| = 1) \wedge \text{ok}' \wedge (\delta = \epsilon) \wedge (\delta' = \epsilon) \\
 &\equiv \text{ok} \Rightarrow \text{false} \\
 &\equiv \neg \text{ok} \\
 &\equiv \neg (\|\delta\| = 1) \\
 &\equiv \|\delta\| < 1
 \end{aligned}$$

All healthiness conditions deriving from the distributional framework (Dist1–Dist4) obviously hold here as well; with a small modification we can recast the notion of total correctness by restricting Dist1 to a variant Dist1-TC (which implies Dist1), stating that:

$$\|\delta\| = \|\delta'\|$$

This requires a program to terminate with the same probability p with which it has started:

$$\|\delta\| = p \wedge \mathit{Pre} \Rightarrow \|\delta'\| = p \wedge \mathit{Post}$$

Standard designs have observations $o\kappa, o\kappa', \sigma, \sigma'$.

$$\begin{aligned}o\kappa, o\kappa' &: \mathbb{B} \\ \sigma, \sigma' &: \mathcal{S}\end{aligned}$$

Probabilistic designs have observations δ, δ'

$$\delta, \delta' : \mathcal{S} \rightarrow [0, 1]$$

Informally we require the two approaches to yield the same results when we are dealing with point distributions, *i.e.* when the probability of being in a given state is 1.

We define the linking predicate L as:

$$L((\delta, \delta'), (\sigma, \sigma', o\mathcal{K}, o\mathcal{K}')) \triangleq o\mathcal{K} \Leftrightarrow (\|\delta'\| = 1) \wedge o\mathcal{K}' \Leftrightarrow (\|\delta'\| = 1) \\ \wedge \delta = \eta_\sigma \wedge \delta' = \eta_{\sigma'}$$

This linking predicate allows us to introduce the following *Galois connections*; first we define the weakest probabilistic design corresponding to a standard design \mathcal{P}_S :

$$\forall \sigma, \sigma', o\mathcal{K}, o\mathcal{K}' \bullet L((\delta, \delta'), (\sigma, \sigma', o\mathcal{K}, o\mathcal{K}')) \Rightarrow \mathcal{P}_S(\sigma, \sigma', o\mathcal{K}, o\mathcal{K}')$$

Analogously, the strongest standard design corresponding to a probabilistic design \mathcal{P}_D is:

$$\exists \delta, \delta' \bullet L((\delta, \delta'), (\sigma, \sigma', o\mathcal{K}, o\mathcal{K}')) \wedge \mathcal{P}_D(\delta, \delta')$$

This linking predicate is a bit too strong, as it maps many interesting program constructs to *abort*.

So far we have seen standard designs as relations:

$$\mathcal{P}_S : \mathcal{S} \times \mathbb{B} \rightarrow \mathcal{S} \times \mathbb{B}$$

but in order to build a more useful link we turn to this other interpretation:

$$\mathcal{P}_{\rho S} : \mathcal{S} \times \mathbb{B} \rightarrow \rho\mathcal{S} \times \mathbb{B}$$

which maps a state to what we may term its program image $\mathcal{P}(\sigma)$:

$$\mathcal{P}(\sigma) = \{\sigma' \mid \mathcal{P}_S(\sigma, \sigma')\}$$

This allows us to define the following linking predicate:

$$L_{\wp}((\delta, \delta'), (\sigma, \alpha', o\mathcal{K}, o\mathcal{K}')) \hat{=} o\mathcal{K} \Leftrightarrow (\|\delta'\| = 1) \wedge o\mathcal{K}' \Leftrightarrow (\|\delta'\| = 1) \\ \wedge \delta = \eta_{\sigma} \wedge \text{supp}(\delta') = \alpha'$$

We can state the variants of the previous Galois connections as:

$$\forall \sigma, \alpha', o\mathcal{K}, o\mathcal{K}' \bullet L_{\wp}((\delta, \delta'), (\sigma, \alpha', o\mathcal{K}, o\mathcal{K}')) \Rightarrow \mathcal{P}_{\wp S}(\sigma, \alpha', o\mathcal{K}, o\mathcal{K}') \\ \exists \delta, \delta' \bullet L_{\wp}((\delta, \delta'), (\sigma, \alpha', o\mathcal{K}, o\mathcal{K}')) \wedge \mathcal{P}_D(\delta, \delta')$$

This link maps all constructs from the probabilistic theory to the correspondent ones in the standard theory of designs, with the exception of generic choice, which is mapped to the nondeterministic choice.

Strongest standard design derivation:

$$\begin{aligned}
 & \exists \delta, \delta' \bullet L_{\wp}((\delta, \delta'), (\sigma, \alpha', \mathit{ok}, \mathit{ok}')) \wedge \mathcal{P}_D(\delta, \delta') \\
 \equiv & \exists \delta, \delta' \bullet \mathit{ok} \Leftrightarrow (\|\delta'\| = 1) \wedge \mathit{ok}' \Leftrightarrow (\|\delta'\| = 1) \wedge \delta = \eta_{\sigma} \wedge \text{supp}(\delta') = \alpha' \\
 & \wedge \mathcal{P}_D(\delta, \delta') \\
 \equiv & \exists \delta, \delta' \bullet \mathit{ok} \Leftrightarrow (\|\delta'\| = 1) \wedge \mathit{ok}' \Leftrightarrow (\|\delta'\| = 1) \wedge \delta = \eta_{\sigma} \wedge \text{supp}(\delta') = \alpha' \\
 & \wedge ((\|\delta\| = 1) \Rightarrow \|\delta'\| = 1 \wedge \mathcal{P}_{DR}(\delta, \delta')) \\
 \equiv & \exists \delta, \delta' \bullet \delta = \eta_{\sigma} \wedge \text{supp}(\delta') = \alpha' \wedge (\mathit{ok} \Rightarrow \mathit{ok}' \wedge \mathcal{P}_{DR}(\delta, \delta')) \\
 \equiv & \exists \delta' \bullet \text{supp}(\delta') = \alpha' \wedge (\mathit{ok} \Rightarrow \mathit{ok}' \wedge \mathcal{P}_{DR}(\eta_{\sigma}, \delta')) \\
 \equiv & \mathit{ok} \Rightarrow \mathit{ok}' \wedge \mathcal{P}_{\wp SR}(\sigma, \alpha') \\
 \equiv & \mathcal{P}_{\wp S}(\sigma, \alpha')
 \end{aligned}$$

We have talked about:

- probabilistic theory of designs, which relies on a UTP-style framework based on distributions over the state space;
- embedding of the standard UTP theory, by requiring guaranteed termination from all program constructs;
- different Galois connections for projections/retreats with different characteristics.

Thanks for your attention!

Any questions?