## Proving Uniformity and Independence

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## A puzzle

A random walk on a cycle

- Start at position $s \in\{0,1, \ldots, n-1\}$
- Each iteration, flip a fair coin
- Heads: increment position (modulo $n$ )
- Tails decrement position (modulo $n$ )
- Return: last edge $(r, r+1)$ to be traversed


## A question

## What is the distribution of the returned edge, and how does it depend on the starting position s?

## A puzzle



## A puzzle



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Somewhat surprisingly
Distribution of final edge is uniform: Starting position $s$ doesn't matter!

## Basic properties of probabilistic programs

Uniformity of a variable $X$
For any two values $w, v$ in the (finite) range of $X$, we have:

$$
\operatorname{Pr}[X=w]=\operatorname{Pr}[X=v]
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in output distribution.

## Can be quite subtle to verify!

The idea today

# Use logic for relational verification to verify uniformity and independence 

A crash course:
the relational logic pRHL

## A curious program logic: pRHL [Barthe, Grégoire, Zanella-Béguelin]

pWhile: An imperative language with random sampling
$c::=x \leftarrow e \mid x \leftrightarrow$ flip $(p) \mid$ if $e$ then $c$ else $c \mid$ while $e$ do $c \mid$ skip $\mid c ; c$

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## A curious program logic: pRHL [Bathe, Grégoire, Zanella-Béguelin]

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pRHL is a program logic that is:

- Probabilistic: Programs can draw samples
- Relational: Describe executions of two programs


## Judgments in pRHL

$\{P($ in $\langle 1\rangle$, in $\langle 2\rangle)\} c \sim c^{\prime}\{Q($ out $\langle 1\rangle$, out $\langle 2\rangle)\}$

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## Assertions

- Non-probabilistic
- FO formulas over program variables tagged with $\langle 1\rangle$ or $\langle 2\rangle$


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Deep connection to probabilistic couplings

- Proofs specify how to correlate random samplings in runs
- Reduce sources of randomness, simplify verification


## For our purposes today: equality of distributions

If this is provable:

$$
\vdash\{P\} c \sim c^{\prime} \quad\left\{e\langle 1\rangle=e^{\prime}\langle 2\rangle\right\}
$$

## Then:

On any two input memories related by $P$, the distribution of $e$ in the first output is equal to the distribution of $e^{\prime}$ in the second output.

## In particular: express equality of probabilities

If this is provable for booleans $b, b^{\prime}$ :

$$
\vdash\{P\} \quad c \sim c^{\prime}\left\{b\langle 1\rangle=b^{\prime}\langle 2\rangle\right\}
$$

## Then:

On any two input memories related by $P$, the probability of $b$ in the first output is equal to the probability of $b^{\prime}$ in the second output.

## Random sampling rules in pRHL

Simplified version

FLIPEQ

$$
\vdash\{\top\} x \ll \mathbb{f l i p}(p) \sim x^{\prime} \mathbb{\&} \operatorname{flip}(p)\left\{x\langle 1\rangle=x^{\prime}\langle 2\rangle\right\}
$$

FLIPNEG

$$
\overline{\vdash\{T\} x \& \in \operatorname{fip}(p) \sim x^{\prime} \& \operatorname{flip}(1-p)\left\{x\langle 1\rangle=\neg x^{\prime}\langle 2\rangle\right\}}
$$

## Random sampling rules in pRHL

Simplified version

FLIPEQ $\underset{\vdash\{T\} x \& \operatorname{slip}(p) \sim x^{\prime} \& \operatorname{flip}(p)\left\{x\langle 1\rangle=x^{\prime}\langle 2\rangle\right\}}{\digamma}$
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Reading: for any $p \in[0,1]$,

1. [FLIPEQ]: Distributions of $\operatorname{flip}(p)$ and $\operatorname{flip}(p)$ are equal
2. [FLIPNEG]: Distributions of $\operatorname{flip}(p)$ and negated $\operatorname{flip}(1-p)$ are equal

## Rest of rules are standard ( $\approx$ Hoare logic)

Assignments
AsSN $\overline{\vdash\left\{Q\left[e\langle 1\rangle, e^{\prime}\langle 2\rangle / x\langle 1\rangle, x^{\prime}\langle 2\rangle\right]\right\} \quad x \leftarrow e_{1} \sim x^{\prime} \leftarrow e_{2}\{Q\}}$

Sequencing

$$
\text { SEQ } \frac{\vdash\{P\} c_{1} \sim c_{1}^{\prime}\{Q\} \quad \vdash\{Q\} c_{2} \sim c_{2}^{\prime}\{R\}}{\vdash\{P\} c_{1} ; c_{2} \sim c_{1}^{\prime} ; c_{2}^{\prime}\{R\}}
$$

Loops
While $\frac{\vdash\{P \wedge b\langle 1\rangle\} c \sim c^{\prime}\{P\} \quad \mid=P \Longrightarrow b\langle 1\rangle=b^{\prime}\langle 2\rangle}{\vdash\{P\} \text { while } b \text { do } c \sim \text { while } b^{\prime} \text { do } c^{\prime}\{P \wedge \neg b\langle 1\rangle\}}$

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Loops
While $\frac{\vdash\{P \wedge b\langle 1\rangle\} c \sim c^{\prime}\{P\} \quad \models P \Longrightarrow b\langle 1\rangle=b^{\prime}\langle 2\rangle}{\vdash\{P\} \text { while } b \text { do } c \sim \text { while } b^{\prime} \text { do } c^{\prime}\{P \wedge \neg b\langle 1\rangle\}}$

## Benefits of pRHL

## Probabilistic properties without probabilistic reasoning

- Abstract away all probabilities
- All reasoning is about relation between samples

Highly similar to Hoare logic

- Most things "just work"
- Compositional reasoning


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Apply to non-relational properties, like uniformity and independence.

## Verifying uniformity: <br> simulating a fair coin

## The algorithm

## Goal

Generate one fair coin flip, using only coin flips with a fixed bias $p \in(0,1)$.

## Procedure

1. Flip two coins with bias $p$
2. Re-flip as long as they are equal
3. Return the first coin flip the first time they are different

## In code

Consider the program fair:

$$
\begin{aligned}
& x \leftarrow t t ; \\
& y \leftarrow t t ; \\
& \text { while } x=y \text { do } \\
& \qquad x \& \mathbb{\&} \operatorname{flip}(p) ; \\
& y \& \in \operatorname{flip}(p) ; \\
& \text { return }(x)
\end{aligned}
$$

To show: generates fair coin flip

## Distribution of return value is uniform

## Observation: uniformity can be proved in pRHL

For every two booleans $w, v$, show:

$$
\vdash\{p\langle 1\rangle=p\langle 2\rangle\} \text { fair } \sim \text { fair }\{(x\langle 1\rangle=w) \Longleftrightarrow(x\langle 2\rangle=v)\}
$$

Reading: for every two booleans $w, v$,

$$
\operatorname{Pr}[x=w]=\operatorname{Pr}[x=v] \quad \text { in the output of fair. }
$$

Four choices in all for $w, v$

- We show the cases with $w \neq v$


## Step 1: rearrange program

Two equivalent programs: fair and fair ${ }^{\prime}$

$$
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& y \leftarrow t t ; \\
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& y \leftarrow t t ; \\
& \text { while } x=y \text { do } \\
& \qquad x \stackrel{\mathbb{s}}{=} \operatorname{flip}(p) ; \\
& y \& \operatorname{flip}^{2}(p) ; \\
& \operatorname{return}(x)
\end{aligned}
$$

For the cases $w \neq v$, suffices to show:

$$
\vdash\{p\langle 1\rangle=p\langle 2\rangle\} \text { fair } \sim \text { fair }^{\prime}\{x\langle 1\rangle=\neg x\langle 2\rangle\}
$$

## Step 2: apply the loop rule

while $x=y$ do<br>$x \stackrel{s}{\leftarrow} \operatorname{flip}(p)$;<br>$y \stackrel{\&}{ } \operatorname{flip}^{(p)}$;<br>return $(x)$

while $x=y$ do

$$
\begin{aligned}
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& x \& \operatorname{flip}(p) ; \\
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$$
\begin{aligned}
& \text { while } x=y \text { do } \\
& \qquad x \& \mathbb{S}^{\mathbb{S}} \operatorname{flip}(p) ; \\
& y \& \operatorname{flip}(p) ; \\
& \operatorname{return}(x)
\end{aligned}
$$

In the body: apply [FLIPEQ] for both pairs of samples

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& \text { while } x=y \text { do } \\
& \qquad x \& \operatorname{flip}(p) ; \\
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$$

$$
\text { while } x=y \text { do }
$$

$$
y \stackrel{\mathbb{s}}{ } \operatorname{flip}(p) ;
$$

$$
x \stackrel{\mathbb{S}}{\mathscr{S}} \text { fip }(p) ;
$$

$$
\text { return }(x)
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- We have: $x\langle 1\rangle=y\langle 2\rangle$


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- And: $x\langle 2\rangle=y\langle 1\rangle$


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In the body: apply [FLIPEQ] for both pairs of samples

- We have: $x\langle 1\rangle=y\langle 2\rangle$
- And: $x\langle 2\rangle=y\langle 1\rangle$

Establishes main invariant:

$$
x\langle 2\rangle=(\text { if } x\langle 1\rangle=y\langle 1\rangle \text { then } y\langle 2\rangle \text { else } \neg x\langle 1\rangle)
$$

## Step 3: putting it all together

Applying [Assn], [Sea] shows:

$$
\vdash\{p\langle 1\rangle=p\langle 2\rangle\} \text { fair } \sim \text { fair }\{(x\langle 1\rangle=w) \Longleftrightarrow(x\langle 2\rangle=v)\}
$$

when $w \neq v$; can also show same judgment when $w=v$. Conclude

## fair returns a uniform boolean

Extensions:
verifying independence

## Verifying independence: the easier way

Observation: reduce independence to uniformity

# $(x, y)$ is uniform over pairs <br> $x$ and $y$ are independent 

Limitation

- Only can show independence for uniform variables


## Verifying independence: the harder way

Use self-composition

- Let $c[1], c[2]$ be two copies of $c$ with disjoint variables
- Prove a pRHL judgment relating

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c \sim c[1] ; c[2]
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Independence of two variables $X, Y$
For any two values $w, v$, we have:

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in output distribution.
Benefits

- Can prove independence for non-uniform variables
- Similar ideas can cover conditional independence


## Summing up

## See the paper for

Lots more examples

- Cycle random walk
- Pairwise and $k$-wise independence
- Bayesian network
- Ballot theorem

Details about the implementation

- Most examples formalized in EasyCrypt framework

Future directions

- Automate this approach
- Explore relational verification for non-relational properties
- Integrate with more general probabilistic verification tools


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