

# Mechanizing the Probabilistic Guarded Command Language

Joe Hurd

Computing Laboratory  
University of Oxford

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Joint work with Carroll Morgan (UNSW), Annabelle McIver (Macquarie),  
Orieta Celiku (CMU) and Aaron Coble (Cambridge)

# Talk Plan

- 1 Introduction
- 2 Formalizing pGCL
- 3 Verification Conditions
- 4 Current Work
- 5 Summary

# Probabilistic Programs

Giving programs access to a random number generator is useful for many applications:

- Symmetry breaking
  - Rabin's mutual exclusion algorithm
- Eliminating pathological cases
  - Randomized quicksort
- Gain in (best known?) theoretical complexity
  - Sorting nuts and bolts
- Solving a problem in an extremely simple way
  - Finding minimal cuts

**Research goal:** Apply formal methods to programs with probabilistic nondeterminism.

# Probabilistic Guarded Command Language

- pGCL stands for Probabilistic Guarded Command Language.
- It's Dijkstra's GCL extended with probabilistic choice

$$c_1 \oplus_p c_2$$

- Like GCL, the semantics is based on weakest preconditions.
- **Important:** retains nondeterministic choice

$$c_1 \sqcap c_2$$

- Developed by Morgan, McIver et al. in Oxford and then Sydney, 1994–

# The HOL4 Theorem Prover

- Developed by Mike Gordon's Hardware Verification Group in Cambridge, first release was HOL88.
- Latest release called HOL4, developed jointly by Cambridge, Utah and ANU.
- Implements classical Higher Order Logic (a.k.a. simple type theory).
- Sprung from the Edinburgh LCF project, so has a small logical kernel to ensure soundness.

# Motivation

## Why formalize?

- The theoretical results and program algebra are checked by logically deriving them from a simple set of definitions.
  - Example: Deriving the rules of Floyd-Hoare logic from a denotational semantics.
- When the program algebra is mechanized its feasibility can be checked by directly applying it to example programs.
  - Analysis tools that deduce from the semantics can be used to check other tools or generate test vectors.

# pGCL Semantics

- Given a standard GCL program  $C$  and a postcondition  $Q$ , let  $P$  be the weakest precondition that satisfies

$$[P]C[Q]$$

- Precondition  $P$  is weaker than  $P'$  if  $P' \implies P$ .
- Think of the program  $C$  as a function that transforms postconditions into weakest preconditions.
- pGCL generalizes this to probabilistic programs:
  - Conditions  $\alpha \rightarrow \mathbb{B}$  become *expectations*  $\alpha \rightarrow [0, +\infty]$ .
  - Expectation  $P$  is weaker than  $P'$  if  $P' \sqsubseteq P$ .
  - Think of programs as *expectation transformers*.

# Expectations

- Expectations are reward functions, from states to expected rewards.
- Modelled in HOL as functions  $\alpha \rightarrow [0, +\infty]$ .
- Define the following operations on expectations:
  - $\text{Min } e_1 \ e_2 \equiv \lambda s. \min (e_1 \ s) (e_2 \ s)$
  - $e_1 \sqsubseteq e_2 \equiv \forall s. e_1 \ s \leq e_2 \ s$
  - $\text{Cond } b \ e_1 \ e_2 \equiv \lambda s. \text{if } b \ s \text{ then } e_1 \ s \text{ else } e_2 \ s$
  - $\text{Lin } p \ e_1 \ e_2 \equiv \lambda s. p \ s \times e_1 \ s + (1 - p \ s) \times e_2 \ s$



# Expectation Transformers

- Expectation transformers are functions from expectations to expectations.
- Expectation transformers that correspond to probabilistic programs satisfy healthiness conditions:

$$\begin{aligned}
 \text{feasible } t &\equiv t \text{ Zero} = \text{Zero} \\
 \text{monotonic } t &\equiv \forall e_1, e_2. e_1 \sqsubseteq e_2 \implies t e_1 \sqsubseteq t e_2 \\
 \text{scaling } t &\equiv \forall e, c. t (\lambda s. c \times e s) = \lambda s. c \times t e s \\
 \text{subadditive } t &\equiv \forall e_1, e_2. t (\lambda s. e_1 s + e_2 s) \sqsubseteq \lambda s. t e_1 s + t e_2 s \\
 \text{subtractive } t &\equiv \forall e, c. c \neq \infty \implies t (\lambda s. e s - c) \sqsubseteq \lambda s. t e s - c
 \end{aligned}$$

- Expectations form a lattice, so expectation transformers can be up\_continuous, have least and greatest fixed points, etc.

# Healthiness Condition

- The definition of healthiness for expectation transformers is analogous to healthiness of predicate transformers in standard GCL:

$$\text{healthy } t \equiv \text{feasible } t \wedge \text{sublinear } t \wedge \text{up\_continuous } t$$

where

$$\text{sublinear } t \equiv \text{scaling } t \wedge \text{subadditive } t \wedge \text{subtractive } t$$

- Sublinearity in pGCL is the generalization of the conjunctivity condition in GCL.

# States

- Fix states to be mappings from variable names to integers:

$$\text{state} \equiv \text{string} \rightarrow \mathbb{Z}$$

- For convenience, define a state update function:

$$\text{assign } v \ f \ s \equiv \lambda w. \text{ if } v = w \text{ then } f \ s \text{ else } s \ w$$

# pGCL Commands

Model pGCL commands with a HOL datatype:

```
command ≡ Abort
         | Skip
         | Assign of string × (state → ℤ)
         | Seq of command × command
         | Nondet of command × command
         | Prob of (state → [0, 1]) × command × command
         | While of (state → ℬ) × command
```

**Note:** The probability in Prob can depend on the state.

# Derived Commands

Define all other commands as syntactic sugar:

$$\begin{aligned}
 v := f &\equiv \text{Assign } v \ f \\
 c_1 ; c_2 &\equiv \text{Seq } c_1 \ c_2 \\
 c_1 \sqcap c_2 &\equiv \text{Nondet } c_1 \ c_2 \\
 c_1 \ p \oplus \ c_2 &\equiv \text{Prob } (\lambda s. \ p) \ c_1 \ c_2 \\
 \text{if } b \text{ then } c_1 \ \text{else } c_2 &\equiv \text{Prob } (\lambda s. \ \text{if } b \ s \ \text{then } 1 \ \text{else } 0) \ c_1 \ c_2 \\
 v := \{e_1, \dots, e_n\} &\equiv v := e_1 \sqcap \dots \sqcap v := e_n \\
 v := \langle e_1, \dots, e_n \rangle &\equiv v := e_1 \ 1/n \oplus v := \langle e_2, \dots, e_n \rangle \\
 b_1 \rightarrow c_1 \mid \dots \mid b_n \rightarrow c_n &\equiv \\
 \begin{cases} \text{Abort} & \text{if none of the } b_i \text{ hold on the current state} \\ \prod_{i \in I} c_i & \text{where } I = \{i \mid 1 \leq i \leq n \wedge b_i \text{ holds}\} \end{cases}
 \end{aligned}$$

# Weakest Preconditions

Define weakest preconditions (wp) directly on commands:

- $\vdash (\text{wp } \text{Abort} = \lambda e. \text{Zero})$
- $\wedge (\text{wp } \text{Skip} = \lambda e. e)$
- $\wedge (\text{wp } (\text{Assign } v f) = \lambda e, s. e (\text{assign } v f s))$
- $\wedge (\text{wp } (\text{Seq } c_1 c_2) = \lambda e. \text{wp } c_1 (\text{wp } c_2 e))$
- $\wedge (\text{wp } (\text{Nondet } c_1 c_2) = \lambda e. \text{Min } (\text{wp } c_1 e) (\text{wp } c_2 e))$
- $\wedge (\text{wp } (\text{Prob } p c_1 c_2) = \lambda e. \text{Lin } p (\text{wp } c_1 e) (\text{wp } c_2 e))$
- $\wedge (\text{wp } (\text{While } b c) = \lambda e. \text{Ifp } (\lambda e'. \text{Cond } b (\text{wp } c e') e))$

# Commands are Healthy

- The major theorem of our formalization:

$$\vdash \forall c. \text{healthy } (\text{wp } c)$$

- Proof by structural induction (800 lines of HOL4 script).
- The hardest part was sublinearity of while loops.
- Needed several lemmas, for example:

$$\begin{aligned} &\vdash \forall t, e_1, e_2. \\ &\quad \text{healthy } t \wedge \text{bounded } t \wedge e_2 \sqsubseteq e_1 \implies \\ &\quad t (\lambda s. e_1 s - e_2 s) \sqsubseteq \lambda s. t e_1 s - t e_2 s \end{aligned}$$

# Example: Monty Hall

contestant *switch*  $\equiv$

$pc := \{1, 2, 3\};$

$cc := \langle 1, 2, 3 \rangle;$

$pc \neq 1 \wedge cc \neq 1 \rightarrow ac := 1$

|  $pc \neq 2 \wedge cc \neq 2 \rightarrow ac := 2$

|  $pc \neq 3 \wedge cc \neq 3 \rightarrow ac := 3;$

if  $\neg$ *switch* then Skip else

$cc :=$  (if  $cc \neq 1 \wedge ac \neq 1$  then 1

else if  $cc \neq 2 \wedge ac \neq 2$  then 2 else 3)

The postcondition is simply the desired goal of the contestant, i.e.,

$win \equiv$  if  $cc = pc$  then 1 else 0.



# Example: Monty Hall

- Verification proceeds by:
  - ① Rewriting away all the syntactic sugar.
  - ② Expanding the definition of  $wp$ .
  - ③ Carrying out the numerical calculations.
- After 22 seconds and 250536 primitive inferences in the logical kernel:

$\vdash wp(\text{contestant } switch) \text{ win} = \lambda s. \text{ if } switch \text{ then } 2/3 \text{ else } 1/3$

- In other words, by switching the contestant is twice as likely to win the prize.
- Not trivial to do by hand, because the intermediate expectations get rather large.

# Weakest Liberal Preconditions

Weakest liberal preconditions (wlp) model partial correctness.

- $\vdash$  (wlp Abort =  $\lambda e.$  **Infty**)
- $\wedge$  (wlp Skip =  $\lambda e.$   $e$ )
- $\wedge$  (wlp (Assign  $v f$ ) =  $\lambda e, s.$   $e$  (assign  $v f s$ ))
- $\wedge$  (wlp (Seq  $c_1 c_2$ ) =  $\lambda e.$  wlp  $c_1$  (wlp  $c_2 e$ ))
- $\wedge$  (wlp (Nondet  $c_1 c_2$ ) =  $\lambda e.$  Min (wlp  $c_1 e$ ) (wlp  $c_2 e$ ))
- $\wedge$  (wlp (Prob  $p c_1 c_2$ ) =  $\lambda e.$  Lin  $p$  (wlp  $c_1 e$ ) (wlp  $c_2 e$ ))
- $\wedge$  (wlp (While  $b c$ ) =  $\lambda e.$  **gfp** ( $\lambda e'.$  Cond  $b$  (wlp  $c e'$ )  $e$ ))

# Weakest Liberal Preconditions: Example

- Consider the simplest infinite loop:

$$\text{loop} \equiv \text{While } (\lambda s. \top) \text{ Skip}$$

- For any postcondition  $post$ , we have

$$\vdash \text{wp loop } post = \text{Zero} \wedge \text{wlp loop } post = \text{Infty}$$

- These correspond to the total and partial Hoare triples

$$[\perp] \text{ loop } [post] \quad \{\top\} \text{ loop } \{post\}$$

as we would expect from an infinite loop.

# Calculating wlp Lower Bounds

- Suppose we have a pGCL command  $c$  and a postcondition  $q$ .
- We wish to derive a lower bound on the weakest liberal precondition.
  - In general, programs are shown to have desirable properties by proving *lower bounds*.
  - Example:  $(\lambda s. 0.95) \sqsubseteq \text{wlp prog (if ok then 1 else 0)}$
- Can think of this as the query  $P \sqsubseteq \text{wlp } c \ q$ .
- **Idea:** use a Prolog interpreter to solve for the variable  $P$ .

# Calculating wlp: Rules

Simple rules:

- $\text{Infty} \sqsubseteq \text{wlp Abort } Q$
- $Q \sqsubseteq \text{wlp Skip } Q$
- $R \sqsubseteq \text{wlp } C_2 Q \wedge P \sqsubseteq \text{wlp } C_1 R$   
 $\implies$   
 $P \sqsubseteq \text{wlp (Seq } C_1 C_2) Q$

**Note:** the Prolog interpreter automatically calculates the ‘middle condition’ in a Seq command.

# Calculating wlp: While Loops

- Define an assertion command:  $\text{Assert } p \ c \equiv c$ .
- Provide a while rule that requires an assertion:
  - $R \sqsubseteq \text{wlp } C \ P \ \wedge \ P \sqsubseteq \text{Cond } B \ R \ Q$   
 $\implies$   
 $P \sqsubseteq \text{wlp } (\text{Assert } P \ (\text{While } B \ C)) \ Q$
- The second premise generates a *verification condition* as an extra subgoal.
- It is left to the user to provide a useful loop invariant in the Assert around the while loop.

# Rabin's Mutual Exclusion Algorithm

- Suppose  $N$  processors are executing concurrently, and from time to time some of them need to enter a critical section of code.
- The mutual exclusion algorithm of Rabin (1982, 1992) works by electing a leader who is permitted to enter the critical section:
  - ① Each of the waiting processors repeatedly tosses a fair coin until a head is shown
  - ② The processor that required the largest number of tosses wins the election.
  - ③ If there is a tie, then have another election.
- Could implement the coin tossing using
$$n := 0 ; b := 0 ; \text{While } (b = 0) (n := n + 1 ; b := \langle 0, 1 \rangle)$$

# Rabin's Mutual Exclusion Algorithm

For our verification, we do not model  $N$  processors concurrently executing the above voting scheme, but rather the following data refinement of that system:

- 1 Initialize  $i$  with the number of processors waiting to enter the critical section who have just picked a number.
- 2 Initialize  $n$  with 1, the lowest number not yet considered.
- 3 If  $i = 1$  then we have a unique winner: return `SUCCESS`.
- 4 If  $i = 0$  then the election has failed: return `FAILURE`.
- 5 Reduce  $i$  by eliminating all the processors who picked the lowest number  $n$  (since certainly none of them won the election).
- 6 Increment  $n$  by 1, and jump to Step 3.



# Rabin's Mutual Exclusion Algorithm

The following pGCL program implements this data refinement:

$$\begin{aligned} \text{rabin} \quad \equiv \quad & \text{While } (1 < i) ( \\ & \quad n := i ; \\ & \quad \text{While } (0 < n) \\ & \quad \quad (d := \langle 0, 1 \rangle ; i := i - d ; n := n - 1) \\ & ) \end{aligned}$$

The desired postcondition representing a unique winner of the election is

$$\text{post} \quad \equiv \quad \text{if } i = 1 \text{ then } 1 \text{ else } 0$$

# Rabin's Mutual Exclusion Algorithm

- The precondition that we aim to show is

$$pre \equiv \text{if } i = 1 \text{ then } 1 \text{ else if } 1 < i \text{ then } 2/3 \text{ else } 0$$

*“For any positive number of processors wanting to enter the critical section, the probability that the voting scheme will produce a unique winner is 2/3, except for the trivial case of one processor when it will always succeed.”*

- **Surprising:** The probability of success is independent of the number of processors.
- We formally verify the following statement of partial correctness:

$$pre \sqsubseteq \text{wlp rabin } post$$

# Rabin's Mutual Exclusion Algorithm

- Need to annotate the While loops with invariants.
- The invariant for the outer loop is simply  $pre$ .
- For the inner loop we used

if  $0 \leq n \leq i$  then  $2/3 \times \text{invar1 } i \ n + \text{invar2 } i \ n$  else 0

where

$\text{invar1 } i \ n \equiv$

$1 - (\text{if } i = n \text{ then } (n + 1)/2^n \text{ else if } i = n + 1 \text{ then } 1/2^n \text{ else } 0)$

$\text{invar2 } i \ n \equiv \text{if } i = n \text{ then } n/2^n \text{ else if } i = n + 1 \text{ then } 1/2^n \text{ else } 0$

- Coming up with these was the hardest part of the verification.

# Rabin's Mutual Exclusion Algorithm

The verification proceeded as follows:

- 1 Annotate the program to create the goal:

$$pre \sqsubseteq wlp \text{ annotated\_rabin } post$$

- 2 This is now in the correct form to apply the VC generator.
- 3 Finish off the VCs with 58 lines of HOL-4 proof script.

```
|- Leq (\s. if s"i" = 1 then 1
         else if 1 < s"i" then 2/3 else 0)
      (wlp rabin (\s. if s"i" = 1 then 1 else 0))
```

# Relational Semantics

- This formalization started from the weakest precondition semantics of pGCL programs.
- Instead can derive this from a relational semantics between initial states and probability distributions over final states:

$$\alpha \times (\alpha \rightarrow [0, 1]) \rightarrow \mathbb{B}$$

- Formalizing this would verify the connection between pGCL expectations and probability theory expectations.

# Loop Rules

- Practical program analysis tools need robust ways of reasoning about programs with loops.
- The usual slogan

total correctness = partial correctness + termination

doesn't hold for (this formalization of) pGCL!

- Counterexample verified in HOL4:

$$\vdash \text{wlp } (\text{While } (n = 0) (n := \langle 0, 1 \rangle)) \text{ One} \neq \text{One}$$

- What is the best way of working around this?

# Summary

- Formalized the theory of pGCL in higher-order logic.
- Created an automatic tool for deriving sufficient conditions for partial correctness.
  - Useful product of mechanizing a program semantics.
- There's still much to be done formalizing the theory and implementing practical program analysis tools.

# Related Work

- Formal methods for probabilistic programs:
  - Christine Paulin's work in Coq, 2002.
  - Prism model checker, Kwiatkowska et. al., 2000–
- Mechanized program semantics:
  - Formalizing Dijkstra, Harrison, 1998.
  - Mechanizing program logics in higher order logic, Gordon, 1989.