# Multiple-expectation Systems

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Motivation

- Motivation
  - Extension to Probabilistic B
  - Background
- Our Results/Contribution
  - Multiple Probabilistic Specification Substitutions
  - Fundamental Theorem
  - Case Study: Duelling Cowboys







### Outline

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# Extending probabilistic B

- To extend the scope of probabilistic B (pB) to cover systems with multiple probabilistic properties;
- Need to introduce multiple probabilistic specification substitution;
- Investigate the new substitution in the framework of layered developments.





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# How pGSL extends GSL

#### Expectations replace predicates

Predicates (functions from state to Boolean) are widened to Expectations (functions from state to non-negative real).

- For consistency with Boolean logic, we use embedded
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#### Summary

```
[x:=E]exp
```

The expectation obtained after replacing all free occurrences of x in exp by E

```
[skip] exp \qquad exp \\ [prog_1 p \oplus prog_2] exp \qquad p \times [prog_1] exp \\ + (1-p) \times [prog_2] exp \\ prog_1 \sqsubseteq prog_2 \qquad [prog_1] exp \Rightarrow [prog_2] exp \\ [prog_1 \parallel prog_2] exp \qquad [prog_1] exp \min [prog_2] exp \\ [0v pred \Rightarrow proglexp \min (v) \cdot (pred \mid prog_2) exp) \\ [0v pred \Rightarrow proglexp \min (v) \cdot (pred \mid prog_2) exp) \\ [0v pred \Rightarrow proglexp \min (v) \cdot (pred \mid prog_2) exp) \\ [0v pred \Rightarrow proglexp \min (v) \cdot (pred \mid prog_2) exp) \\ [0v pred \Rightarrow proglexp \min (v) \cdot (pred \mid prog_2) exp) \\ [0v pred \Rightarrow proglexp \min (v) \cdot (pred \mid prog_2) exp) \\ [0v pred \Rightarrow prog_1] exp \\ [0v pred \Rightarrow prog_2] exp \\ [0v pred \Rightarrow pred \Rightarrow prog_2] exp \\ [0v pred \Rightarrow pred
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exp



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prog_1 \sqsubseteq prog_2 \qquad [prog_1] exp \implies [prog_2] exp
[prog_1 \parallel prog_2] exp \qquad [prog_1] exp \min [prog_2] exp
```

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### Summary

```
[x:=E]exp
                                  The expectation obtained after re-
                                  placing all free occurrences of x in
                                  exp by E
[skip]exp
                                  exp
[prog<sub>1 p</sub>⊕ prog<sub>2</sub>]exp
                                   p \times [prog_1]exp + (1-p) \times [prog_2]exp
                                  [prog_1]exp \Rightarrow [prog_2]exp
prog_1 \sqsubseteq prog_2
```



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[x := E] exp \qquad \qquad \text{The expectation obtained after replacing all free occurrences of } x \text{ in } exp \text{ by } E
[skip] exp \qquad exp
[prog_{1} \ _{p} \oplus \ prog_{2}] exp \qquad p \qquad \times \ [prog_{1}] exp \qquad + \ (1-p) \qquad \times \ [prog_{2}] exp \qquad + \ (1-p) \qquad \times \ [prog_{2}] exp \qquad prog_{1} \ \sqsubseteq \ prog_{2}] exp \qquad [prog_{1}] exp \qquad min \ [prog_{2}] exp \qquad [prog_{1}] exp \qquad min \ [prog_{2}] exp \qquad [evp \ _{p} \ _{p} \ ] exp \qquad min \ (y) \cdot (pred \mid [prog] exp)
```





#### Syntax

 $v: \{A, B\}$ , where A and B are expectations over state x.

- v ⊆ x
- B can refer to the original state by using subscripted variables  $x_0$ .

The expected value of B over the set of final distributions is at least the expected value of A over the initial distribution.

#### Semantics

$$[v:\{A,B\}]$$
  $C = A \times [x_0:=x] \left( \Box x \cdot \left( \frac{C}{B \times \langle w=w_0 \rangle} \right) \right)$ 

(w is the set of unchanged variables, i.e. x - v). (Similar work can be seen in White[1996] and Ying[2003]

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### Fundamental theorem

#### Probabilistic Theorem

Assume that  $prog_1 = v : \{A, B\}$ . For any program prog<sub>2</sub>,

$$prog_1 \sqsubseteq prog_2$$

if and only if

$$A \Rightarrow [x_0 := x][prog_2] B^w$$
,

where 
$$B^w \cong B \times \langle w = w_0 \rangle$$
.





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# Multi-way probabilistic choice

For  $i \in (1..n)$ , let  $p_i$  be a probabilistic expression over the state satisfying

$$\sum_{i=1}^{n} \rho_i \le 1 ; \tag{1}$$

Let *S<sub>i</sub>* be a probabilistic substitution. The multi-way probabilistic choice is defined as follows:

$$\begin{bmatrix} S_{1} & @p_{1} \\ S_{2} & @p_{2} \\ \dots & \\ S_{n} & @p_{n} \end{bmatrix} E \equiv \begin{pmatrix} p_{1} \times [S_{1}] E \\ + & p_{2} \times [S_{2}] E \\ + & \dots & \\ + & p_{n} \times [S_{n}] E . \end{pmatrix} (2)$$

where E is an arbitrary expectation of the state.





# Set of pre- and post-expectations

For  $i \in (1..n)$ , let  $p_i$  be a probabilistic expression over the state x and free from  $x_0$  and satisfying:

$$\sum_{i=1}^{n} p_i \le 1 ; \tag{3}$$

let  $Q_i$  be predicates defined over  $x_0$ , v (where v is a subset of x) and satisfying, for all  $Q_i$ , that we have

$$\forall x_0 \cdot (\exists v \cdot Q_i) . \tag{4}$$

#### Semantics

Let  $p_0 = 1 - \sum_{i=1}^{n} p_i$ , we define

$$v: \begin{vmatrix} \{p_1, \langle Q_1 \rangle\} \\ \{p_2, \langle Q_2 \rangle\} \\ \dots \\ \{p_n, \langle Q_n \rangle\} \end{vmatrix} \cong \begin{vmatrix} \langle v : \{1, \langle Q_1 \rangle\} \rangle & @p_1 \\ \langle v : \{1, \langle Q_2 \rangle\} \rangle & @p_2 \\ \dots \\ \langle v : \{1, \langle Q_n \rangle\} \rangle & @p_n \\ \langle x : \{1, 1\} \rangle & @p_0 \end{vmatrix}.$$
 (5)

#### **Semantics**

Let  $p_0 = 1 - \sum_{i=1}^{n} p_i$ , we define

$$v: \begin{vmatrix} \{\rho_{1}, \langle Q_{1} \rangle\} \\ \{\rho_{2}, \langle Q_{2} \rangle\} \\ \cdots \\ \{\rho_{n}, \langle Q_{n} \rangle\} \end{vmatrix} \qquad \widehat{=} \qquad \begin{vmatrix} (v: \{1, \langle Q_{1} \rangle\}) & @p_{1} \\ (v: \{1, \langle Q_{2} \rangle\}) & @p_{2} \\ \cdots \\ (v: \{1, \langle Q_{n} \rangle\}) & @p_{n} \\ (x: \{1, 1\}) & @p_{0} \end{cases}$$
(5)

## Examples

#### A fair coin

$$S_1 \quad \widehat{=} \quad c: \begin{cases} \left\{ \frac{1}{2}, \langle c = H \rangle \right\} \\ \left\{ \frac{1}{2}, \langle c = T \rangle \right\} \end{cases} \tag{6}$$

#### A non-deterministic coin:

A coin which guarantees to return Heads at least 1/3 of the time and Tails at least 1/3 of the time.

$$S_2 \quad \widehat{=} \quad c: \left| \begin{array}{l} \left\{ \frac{1}{3}, \langle c = H \rangle \right\} \\ \left\{ \frac{1}{2}, \langle c = T \rangle \right\} \end{array} \right. \tag{7}$$





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We consider a special set of multiple probabilistic specification substitutions where for any pair  $Q_i$  and  $Q_i$ , where  $i \neq j$ , we have

$$Q_i \wedge Q_j = false$$
, (8)

#### Probabilistic Theorem

For all programs T, if

$$(x:\{1,1\}) \subseteq T \text{ and } (9)$$

$$(v: \{p_i, \langle Q_i \rangle\}) \subseteq T, \text{ for all } i \in (1..n),$$
 (10)

then we have

$$v: \begin{cases} \{\rho_1, \langle Q_1 \rangle\} \\ \{\rho_2, \langle Q_2 \rangle\} \\ \dots \\ \{\rho_n, \langle Q_n \rangle\} \end{cases} \sqsubseteq T.$$
 (11)



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# Two cowboys

#### Conditions

There are two cowboys *X* and *Y* fighting a duel. They take turns to shoot at each other.

- The probability for X to hit his opponent is  $\frac{2}{3}$ .
- The probability for Y to hit his opponent is  $\frac{1}{2}$ .
- Assuming that X has the advantage of shooting first.

#### Question?

What are the guaranteed survival probabilities for both cowboys?





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# Formal specification

Let  $p_X$  and  $p_Y$  be the survival probability for X and Y, respectively. Let s be the cowboy which survives the duelling.

### Specification

$$s \leftarrow$$
 TwoCowboyXYSpec  $\widehat{=}$   $s : \begin{cases} \{p_X, \langle s = X \rangle\} \\ \{p_Y, \langle s = Y \rangle\} \end{cases}$ 





```
s \leftarrow \mathsf{TwoCowboyXYImp}
     VAR t, n IN
                    (s := X; n := 1) g \oplus t := Y
     END
```



```
s \leftarrow \mathsf{TwoCowboyXYImp}
     VAR t, n IN
          t := X : s := X : n := 2: // init
                   (s := X; n := 1) g \oplus t := Y
     END
```



```
s \leftarrow \mathsf{TwoCowboyXYImp}
    VAR t, n IN
         t := X : s := X : n := 2:
         WHILE n = 2 DO // Loop
                  (s := X; n := 1) g \oplus t := Y
         EXPECTATIONS ...
         END
    END
```



```
s \leftarrow \mathsf{TwoCowboyXYImp} \quad \widehat{=} \quad
     VAR t, n IN
          t := X; s := X; n := 2;
          WHILE n = 2 DO
               IF t = X THEN // body
                    (s : = X; n : = 1) \frac{1}{2} \oplus t : = Y
               ELSE
                    (s := Y; n := 1) t := X
               END
          EXPECTATIONS ...
          END
     END
```



In order to prove that  $TwoCowboyXYSpec \sqsubseteq TwoCowboyXYImp$ , we have to prove that

$$(s: \{p_X, \langle s = X \rangle\}) \subseteq \text{TwoCowboyXYImpl}$$
 (12)

and

$$(s: \{p_Y, \langle s = Y \rangle\}) \subseteq \text{TwoCowboyXYImpl}.$$
 (13)

Then we can apply the fundamental theorem for single probabilistic specification substitution for (12) and (13) separately. For (12) we have to prove that

$$p_X \Rightarrow [init; Loop] \langle s = X \rangle.$$
 (14)





# Recall proof rules for probabilistic loops

For a probabilistic loop, such as

$$100p = WHILE G DO S INVARIANT I EXPECTATION E END.$$

then  $A \Rightarrow [init; loop]B$  holds if the following satisfies:

$$\begin{array}{c|ccc} P1 & A & \Rightarrow & [init]E \\ \hline P2 & \langle G \wedge I \rangle *E & \Rightarrow & [S]E \\ \hline P3 & \langle \neg G \wedge I \rangle *E & \Rightarrow & B \\ \end{array}$$

(Here, I only concentrate on the maintenance of the expectation E)





### Tabular method

For proving (14), we try to "guess" the expectation of the loop by tabulating the probabilities of establishing the post-expectation s = X after executing one iteration of the loop.

	n = 2	$s = X \wedge n = 1$	$s = Y \wedge n = 1$
t = X	4/5	1	0
t = Y	2/5	1	0

We have the expectation of the loop is

$$E \quad \widehat{=} \quad \langle s = X \wedge n = 1 \rangle + \langle n = 2 \wedge t = X \rangle \times \frac{4}{5} + \langle n = 2 \wedge t = Y \rangle \times \frac{2}{5} . \tag{15}$$





Apply the proof rule P1, we need to prove that

$$p_X \Rightarrow [t:=X;s:=X;n:=2]E, \qquad (16)$$

which is equivalent to

$$p_X \quad \Rightarrow \quad \frac{4}{5} \ . \tag{17}$$

So we can choose  $p_X \equiv \frac{4}{5}$ , which will be the guaranteed surviving probability for X. With similar reasoning, we have  $p_Y \equiv \frac{1}{5}$ .





# Development in layers

#### Three Cowboys

Assume that we have another cowboy, namely Z with the probability of hitting his opponent is  $\frac{1}{3}$ . With similar setting, what are the surviving probabilities for the cowboys.

Here, we can use the specification of the two cowboys situation when write the implementation, for example, the case when *Y* has the turn to shoot can be specified as follows:

and the reasoning can be done similarly as in the case for two cowboys.





Motivation

- Abstractly specify and refine probabilistic systems with multiple properties.
- Development of these systems can be separated into layers.
- When the state is small, the expectation for loops can be found using the tabular method.





# For further reading I



Abstraction, Refinement and Proof for Probabilistic Systems. Springer-Verlag, 2004.

T.S. Hoang, Z. Jin, K. Robinson, C. Morgan and A. McIver. Development via Refinement in Probabilistic B — Foundation and Case Study.

Proceedings of the 4th International Conference of B and Z Users, volume 3455 of LNCS, 2005.

N. White.

Probabilistic Specification and Refinement *Master Thesis*, Keble College, 1996.

M.S. Ying.

Reasoning about probabilistic sequential programs in a probabilistic logic.

Acta Informatica, volume 39, 2003.



