#### LMS/BCS-FACS Evening Seminar — London and Sydney.

## Probabilistic Datatypes: Automating verification for abstract probabilistic reasoning



**Annabelle McIver** Chris Chen Carroll Morgan

Probabilistic power domains (Jones, Plotkin, Saheb-Djahromi)

- Randomised algorithms for resource, security, performance.
  - 1980-ish
- Efficient reasoning principles, semantics, algebras, logics.

- Source-level reasoning for sequential programs (no non-determinism) (Kozen)
- Source-level reasoning for sequential programs, pGCL (yes non-determinism!)



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# Efficient Reasoning for Probabilistic Programs

 Non-determinism — enables transition abstraction.  $P \sqsubseteq Q$  means that program Q satisfies all the properties of P

 Source-level reasoning.  $\{\phi\} P \{\psi\}$ is a *Probabilistic* Hoare-style triple;  $\phi$  and  $\psi$  are real-valued functions.

\* Tools for verifying all the triples...automatically if possible!

Source-level reasoning for sequential programs, pGCL (yes non-determinism!)



# A challenge program

#### Set c uniformly between 0 and N-1



- \* Why is this fast?
- \* Why is this a challenge?
- What is Inv ?
- \* What should we do?





# A challenge program, N=5, i=3





# A challenge program, N=5, i=3





c,v=1,4



# A challenge program, N=5, i=3



c,v=1,2











## What should we do?





# Abstract datatypes

The motivation behind the work in very-high-level languages is to ease the programming task by providing the programmer with a language containing primitives or abstractions suitable to his problem area. The programmer is then able to spend his effort in the right place; he concentrates on solving his problem, and the resulting program will be more reliable as a result. Clearly, this is a worthwhile goal.

Unfortunately, it is very difficult for a designer to select in advance all the abstractions which the users of his language might need. If a language is to be used at all, it is likely to be used to solve problems which its designer did not envision, and for which the abstractions embedded in the language are not sufficient.

> Barbara Liskov and Stephen Zilles, Proceedings of the ACM SIGPLAN symposium on Very high level languages, 1974

#### Abstract



# An example: a quick prototype

class Set(N): # New abstract version, capacity limited. # DTI: |ss|<=N # Data-type invariant. local: # PRE: N>=0 # Data-type precondition. ss= {} def makeEmpty: ss= {} def add(s): PRE: |ss|!=N # Must hold when add() is called. ss= ss∪{s} def isIn(s): return s∈ss

Taken from: Formal Methods, Informally, Carroll Morgan (to appear, CUP).

# Programming with specifications

class Set(N): # New abstract version, capacity limited.
 # DTI: ss <=N # Data-type invariant.</pre>

local:
 # PRE: N>=0
 ss= {}
def makeEmpty:
 ss= {}
def add(s):
 Pre: [ss]!=N
 ss= ss∪{s}
def isIn(s):
 return s∈ss

# Data-type precondition.

# Must hold when add() is called.

# Implementor side



SetmySet = Set(10);
mySet.add(3);
mySet.add(4);
Bool z = mySet.isIn(5);

#### User side





# What does this mean for verification?

**Definition 1.** An encapsulated datatype is a triple (I, OP, F), where I and F are two distinguished operations called respectively the initialisation and finalisation, and OP is an indexed set of publicly accessible operations. [12]

**Definition 2.** A datatype (I, OP, F) is refined by (I', OP', F') [12] if, for every program  $\mathcal{P}$  expressible using the constructs mentioned above, including calls on corresponding operators in OP and OP', we have

 $I; \mathcal{P}(OP); F \subseteq I'; \mathcal{P}(OP'); F'$ ,

where ";" indicates sequential composition.





# Today's talk

- \* What happens when some of the behaviour can be probabilistic?
- \* Do the traditional proof methods (eg simulation) still work?
- datatypes?
- \* If they don't, what must be changed?
  - \* A semantics and refinement that distinguishes hidden and visible state;
  - \* Does it apply to our challenge program?
  - \* We need to talk about information leaks...



\* Can we still use the abstract specification to prove properties of programs that use probabilistic



# What happens when the datatype is probabilistic?



# Abstract (specification) datatype for a coin flip



- var coin # Local variable
- Flip<sub>A</sub>: # Flip on demand. coin:= H $_{1/2} \oplus$  coin:= T; return coin

  - (a) The abstract datatype

### Probabilistic choice

# Programming with the coin datatype

var coin # Local variable skip  $I_A$ : Flip<sub>A</sub>: # Flip on demand. coin:= H  $_{1/2}\oplus$  coin:= T; return coin skip  $F_A$ :

(a) The abstract datatype







F

Demonic nondeterministic choice

What is the probability that g=v finally?

## $v := H \sqcap T;$ g:= Flip()



# Reasoning with the coin datatype

I v:= H□T; g:= Flip() F

This is what would be expected, but can we prove it formally? What happens when we implement the datatype?



What is the probability that g=v finally?

1/2 because the flip comes after the hondeterminism...?

## We can reason about the program by using the abstract datatype and Hoare Triples...

```
var coin # Local variable
I_A: skip
Flip<sub>A</sub>: # Flip on demand.
              coin:= H
        _{1/2}\oplus coin:= T;
        return coin
F_A:
      skip
```

(a) The abstract datatype



### { Pre Condition}

What about probability?

 $v := H \sqcap T;$ 

g:= Flip()

F

{ Post Condition}



var coin # Local variable  $I_A$ : skip Flip<sub>A</sub>: # Flip on demand. coin:= H  $_{1/2} \oplus$  coin:= T; return coin  $F_A$ : skip

(a) The abstract datatype

wp.Prog.[g=v] = p

### { p }

 $v := H \sqcap T;$ 

g:= Flip()

F



MONOGRAPHS IN COMPUTER SCIENCE

#### ABSTRACTION, REFINEMENT AND PROOF FOR PROBABILISTIC SYSTEMS

Annabelle McIver Carroll Morgan



Springer













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- ${[H=H]/2 + [T=H]/2 MIN}$ [H=T]/2 + [T=T]/2
  - $v := H \sqcap T;$
  - ${[H=v]/2 + [T=v]/2}$ 
    - g:= Flip()
      - $\{[g=v]\}$ 
        - F
      - $\{[g=v]\}\$



Ι











- $\{1/2\}$
- $v := H \sqcap T;$
- ${[H=v]/2 + [T=v]/2}$ 
  - g:= Flip()
    - $\{[g=v]\}\$ 
      - F
    - $\{[g=v]\}$



## wp.I.[1/2]













# Why is this valid?

 $v := H \sqcap T;$ g:= Flip() F

```
var coin # Local variable
       skip
I_A:
Flip<sub>A</sub>: # Flip on demand.
              coin:= H
        _{1/2}\oplus coin:= T;
        return coin
F_A: skip
    (a) The abstract datatype
```

$$g=v]$$



# Justification using the "copy rule"...



## The copy rule: inlining the code should mean the same thing







# How does all of this work with refinement of datatypes?



# A refinement example

var coin # Local variable

 $I_A$ : skip Flip<sub>A</sub>: # Flip on demand. coin:= H  $_{1/2} \oplus$  coin:= T; return coin  $F_A$ : skip (a) The abstract datatype Eige 1: Abstract and c

Why should the concrete refine the abstract?



var coin,c # Local variables  $I_C$ : c:= H<sub>1/2</sub> $\oplus$  T # Pre-flip. Flip<sub>C</sub>: coin:= c; c:= H<sub>1/2</sub> $\oplus$  T; # Pre-flip. return coin

 $F_C$ : skip

(b) The concrete datatype

Fig. 1: Abstract and concrete probabilistic datatypes



# A refinement example

var coin # Local variable

 $I_A$ : skip Flip<sub>A</sub>: # Flip on demand. coin:= H  $_{1/2} \oplus$  coin:= T; return coin  $F_A$ : skip (a) The abstract datatype

Fig. 1: Abstract and concrete probabilistic datatypes

The abstract datatype uses a "hidden" variable c to store a pre-flipped value — why shouldn't that matter?



(b) The concrete datatype



# Remember the definition of refinement?

**Definition 2.** A datatype (I, OP, F) is refined by (I', OP', F') [12] if, for every program  $\mathcal{P}$  expressible using the constructs mentioned above, including calls on corresponding operators in OP and OP', we have

where ";" indicates sequential composition.

Demonic nondeterministic choice cannot be resolved on the basis of internal state that it cannot "access" or "observe"

 $I; \mathcal{P}(OP); F \subseteq I'; \mathcal{P}(OP'); F'$ ,

 $v := H \sqcap T;$ g:= Flip()

In particular whether or not there has been a "preflip" of a local variable is not available, even at "run-time"

## Remember the definition of refinement?

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var coin # Local variable
$I_A$ : skip
Flip <sub>A</sub> : # Flip on demand.
coin:= H
$_{1/2}\oplus$ coin:= T;
return coin
$F_A$ : skip
(a) The abstract datatype

Fig. 1: Abstract and concrete probabilistic datatypes

These datatypes should therefore be equivalent in terms of their behaviour, and either one should result in g and v being equal with probability 1/2

var coin,c # Local variables

 $I_C$ : Flip<sub>C</sub>: coin:= c;

 $F_C$ :

(b) The concrete datatype

 $v := H \sqcap T;$ g:= Flip() F







## Let's look again at the copy rule, this time for the concrete datatype

In this in-lining, the hidden state is revealed! Is pGCL/MDP the wrong semantics for datatypes?

 $v := H \sqcap T;$ 

F

g:=

Flip()



var coin,c # Local variables  

$$I_C$$
: c:= H<sub>1/2</sub> $\oplus$  T # Pre-flip  
Flip<sub>C</sub>: coin:= c;  
c:= H<sub>1/2</sub> $\oplus$  T; # Pre-fli  
return coin  
 $F_C$ : skip  
(b) The concrete datatype





# What is the right semantics for probabilistic datatypes?



## et's take another look at refinement

**Definition 2.** A datatype (I, OP, F) is refined by (I', OP', F') [12] if, for every program  $\mathcal{P}$  expressible using the constructs mentioned above, including calls on corresponding operators in OP and OP', we have

where ";" indicates sequential composition.

The observed behaviour of datatypes makes an (unarticulated assumption) that the calling program has "no access" to the run-time internals of the datatype. This doesn't matter when there is no probability, but, as we've seen does when probability is involved.

 $I; \mathcal{P}(OP); F \subseteq I'; \mathcal{P}(OP'); F'$ ,

## Simulation is a well-known technique for this situation

**Definition 3.** We say that an operation rep is a simulation from (I, OP, F) to (I', OP', F') if -using  $j \in J$  to index corresponding operations in OP and OP'the following inequations hold [12]:





Simulation *should be consistent* with a copy rule, using a semantics that reflects the operational assumptions.



## We can "simulate" concrete behaviours with abstract ones

**Definition 3.** We say that an operation rep is a simulation from (I, OP, F) to (I', OP', F') if -using  $j \in J$  to index corresponding operations in OP and OP'the following inequations hold [12]:



I' rep; $OP'_{j}  \forall j \in J$ rep; $F'$	(1) (2) (3)
Local variables 2⊕T # Pre-flip. c; 2⊕T; # Pre-flip. coin	$rep: c := H \oplus_{1/2} c := T$ $Consistent with the MDP copy rule?$
crete datatype	



# How should simulation work?



$$I_{A}$$

$$v := H \sqcap T;$$

$$g := Flip_{A}()$$

$$F_{A}$$

$$I_{c}$$
  
v:= H \[T;  
g:= Flip()  
$$F_{c}$$





$$I_{A}$$

$$v := H \sqcap T;$$

$$g := Flip_{A}()$$

$$F_{A}$$





 $\mathbf{v} := \mathbf{H} \sqcap \mathbf{v} := \mathbf{T}$ 

$$I_{A}$$
  
v:= H $\sqcap$ T;  
g:= Flip()  
 $F_{A}$ 

IA

 $I_{C}$ 



 $I_{\rm C}$  $v := H \sqcap T;$ g:= Flip<sub>C</sub>()  $F_{\rm C}$ 

Glue a series of little proofs together via the rep...



 $c:= H\oplus_{1/2} c:= T$ 

 $\mathbf{v} := \mathbf{H} \sqcap \mathbf{v} := \mathbf{T}$ 

 $c := H \oplus_{1/2} c := T$ 

But this in NOT valid using an MDP semantics for probability, which does not distinguish hidden statements, i.e. c



If this little proof is valid, it means that probability distributes with non-determinisim

 $\mathbf{v} := \mathbf{H} \sqcap \mathbf{v} := \mathbf{T}$ 

 $c:= H\oplus_{1/2} c:= T$  $\mathbf{v} := \mathbf{H} \sqcap \mathbf{v} := \mathbf{T}$ 





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 $v := H \sqcap T;$ g:= Flip<sub>A</sub>()  $F_{\mathtt{A}}$ 

# How should simulation work?

#### $Flip_A$

$$I_{c}$$

$$v := H \sqcap T;$$

$$g := Flip_{c}()$$

$$F_{c}$$

This is not a valid proof of refinement with hidden state and probability...



# What do we learn from this?

**Definition 3.** We say that an operation rep is a simulation from (I, OP, F) to (I', OP', F') if -using  $j \in J$  to index corresponding operations in OP and OP'the following inequations hold [12]:



Demonic nondeterministic choice cannot be resolved on the basis of internal state that it cannot "access" or "observe"



 $v := H \sqcap T;$ g:= Flip() MDP's do not support a copy rule that distinguishes hidden (c state) from observed state.



## What is the right semantics for probabilistic datatypes so that internal state is "invisible" to external nondeterminism?

Partially Observable Hidden Markov Models!



#### MDP

### $\operatorname{coin} := \operatorname{H}_{1/2} \oplus \operatorname{coin} := \operatorname{T}$

$$\mathcal{V} \to \mathbb{PDV}$$

The abstract datatype has a visible coin to flip; the concrete datatype has a hidden c to flip





# MDP versus POMDP's

#### MDP

## $\operatorname{coin} := \operatorname{H}_{1/2} \oplus \operatorname{coin} := \operatorname{T}$ $v := 0 \quad \Box \quad v := 1$

$$\mathcal{V} \to \mathbb{PDV}$$



External nondeterminism cannot "observe" hidden coin flips inside the module





External nondeterminism cannot "observe" hidden coin flips inside the module



# $\operatorname{coin} := \operatorname{H}_{1/2} \oplus \operatorname{coin} := \operatorname{T}$ $v := 0 \quad \Box \quad v := 1$ $c := H_{1/2} \oplus T$ $v := 0 \quad \Box \quad v := 1$ External nondeterminism cannot "observe" hidden coin flips inside the module

#### MDP

 $u[coin = H \land v = 0, coin = T \land v = 1]$  $u[coin = T \land v = 0, coin = H \land v = 1]$  $u[coin = T \land v = 0, coin = H \land v = 0]$ 

# MDP versus POMDP's $u[coin = T \land v = 1, coin = H \land v = 1]$ $\mathcal{V} \times \mathbb{D}\mathcal{H} \longrightarrow \mathbb{P}\mathbb{D}(\mathcal{V} \times \mathbb{D}\mathcal{H})$ $v = 0 \land u[c = T, c = H]$

 $v = 1 \wedge u[c = T, c = H]$ 





## Simulation now works in POMDP's, consistent with copy rule!



$$I_{A}$$
  
v:= H $\sqcap$ T;  
g:= Flip()  
 $F_{A}$ 

#### Flip<sub>A</sub>

T

$$I_{c}$$
  
v:= H \[T;  
g:= Flip()  
$$F_{c}$$

This little proof is now valid!



# What does this mean for these developers?

var coin # Local variable

 $I_A$ : skip Flip<sub>A</sub>: # Flip on demand. coin:= H  $_{1/2} \oplus$  coin:= T; return coin  $F_A$ : skip (a) The abstract datatype

Why should the concrete refine the abstract?





![](_page_54_Picture_7.jpeg)

![](_page_54_Picture_8.jpeg)

## A small cadenza...

- Probabilistic invariants are sometimes simulation relations, and
- Refinement depends on run-time information leaks concerning about the hidden state

corresponding operators in OP and OP', we have

where ";" indicates sequential composition.

**Definition 2.** A datatype (I, OP, F) is refined by (I', OP', F') [12] if, for every program  $\mathcal{P}$  expressible using the constructs mentioned above, including calls on

 $I; \mathcal{P}(OP); F \subseteq I'; \mathcal{P}(OP'); F'$ ,

# How does this work for our challenge problem?

![](_page_56_Figure_1.jpeg)

- unlike the {inv} assertion style.

# We can model things like secure implementation of cloud storage...

![](_page_57_Picture_1.jpeg)

return "yes/no" without leaking any other information.

![](_page_57_Picture_3.jpeg)

# Conclusions for today

- \* What happens when some of the behaviour can be probabilistic in sequential programs?
- \* Do the traditional proof methods for sequential programs (eg simulation) still work?

 $((g,v),-) \xrightarrow{\mathsf{v}:= 0 \sqcap 1} \{((g,0),-),((g,1),-)\}$ 

 $((g,v), \mathsf{u}\{0,1\}) \quad \stackrel{\mathsf{v}:=\ \mathsf{O}\,\square\ \mathsf{1}}{\longrightarrow} \ \mathsf{1}\ \{((g,0), \mathsf{u}\{0,1\}), ((g,1), \mathsf{u}\{0,1\})\}$ 

 $c:=0_{1/2}\oplus 1$ 

- \* If they don't, what must be changed?

 $\mathsf{c}:= \mathsf{O}_{1/2} \oplus \mathsf{1}$ 

 $\mathcal{V} imes \mathbb{D}\mathcal{H} o \mathbb{P}\mathbb{D}(\mathcal{V} imes \mathbb{D}\mathcal{H})$ 

- \* A semantics and refinement that distinguishes hidden and visible state;
- \* We can now talk about information leaks.

![](_page_58_Picture_9.jpeg)

\* Can we still use the abstract specification to prove properties of programs that use datatypes?

![](_page_58_Picture_12.jpeg)

![](_page_58_Picture_14.jpeg)

![](_page_58_Picture_15.jpeg)

![](_page_58_Picture_16.jpeg)

simulation