

Towards QoS Prediction Based on Composition Structure Analysis and Probabilistic Models

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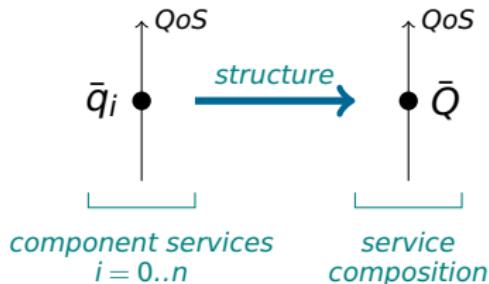
ICSOC 2014, Paris
6 Nov 2014

Introduction

- ▶ Quality of Service (QoS) critical for real-world usability
(performance, cost, user experience...)
 - ▶ design-time analysis ⇒ evolving high quality services
 - ▶ run-time prediction ⇒ proactive adaptation
 - ▶ simulation modeling ⇒ configuration, SLA offering
- ▶ QoS analysis for service compositions: *uncertainty*
 - ▶ component services: 3rd party components
 - ▶ limited information on implementation
 - ▶ many actors ⇒ many factors ⇒ uncertain data / QoS

Motivation: Uncertainty

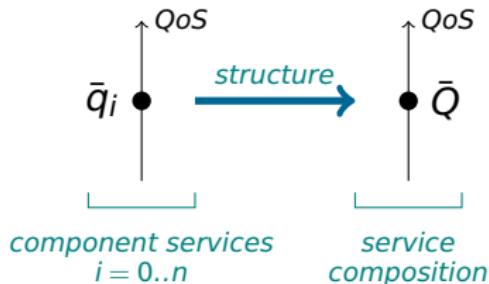
Averages



- ▶ how representative?
- ▶ spread?
- ▶ distribution shape?

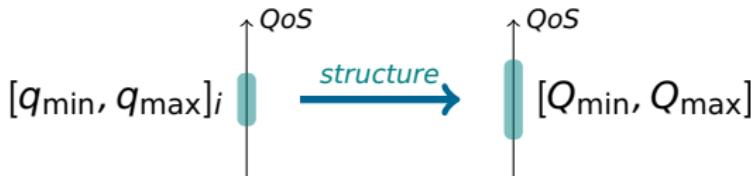
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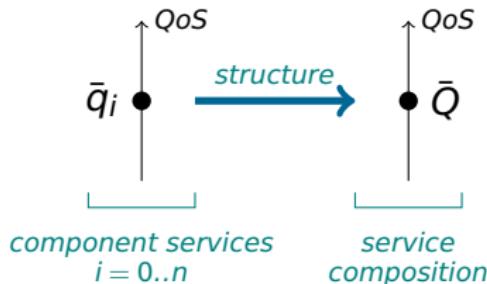
Intervals



- ▶ completeness?
- ▶ level of confidence?
- ▶ granularity?

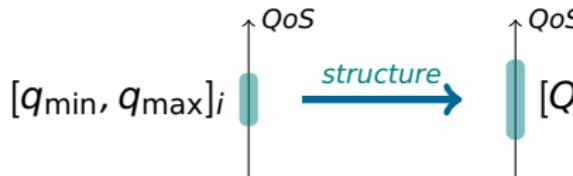
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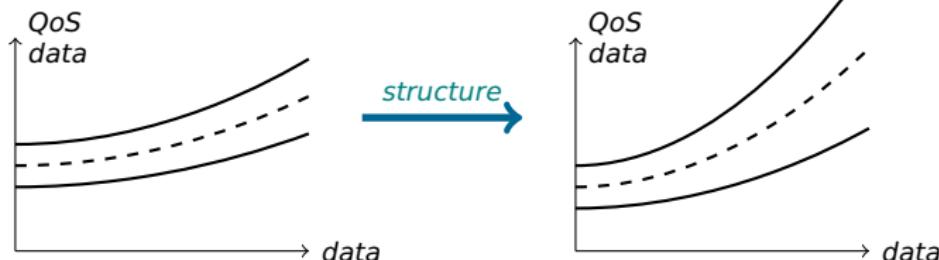


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Both cases: **effects of data?**

Motivation: Uncertainty

Data-dependent QoS bounds

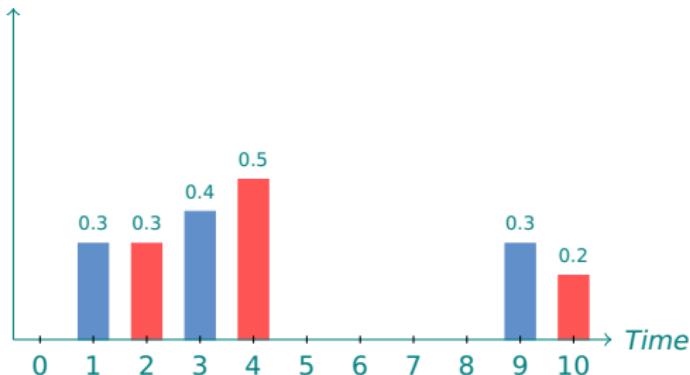
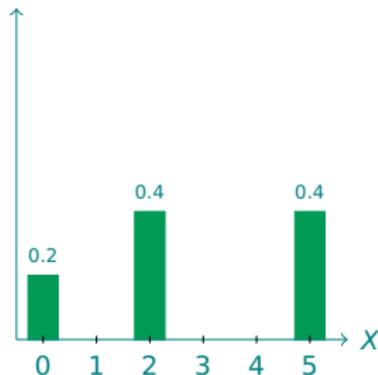


- ▶ safe bounds: data + cost
- ▶ functions of input data (size)
- ▶ complex control
- ▶ interval for every data point
- ▶ loss of precision?
- ▶ data uncertainty?

Unifying analysis

Interpret composition structure in a *probabilistic domain*

if $X > 3$ then call S_1 else call S_2

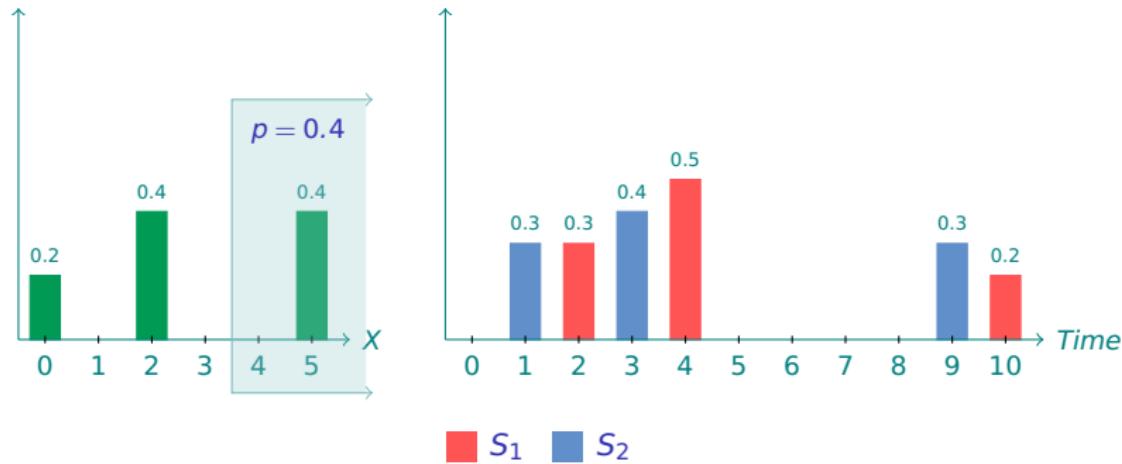


■ S_1 ■ S_2

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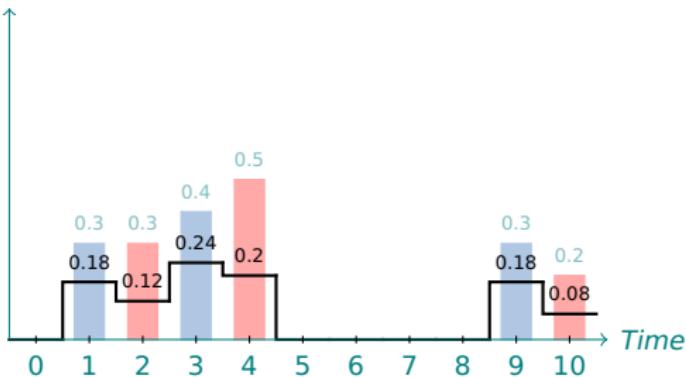
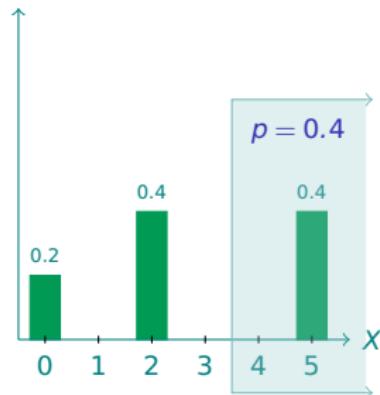
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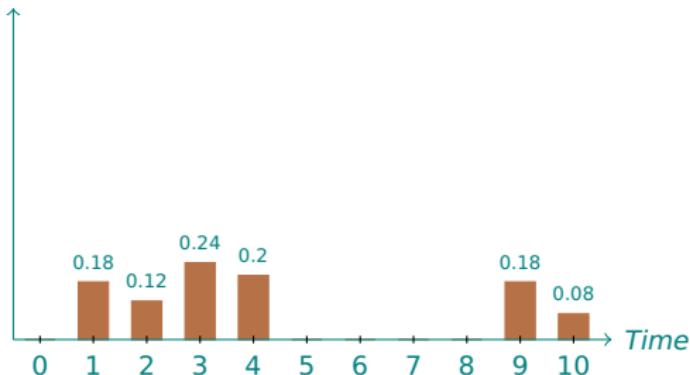
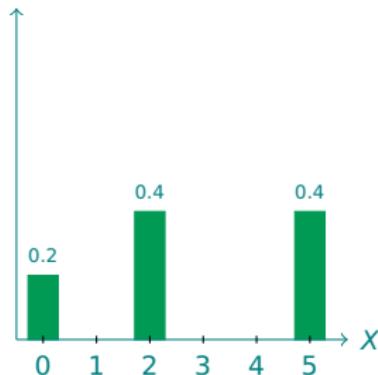


S_1 S_2

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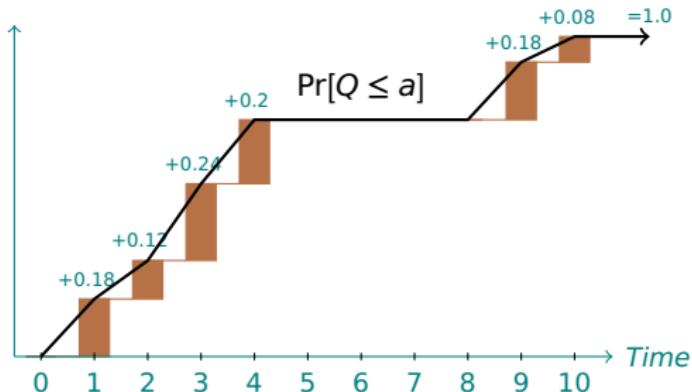
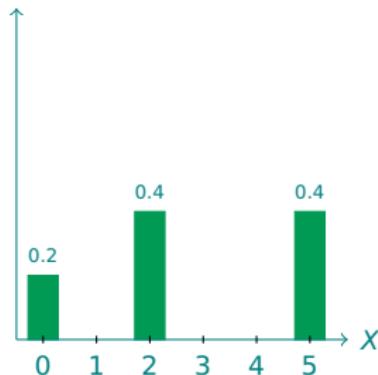
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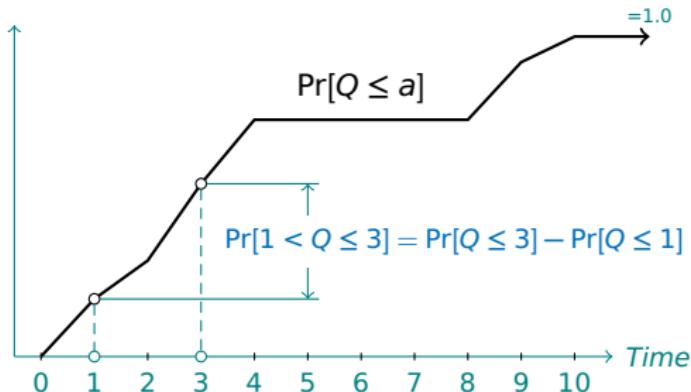
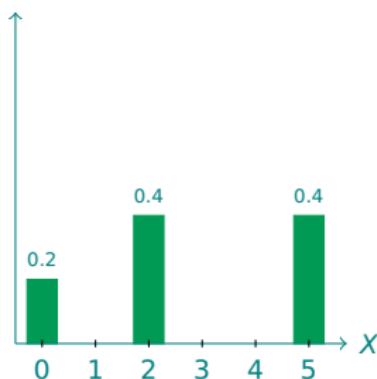
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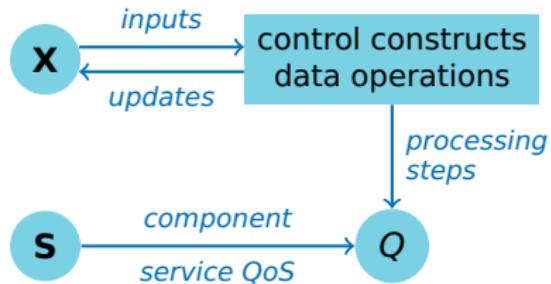
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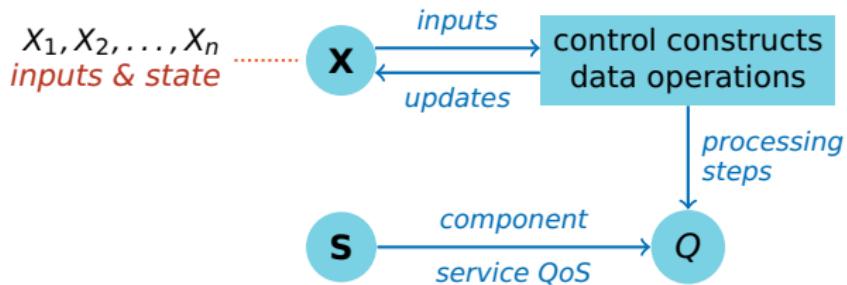
Proposed Approach

- ▶ Unifying analysis: interpreting in a *probabilistic domain*
data and QoS \rightarrow discrete random variables



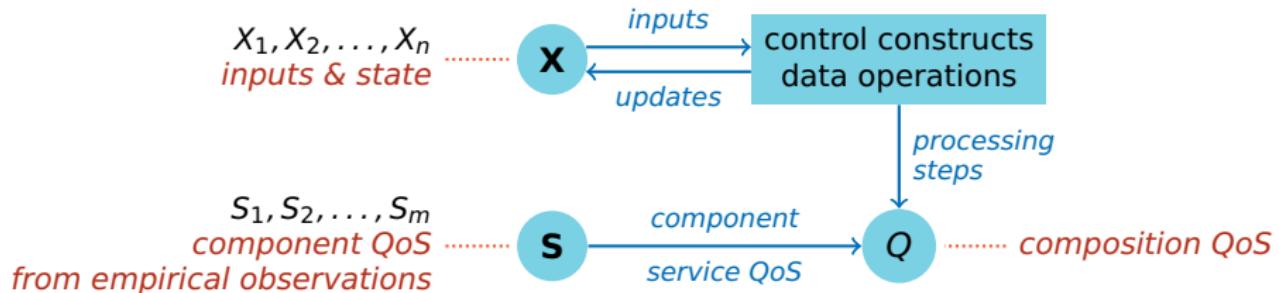
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Sample Language

- ▶ Featuring assignment, arithmetic, boolean conditions, usual control structures, service invocation.
- ▶ *Interpretation step*: ρ before $\rightarrow \rho'$ after.

Assignments and Arithmetics

$X := E$

- ▶ X becomes dependent on all variables in E
- ▶ Example: $X_1 := X_2 + X_3$

Initial distribution *Joint distribution*

x_2	ρ_{x_2}	x_3	ρ_{x_3}	x_1	x_2	x_3	ρ'_{x_1, x_2, x_3}
1	0.3	0	0.4	1	1	0	0.12
2	0.5	1	0.6	2	1	1	0.18
4	0.2			2	2	0	0.20
				3	2	1	0.30
				4	4	0	0.08
				5	4	1	0.12

- ▶ Logical condition: $\rho'(x, \mathbf{v}) = \sum \{\rho(u, \mathbf{v}) | x = E[u, \mathbf{v}]\}$
 $E[u, \mathbf{v}]$: value of E for $X = u$ and $\mathbf{V} = \mathbf{v}$
- ▶ Practical procedure: computing factor tables

Branching

if B then C_1 else C_2

- ▶ Variables in B become *co-dependent*
- ▶ Condition $B \rightarrow p$ is the probability **that that B is true**
 - ▶ Only some **(value, probability)** combinations possible when entering C_1 (same for C_2).

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- ▶ Variables in B become *co-dependent*
- ▶ Condition $B \rightarrow p$ is the probability that that B is true
 - ▶ Only some (value, probability) combinations possible when entering C_1 (same for C_2).
- ▶ Example: $B \equiv X_1 > X_2$

X_1	X_2	X_3	$\rho_{X_1, X_2 X_3}$	
1	1	0	0.3	$\neg B$
2	1	1	0.3	B
2	2	0	0.5	$\neg B$
3	2	1	0.5	B
4	4	0	0.2	$\neg B$
5	4	1	0.2	B

B

X_1	X_2	X_3	$\rho_{X_1, X_2 X_3} \rho_{X_3}$
2	1	1	0.18
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$\Sigma : 0.60 \leftarrow p$

$\neg B$

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1	1	0	0.12
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$\Sigma : 0.40 \leftarrow 1 - p$

- ▶ Resulting $p' = p \times p'_1 + (1 - p) \times p'_2$

Loops

while B do C

- ▶ Interpretation *unfolds it* into:

```
if  $B$  then begin  
     $C$ ;  
    while  $B$  do  $C$   
end  
else  
skip
```

- ▶ Assuming *termination*

Service Invocations

call (*service_i*)

- ▶ For *monotonic, cumulative* QoS metrics \equiv

$$Q := Q \oplus S_i$$

- ▶ Homogeneous data / QoS treatment.

Implementation Notes

- ▶ *Analyzer prototype implemented in Prolog*
 - ▶ ease of symbolic manipulation [ASTs, distributions]

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- ▶ Inputs:
 - ▶ initial distributions for inputs and state variables [**X**]
 - ▶ a description of the composition structure
 - ▶ QoS distributions for component services [empirical, **S**]
- ▶ Outputs:
 - ▶ final distribution for composition QoS [**Q**]
 - ▶ final distributions for composition state vars [**X – optional**]

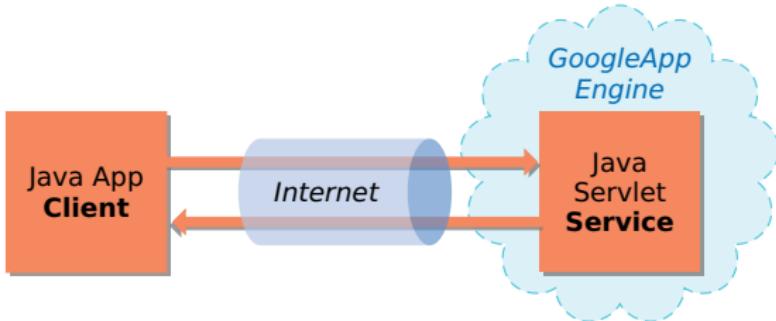
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- ▶ Example:

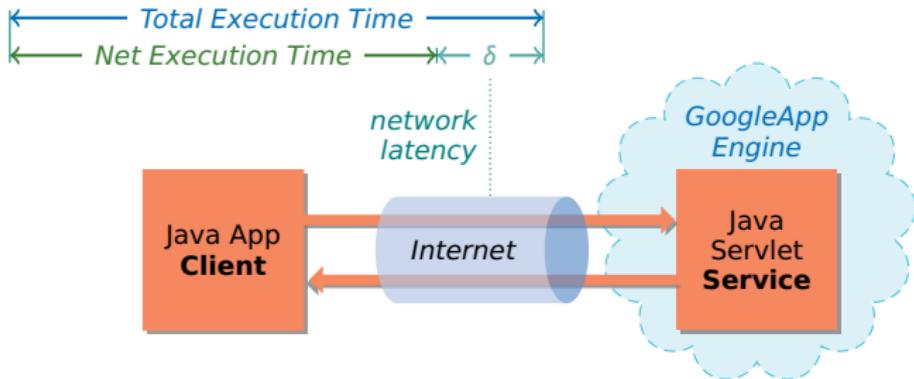
```
?- analyze(if(x>3, call(s1), call(s2)),  
           [x=[0-0.2,2-0.4,5-0.4]],  
           [s1=[2-0.3,4-0.5,10-0.2],  
            s2=[1-0.3,3-0.4,9-0.3]],  
           QoS  
           ).
```

QoS = [1-0.18, 2-0.12, 3-0.24, 4-0.2, 9-0.18, 10-0.08]

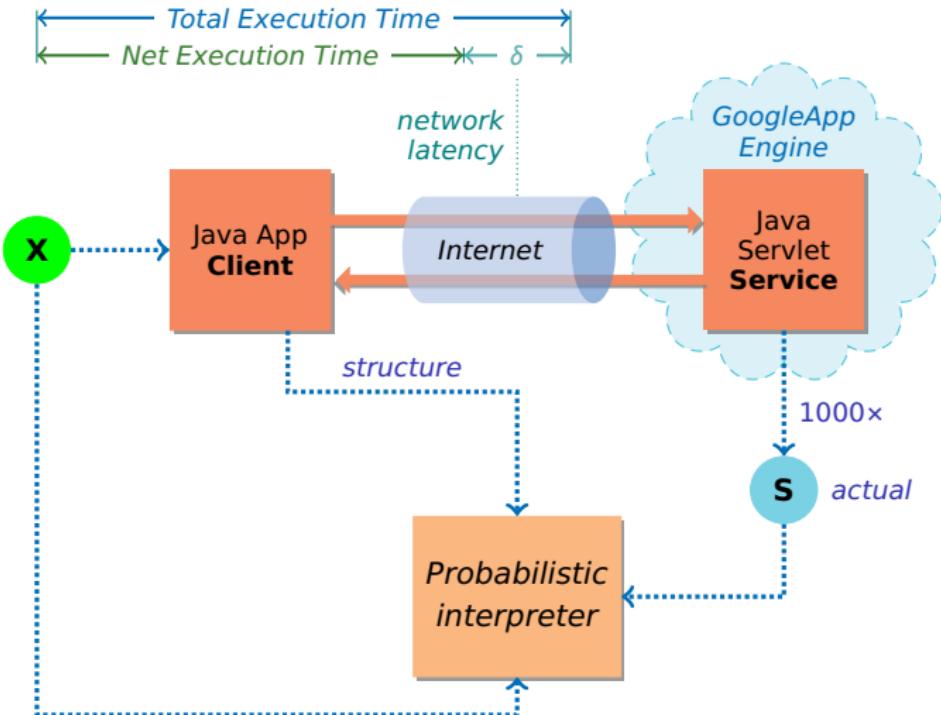
Validation: Execution Time



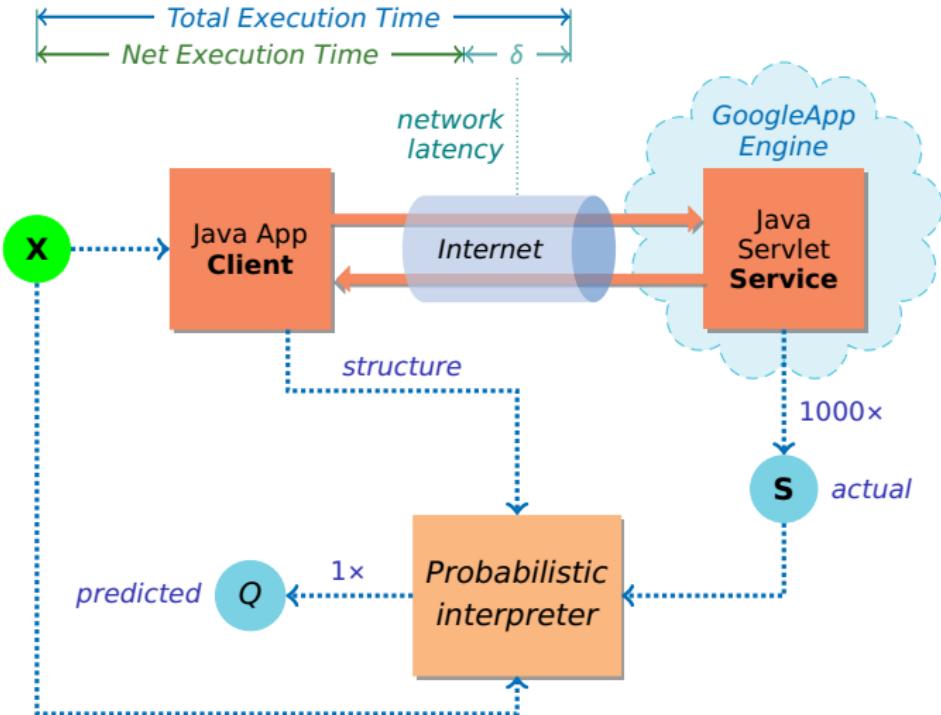
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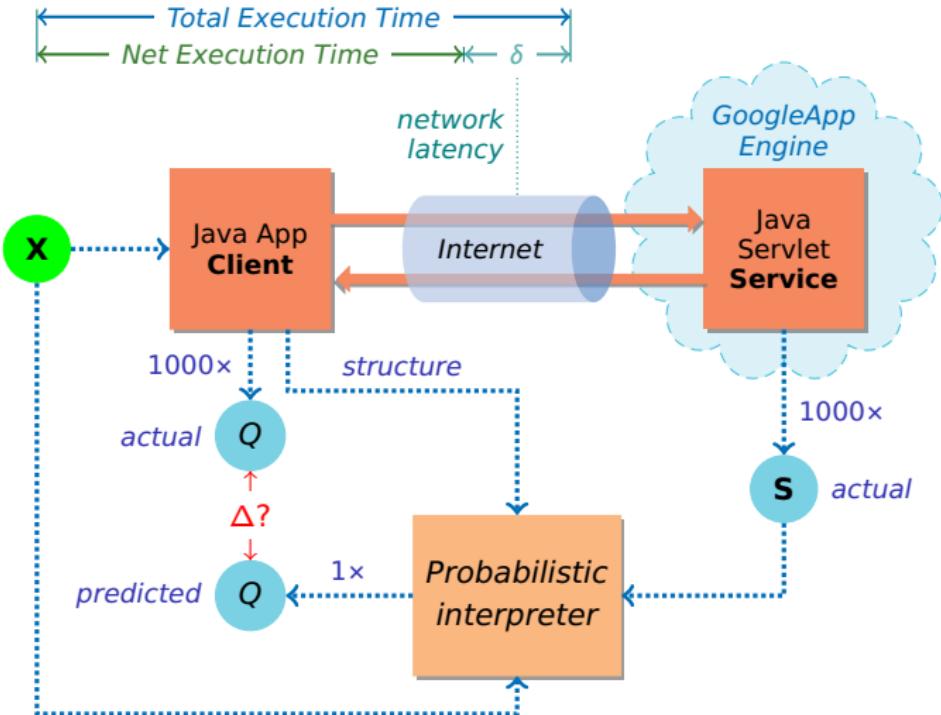
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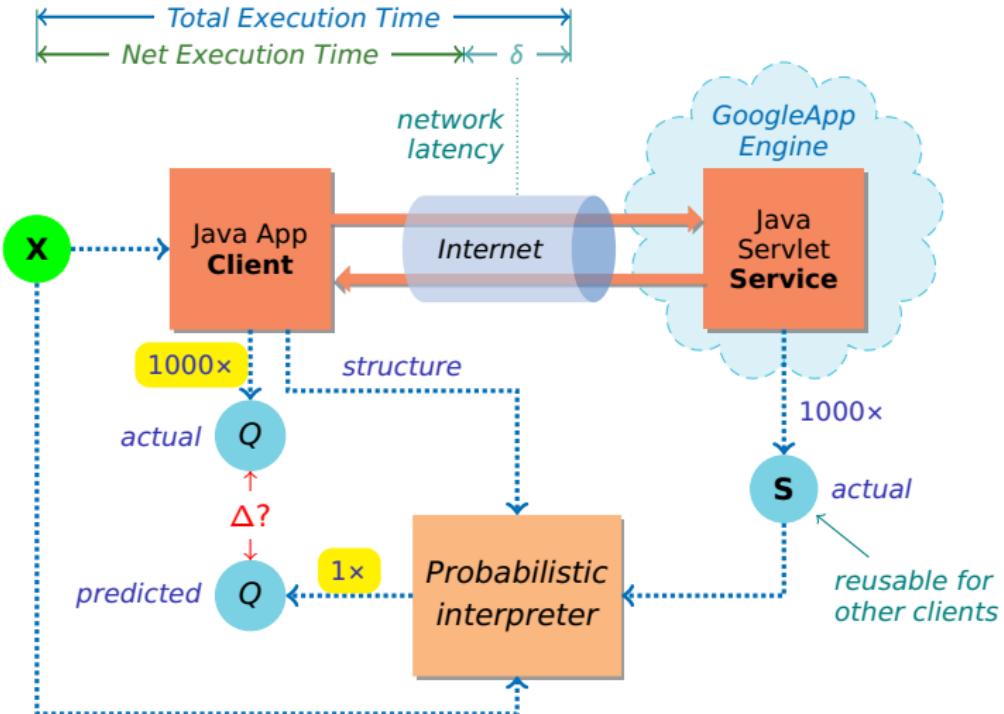
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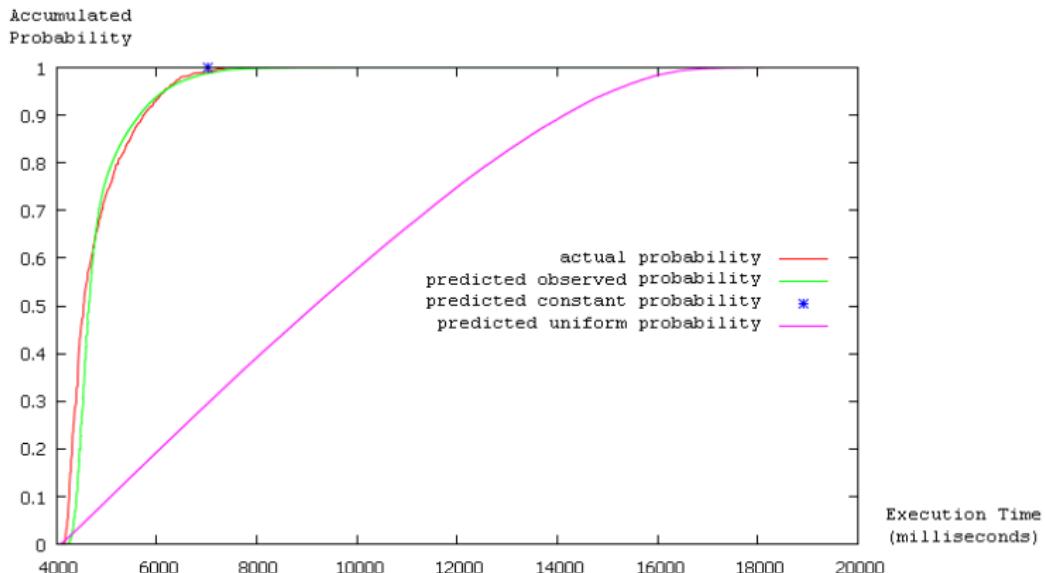
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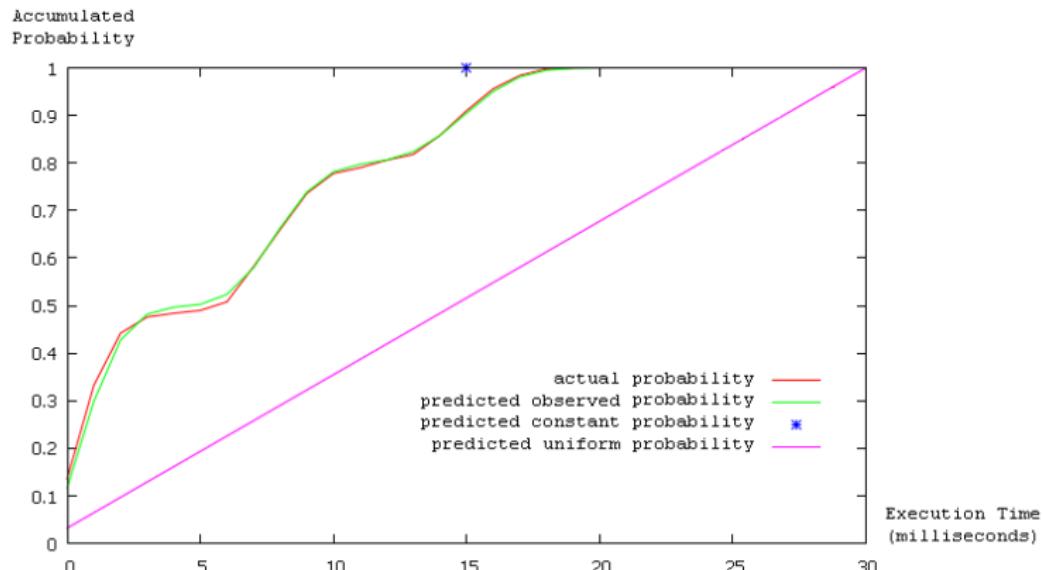


Validation Results: Fitting



- ▶ Very small mean square error ≈ 0.07 [total execution time]
 - ▶ *order-of-magnitude better* than uniform probability

Validation Results: Multi-Modality



- ▶ *Multi-modal distributions*: inflection points
 - ▶ difficult to analyze using standard measures of dispersion
 - ▶ still, very good fitting

Conclusions

- ▶ Modeling uncertainty with probability distributions
⇒ *finer grained, more detailed QoS predictions*
- ▶ Basic ingredients: structure + empirical data on component QoS → *probabilistic interpretation*
- ▶ Future work:
 - ▶ reverse: composition QoS → component QoS
 - ▶ tradeoffs: complexity ↔ precision
 - ▶ internal representation, algorithms
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Thank you!