

Service-Level Agreements for Service-Oriented Computing

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Joint work with Allan Clark and Mirco Tribastone

Bertinoro 2009





A contract between service provider and client.





- A contract between service provider and client.
- May involve availability:
 - Service has > 99% availability.





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 - Service has > 99% availability.
- May involve response time:
 - 97% of requests receive a response within 3 seconds.





- A contract between service provider and client.
- May involve availability:
 - Service has > 99% availability.
- May involve response time:
 - 97% of requests receive a response within 3 seconds.
- May be a combination of several statements such as these.





A modern approach to distributed computing.



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- Applications are built by composing services.



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- Services are replicated across a number of servers.



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- Providers publish services in registries.





- A modern approach to distributed computing.
- Applications are built by composing services.
- Services are replicated across a number of servers.
- Providers publish services in registries.
- Users discover services and bind to them.

Outline



- Analysing service-oriented computing
- Stochastic process algebras
 - Continuous-time Markov Chains
 - Transient analysis
 - Passage-time computation

- Example: Virtual University
 - Analysis results



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Uncertainties in service-oriented computing



- We do not know which service instances will be used.
- The service instances have different performance characteristics.
- The service instances may have different functionality.
- Plus all of the usual problems of distributed systems. . .



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- The service instances may have different functionality.
- Plus all of the usual problems of distributed systems. . .

...lots of difficulties for modellers!



Analysis of service-oriented computing



Put all possible descriptions of service behaviours together in one big model.

Analysis of service-oriented computing



- Put all possible descriptions of service behaviours together in one big model.
 - Hope your favourite large state-space method can cope ...



Statespace







Analysis of service-oriented computing



Separate out service bindings into different cases. Analyse cases separately. Re-combine results.

Analysis of service-oriented computing



- Separate out service bindings into different cases. Analyse cases separately. Re-combine results.
 - Scalable analysis of scalable systems.





• A way of specifying the behaviour of interest.





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 - Model using a process calculus.



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 - Perform parameter sweep across all values.
- A way of specifying uncertainty about bindings.
 - *Server* = {*UEDIN::Server*, *UNIPI::Server*}
 - Analyse all possible configurations by cases.





SRMC (Sensoria Reference Markovian Calculus)



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- PEPA (Performance Evaluation Process Algebra)
 - Use ipc (International PEPA Compiler) to compile to Hydra
- Hydra (Markovian response-time analyser)



Languages and tools



- SRMC (Sensoria Reference Markovian Calculus)
 - Use srmc (Sensoria Reference Markovian Compiler) to compile to PEPA
- PEPA (Performance Evaluation Process Algebra)
 - Use ipc (International PEPA Compiler) to compile to Hydra
- Hydra (Markovian response-time analyser)
 - Use hydra to compute response-time quantiles





SRMC
SINIVIC





SRMC

PEPA PEPA

PEPA

PEPA

PEPA





SRMC

PEPA PEPA PEPA PEPA

Hydra

Hydra

Hydra

Hydra

Hydra

Hydra

Hydra





SRMC PEPA PEPA PEPA PEPA PEPA Hydra Hydra



SRMC PEPA PEPA PEPA PEPA PEPA Hydra Hydra



SRMC PEPA PEPA PEPA PEPA **PEPA** Hydra Hydra



SRMC					
PEPA	PEPA	PEPA	PEPA	PEPA	
Hydra	Hydra	Hydra	Hydra	Hydra	
Hydra	Hydra	Hydra	Hydra	Hydra	
Hydra	Hydra	Hydra		Hydra	
Hydra	Hydra	Hydra		Hydra	
Hydra	Hydra	Hydra			
Hydra		Hydra			
Hvdra					

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Continuous-Time Markov Chains



We work with a stochastic process algebra (PEPA) which has Continuous-Time Markov Chains (CTMCs) as the underlying mathematical model.

Markov chains

Markov chains are finite state stochastic processes. The transition system of a Markov chain can be stored as a generator matrix, Q, constructed such that when we find a transition from state i to state j at rate r we add r to the current value of Q_{ij} .

Steady-state analysis



Investigation of SLAs may require steady-state analysis of a CTMC.

Steady-state analysis

We are concerned with finding the state probability row vector $\pi = [\pi_1, \dots, \pi_n]$ where π_i denotes the stationary probability that the CTMC is in state i.

Computing the stationary distribution



The stationary distribution can be computed using procedures of numerical linear algebra.

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Global balance equation

$$\pi Q = 0$$



Computing the stationary distribution



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Global balance equation

$$\pi Q = 0$$

Normalisation condition

$$\sum \pi = 1$$

Modelling with quantified process algebras



Tiny example

$$\begin{array}{ll} \texttt{P1} = (\texttt{start}, \texttt{r}).\texttt{P2} & \texttt{P2} = (\texttt{run}, \texttt{r}).\texttt{P3} & \texttt{P3} = (\texttt{stop}, \texttt{r}).\texttt{P1} \\ \texttt{System} = (\texttt{P1} \parallel \texttt{P1}) & \end{array}$$

Modelling with quantified process algebras



Tiny example

$$P1 = (start, r).P2$$
 $P2 = (run, r).P3$ $P3 = (stop, r).P1$
System = $(P1 \parallel P1)$

This example defines a system with nine reachable states:

1. P1 || P1

4. P2 || P1

7. P3 || P1

2. P1 || P2

5. P2 || P2

8. P3 || P2

3. P1 || P3

6. P2 || P3

9. P3 || P3

The global balance equations and the normalisation condition ensure there is a unique stationary distribution over these states.



Modelling with quantified process algebras



Tiny example

$$P1 = (start, r).P2$$
 $P2 = (run, r).P3$ $P3 = (stop, r).P1$ System = $(P1 \parallel P1)$

The stationary distribution over the nine reachable states is:

1. 0.1111

4. 0.1111

7. 0.1111

2. 0.1111

5. 0.1111

8. 0.1111

3. 0.1111

6. 0.1111

9. 0.1111

(Each state has two outgoing transitions with rate r so none of them is more likely than the others.)



Transient analysis



Investigation of SLAs often requires the transient analysis of a CTMC.

Transient analysis

We are concerned with finding the transient state probability row vector $\pi(t) = [\pi_1(t), \dots, \pi_n(t)]$ where $\pi_i(t)$ denotes the probability that the CTMC is in state i at time t.





Tiny example

$$P1 = (start, r).P2$$
 $P2 = (run, r).P3$ $P3 = (stop, r).P1$
System = $(P1 \parallel P1)$

Using transient analysis we can evaluate the probability of each state with respect to time. For t=0:

1. 1.0000

4. 0.0000

7. 0.0000

2. 0.0000

5. 0.0000

8. 0.0000

3. 0.0000

6. 0.0000





Tiny example

$$P1 = (start, r).P2$$
 $P2 = (run, r).P3$ $P3 = (stop, r).P1$
System = $(P1 \parallel P1)$

Using transient analysis we can evaluate the probability of each state with respect to time. For t=1:

1. 0.1642

4. 0.1567

7. 0.0842

2. 0.1567

5. 0.1496

8. 0.0804

3. 0.0842

6. 0.0804





Tiny example

$$P1 = (start, r).P2$$
 $P2 = (run, r).P3$ $P3 = (stop, r).P1$
System = $(P1 \parallel P1)$

Using transient analysis we can evaluate the probability of each state with respect to time. For t=2:

1. 0.1056

4. 0.1159

7. 0.1034

2. 0.1159

5. 0.1272

8. 0.1135

3. 0.1034

6. 0.1135



Tiny example

$$P1 = (start, r).P2$$
 $P2 = (run, r).P3$ $P3 = (stop, r).P1$
System = $(P1 \parallel P1)$

Using transient analysis we can evaluate the probability of each state with respect to time. For t=3:

1. 0.1082

4. 0.1106

7. 0.1100

2. 0.1106

5. 0.1132

8. 0.1125

3. 0.1100

6. 0.1125





Tiny example

$$\begin{array}{ll} \texttt{P1} = (\texttt{start}, \texttt{r}). \texttt{P2} & \texttt{P2} = (\texttt{run}, \texttt{r}). \texttt{P3} & \texttt{P3} = (\texttt{stop}, \texttt{r}). \texttt{P1} \\ \texttt{System} = (\texttt{P1} \parallel \texttt{P1}) & \end{array}$$

Using transient analysis we can evaluate the probability of each state with respect to time. For t=4:

1. 0.1106

4. 0.1108

7. 0.1111

2. 0.1108

5. 0.1110

8. 0.1113

3. 0.1111

6. 0.1113





Tiny example

$$P1 = (start, r).P2$$
 $P2 = (run, r).P3$ $P3 = (stop, r).P1$ $System = (P1 \parallel P1)$

Using transient analysis we can evaluate the probability of each state with respect to time. For t=5:

1. 0.1111

4. 0.1110

7. 0.1111

2. 0.1110

5. 0.1110

8. 0.1111

3. 0.1111

6. 0.1111





Tiny example

$$P1 = (start, r).P2$$
 $P2 = (run, r).P3$ $P3 = (stop, r).P1$
System = $(P1 \parallel P1)$

Using transient analysis we can evaluate the probability of each state with respect to time. For t=6:

1. 0.1111

4. 0.1111

7. 0.1111

2. 0.1111

5. 0.1110

8. 0.1111

3. 0.1111

6. 0.1111





Tiny example

$$\begin{array}{ll} \texttt{P1} = (\texttt{start}, \texttt{r}).\texttt{P2} & \texttt{P2} = (\texttt{run}, \texttt{r}).\texttt{P3} & \texttt{P3} = (\texttt{stop}, \texttt{r}).\texttt{P1} \\ \texttt{System} = (\texttt{P1} \parallel \texttt{P1}) & \end{array}$$

Using transient analysis we can evaluate the probability of each state with respect to time. For t=7:

1. 0.1111

4. 0.1111

7. 0.1111

2. 0.1111

5. 0.1111

8. 0.1111

3. 0.1111

6. 0.1111

Uniformisation



Transient and passage-time analysis of CTMCs proceeds by a numerical procedure called *uniformisation*.

Uniformisation

The generator matrix, Q, is "uniformised" with:

$$P = Q/q + I$$

where $q > \max_i |Q_{ii}|$. This process transforms a CTMC into one in which all states have the same mean holding time 1/q.

Uniformisation







Passage-time computation



Passage-time computation is concerned with knowing the probability of reaching a designated target state from a designated source state. It rests on two key sub-computations.

- 1. Finding the time to complete n hops (n = 1, 2, 3, ...), which is an Erlang distribution with parameters n and q.
- 2. Finding the probability that the transition between source and target states occurs in exactly *n* hops.

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A Virtual University



The Sensoria Virtual University (SVU) is a (fictitious) virtual organisation formed by bringing together the resources of the universities at Edinburgh (UEDIN), Munich (LMU), Bologna (UNIBO), Pisa (UNIPI) and others.

The SVU federates the teaching and assessment capabilities of the universities allowing students to enrol in courses irrespective of where they are delivered geographically.



The Sensoria Virtual University







The Sensoria Virtual University







Content upload and download



Students download *learning objects* from the content download portals of the universities involved and upload archives of their project work for assessment. By agreement within the SVU, students may download from (or upload to) the portals at any of the SVU sites, not just the one which is geographically closest.



What we leave out



It is likely that the students make a conscious decision about which portal to bind to. However, we do not anticipate having any data about how the students make their choice so we will not include in our model any representation of the reasoning process leading to selection of one portal or another.



SRMC description of the UEDIN server



```
UEDIN::{
  lambda = 1.65; mu = 0.0275; gamma = 0.125; delta = 3.215;
  avail = \{0.6, 0.7, 0.8, 0.9, 1.0\};
  UploadPortal::{
     Idle = (upload, avail * lambda).Idle + (fail, mu).Down;
     Down = (repair, gamma).Idle;
  DownloadPortal::{
     Idle = (download, avail * delta).Idle + (fail, mu).Down;
     Down = (repair, gamma).Idle;
```

SRMC description of the LMU server



```
LMU::{
   lambda = 0.965; delta = 2.576;
   avail = \{0.5, 0.6, 0.7, 0.8, 0.9\};
  UploadPortal::{
     Idle = (upload, avail * lambda).Idle;
   DownloadPortal::{
      Idle = (download, avail * delta).Idle;
```

SRMC description of the UNIBO server



```
UNIBO::{
  lambda = 1.65; mu = 0.0275; gamma = 0.125; delta = 3.215;
  slambda = 1.25; sdelta = 2.255; avail = \{ 0.8, 0.9, 1.0 \};
  UploadPortal::{
     Idle = (upload, avail * lambda).Idle + (fail, mu).Down
           + (supload, avail * slambda). Idle;
     Down = (repair, gamma).Idle;
  DownloadPortal::{
     Idle = (download, avail * delta).Idle + (fail, mu).Down
           + (sdownload, avail * sdelta). Idle;
     Down = (repair, gamma).Idle;
```

SRMC description of a cautious client



SRMC description of an incautious client



```
Harry::{
   Idle = (start, 1.0).Download;
   Download = (download, _).(download, _).(download, _).Upload;

   Upload = (upload, _).(upload, _).Disconnect;

   Disconnect = (finish, 1.0).Idle;
}
```

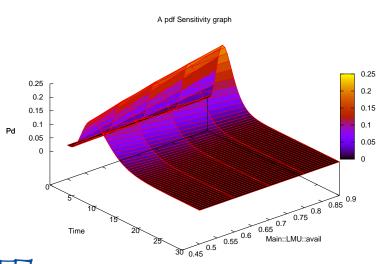
Client, Upload and Download portals





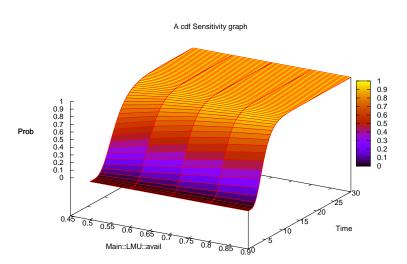
Probability Distribution Function [Harry, LMU, LMU]





Cumulative Distribution Function [Harry, LMU, LMU]



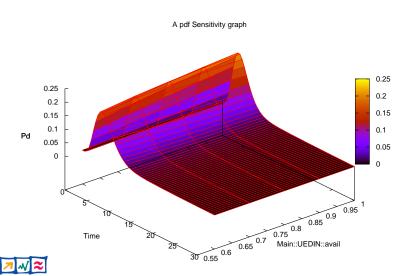




PDF [Harry, UEDIN, LMU], (LMU::avail=0.5)

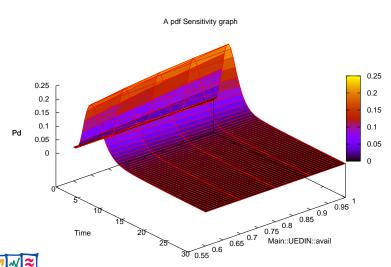
Information Society
Technologies





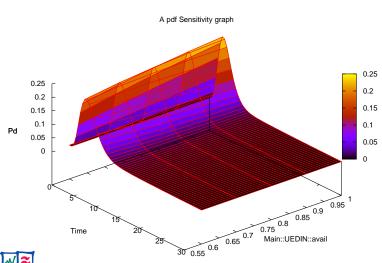
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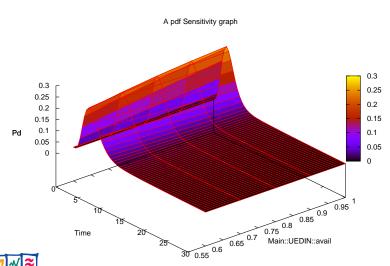
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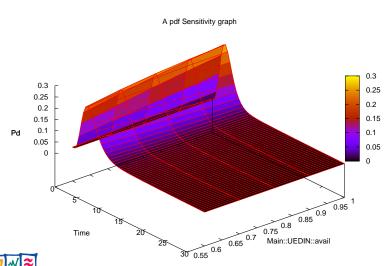
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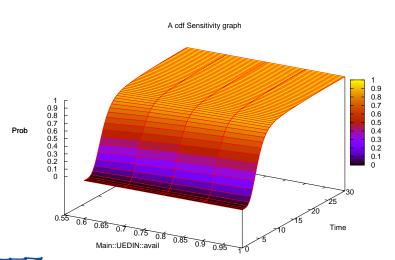
PDF [Harry, UEDIN, LMU], (LMU::avail=0.9)





CDF [Harry, UEDIN, LMU], (LMU::avail=0.5)

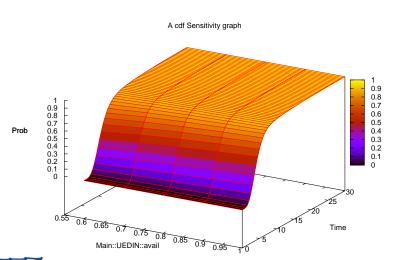






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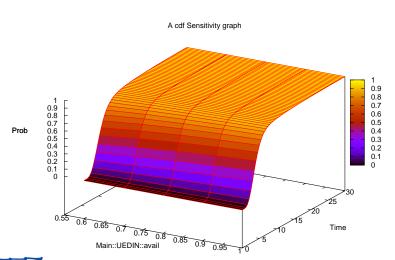






CDF [Harry, UEDIN, LMU], (LMU::avail=0.7)

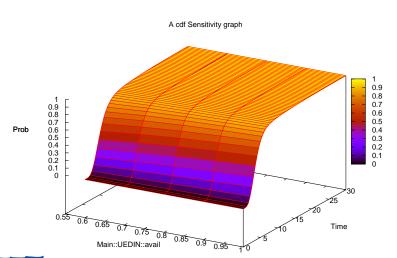






CDF [Harry, UEDIN, LMU], (LMU::avail=0.8)

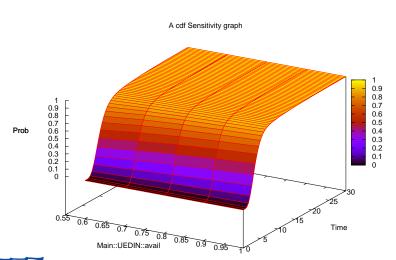






CDF [Harry, UEDIN, LMU], (LMU::avail=0.9)







Service-level agreements for SOC



The type of service-level argeement which we would attempt to state for service-oriented computing systems would include a *confidence interval*, a *path* through the system, a *time bound* and lower and upper *probability bounds*.

Service-level agreements for SOC

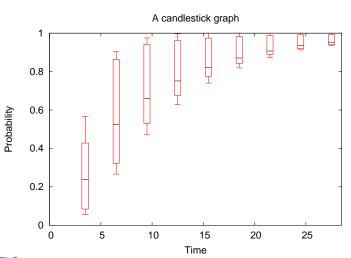


The type of service-level argeement which we would attempt to state for service-oriented computing systems would include a *confidence interval*, a *path* through the system, a *time bound* and lower and upper *probability bounds*.

For example: "Ninety percent of sessions will complete within 29 minutes with probability between 93.9% and 99.3%".

Candlestick 10% to 90%

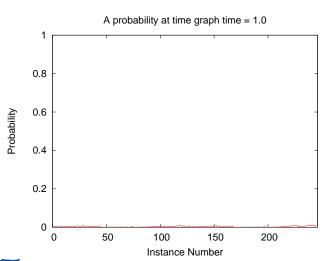






Probability of completion at t = 1.0

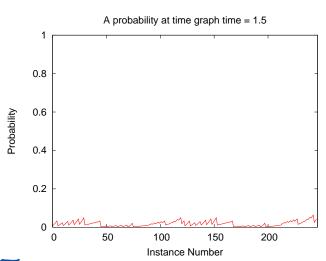






Probability of completion at t = 1.5

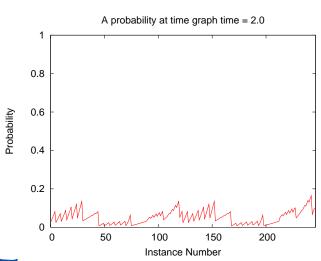






Probability of completion at t = 2.0

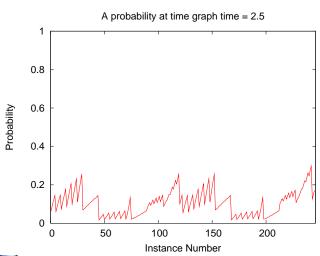






Probability of completion at t = 2.5

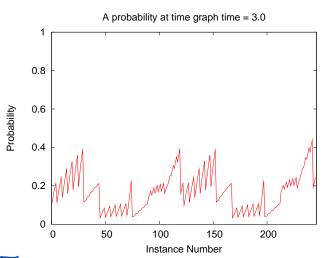






Probability of completion at t = 3.0

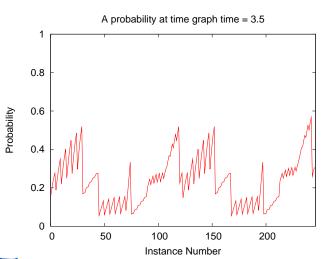






Probability of completion at t = 3.5

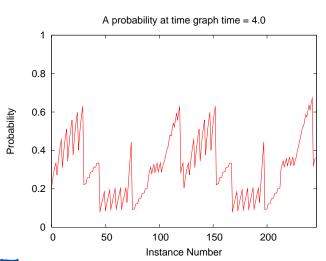






Probability of completion at t = 4.0

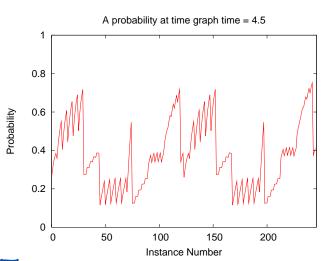






Probability of completion at t = 4.5

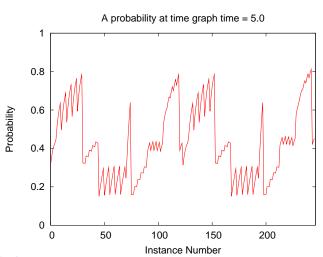






Probability of completion at t = 5.0

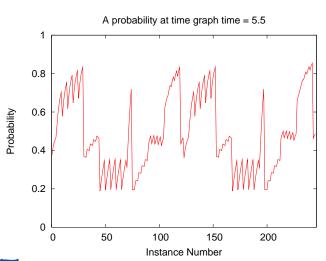






Probability of completion at t = 5.5

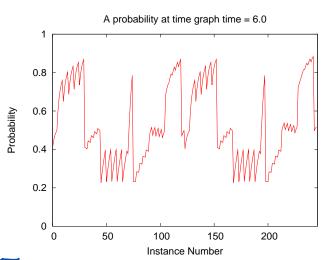






Probability of completion at t = 6.0

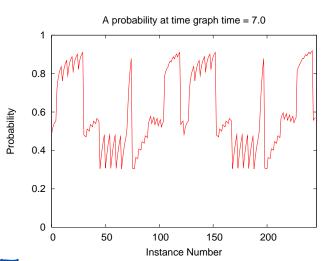






Probability of completion at t = 7.0

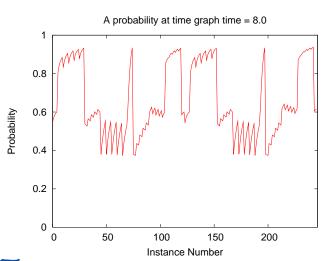






Probability of completion at t = 8.0

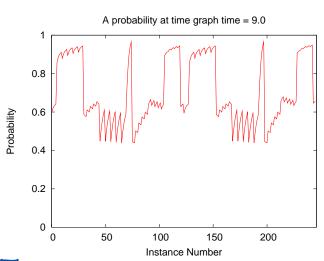






Probability of completion at t = 9.0

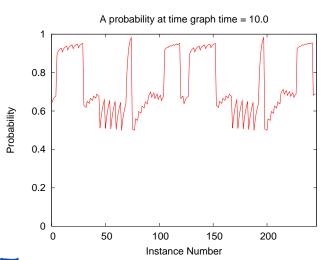






Probability of completion at t = 10.0

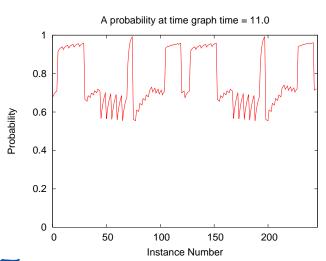






Probability of completion at t = 11.0

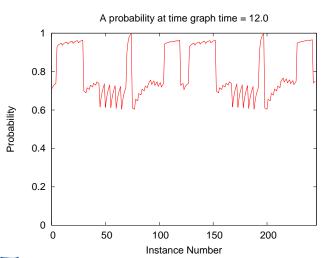






Probability of completion at t = 12.0

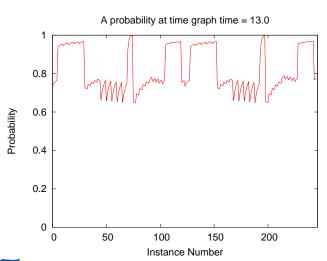






Probability of completion at t = 13.0

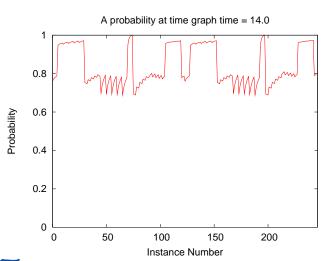






Probability of completion at t = 14.0

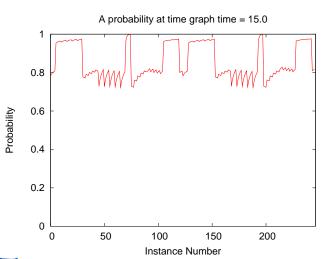






Probability of completion at t = 15.0

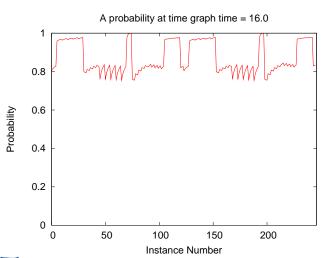






Probability of completion at t = 16.0

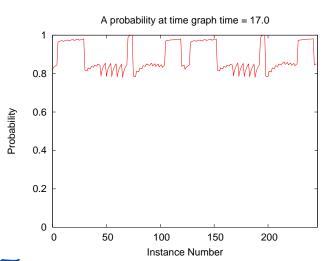






Probability of completion at t = 17.0

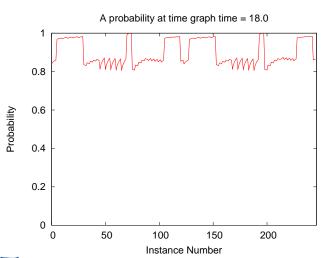






Probability of completion at t = 18.0

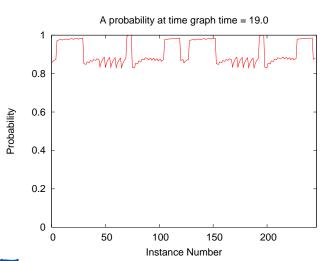






Probability of completion at t = 19.0

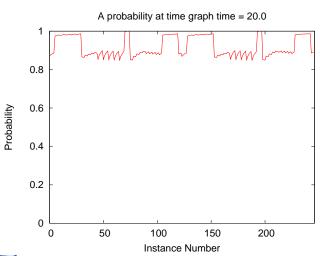






Probability of completion at t = 20.0

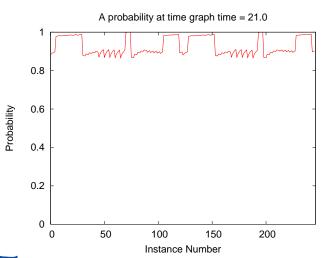






Probability of completion at t = 21.0

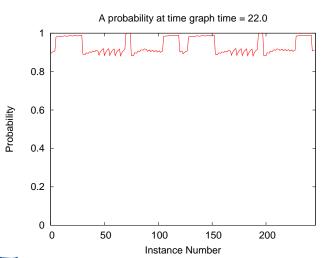






Probability of completion at t = 22.0

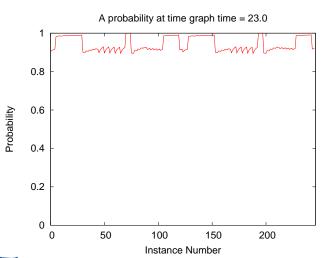






Probability of completion at t = 23.0

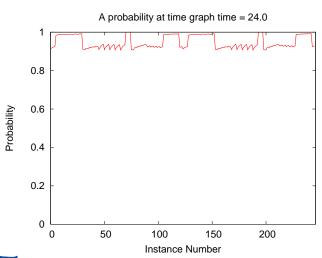






Probability of completion at t = 24.0

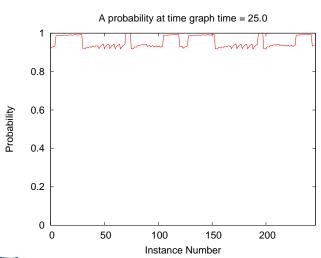






Probability of completion at t = 25.0

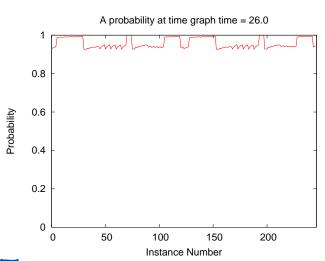






Probability of completion at t = 26.0

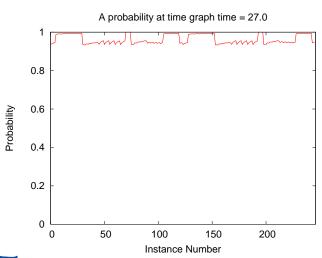






Probability of completion at t = 27.0

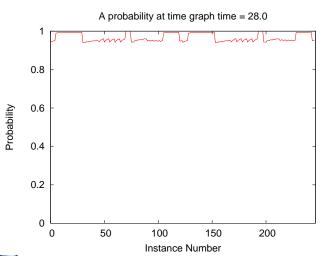






Probability of completion at t = 28.0

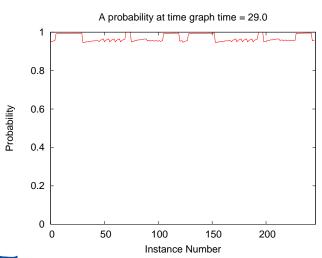






Probability of completion at t = 29.0

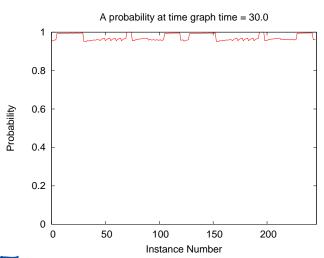






Probability of completion at t = 30.0







Discussion



- We might wonder if our infrastructure is necessary. Do we need to generate separate configurations and recombine the results?
- To test this we compared our SRMC model against a plain CTMC model of the system generated from a Generalised Stochastic Petri Net with vanishing markings.

Experiments with increasing model size



#	Client	Upload Portal	Download Portal
1	{Harry}	{UEDIN}	{UEDIN}
2	{Harry}	{UEDIN, LMU}	{UEDIN}
3	{Harry}	{UEDIN, LMU}	{UEDIN, LMU}
4	{Harry}	{UEDIN, LMU, UNIBO}	{UEDIN, LMU}
5	{Harry}	{UEDIN, LMU, UNIBO}	{UEDIN, LMU, UNIBO}
6	{Harry}	{UEDIN, LMU, UNIBO, UNIPI}	{UEDIN, LMU, UNIBO}
7	{Harry}	{UEDIN, LMU, UNIBO, UNIPI}	{UEDIN, LMU, UNIBO, UNIPI}
8	{Harry, Sally}	{UEDIN, LMU, UNIBO, UNIPI}	{UEDIN, LMU, UNIBO, UNIPI}

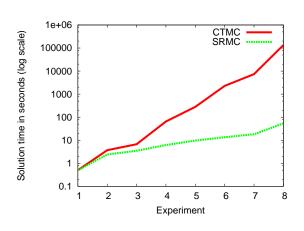
Experiments with increasing model size



		(СТМС		SRMC			
Н	Num.	Num.	Num.	time	Num.	Num.	Num.	time
#	states	config	runs	(secs)	states	config	runs	(secs)
1	32	1	5	0.5	32	1	5	0.5
2	48	1	25	3.7	32	2	30	2.4
3	72	1	25	6.7	32	4	60	3.5
4	120	1	75	65.0	32	6	90	6.2
5	200	1	75	280.0	32	9	123	9.7
6	280	1	225	2280.0	32	12	162	13.5
7	392	1	225	7390.0	32	16	204	18.2
8	784	1	225	\sim 134100.0	32	32	408	54.0

Experiments with increasing model size







Conclusions



- We addressed the inherent uncertainty in service-oriented computing by analysing by cases. We perform parameter sweep for each case. We can evaluate these in parallel (using Condor).
- The analysis methods scale well with increasing problem size.
- We build on tried and trusted compilers and analysers.
- Hopefully a "real world" approach to a "real world" problem.



Thank you!







Conclusions



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