

# Aci SurePath

## Perfect simulation of finite capacity queueing networks

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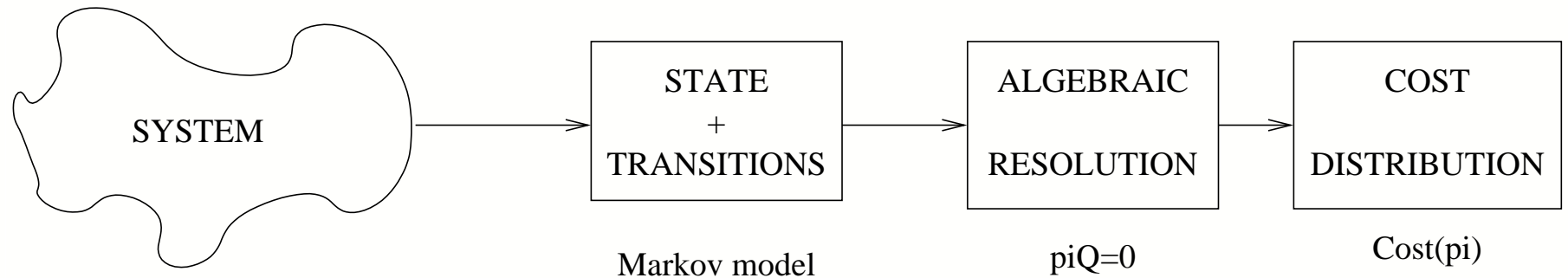


# Outline

1. Motivations, simulation of Markov chains and availability
2. Markovian queueing networks
3. Perfect simulation
4. Events and monotonicity
5. PSI2 architecture
6. Examples and demo
7. Conclusion and future works



# Modeling discrete event systems



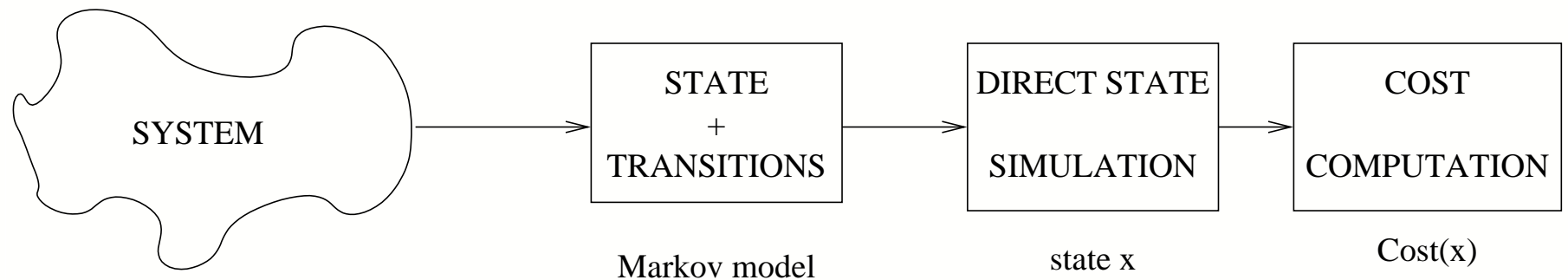
## Difficulties:

- complex structure synchronizations
- rare event probability estimation
- analytical/numerical method
- approximation/bounding techniques

⇒ reduction of the state space



# Modeling discrete event systems

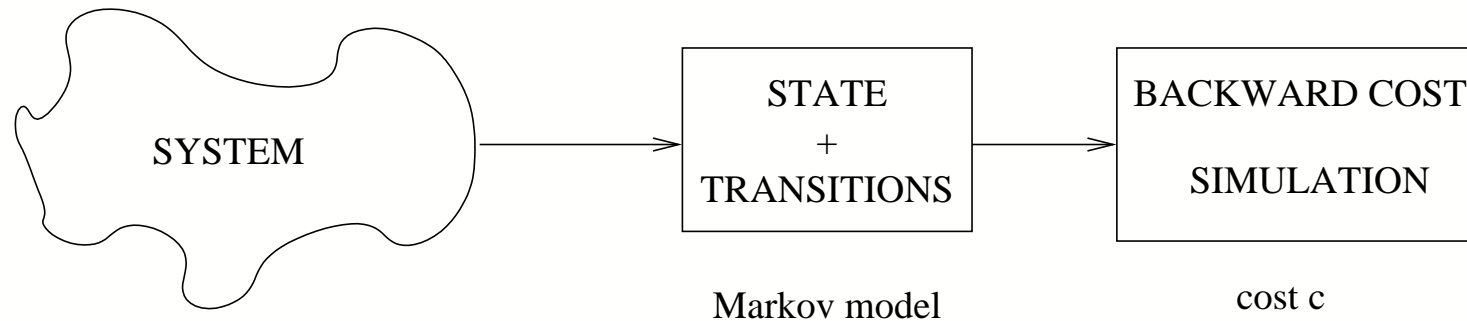


## Difficulties:

- stopping criteria : burn in time
- simulation biases  $\|\pi_n - \pi_\infty\|$
- estimation biases : confidence intervals  $\mathcal{O}\left(\frac{1}{\sqrt{n}}\right)$



# Modeling discrete event systems



## Properties:

- Exact stopping criteria  
⇒ no simulation bias

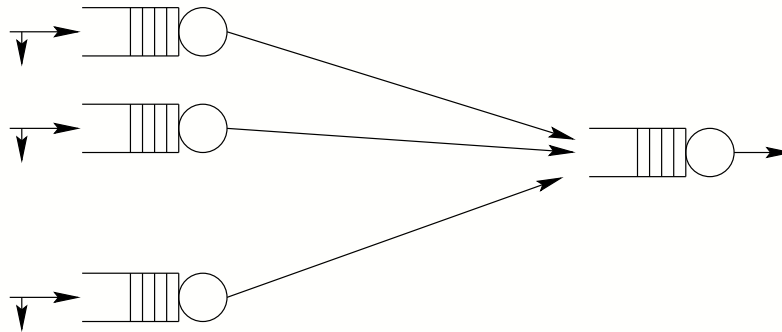
## Constraints:

- $N$  parallel trajectories



# Queueing systems models

## Model of resource contention



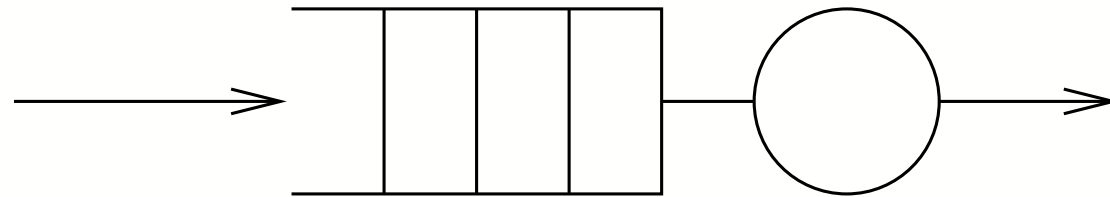
### Properties:

- Finite capacity queues
- Routing policies
- Blocking schemes
  - ⇒ estimation of losses and saturation
  - ⇒ rare events, availability



# Basic queue

Basic model of resource contention :  
time (server) and memory (capacity)



Markovian queues :

- arrival Poisson process  $\lambda$  : +1 transition
- exponential service time  $\mu$  : -1 transition
- Birth and Death process

⇒ monotonous process



# Basic queue : transition function

State space =  $\{0, 1, \dots, C\}$

Two types of events :

$$\Phi(x, arrival) = \min(x + 1, C)$$

$$\Phi(x, departure) = \max(x - 1, 0)$$

Monotonicity according to events :

$$x \leq y \Rightarrow \Phi(x, event) \leq \Phi(y, event)$$

Remark : link with the st-monotonicity





# Network of queues : transition function

Queueing network

-  $K$  queues, capacity  $C_i$  for queue  $i$

- State space

$$\mathcal{X} = \{0, \dots, C_1\} \times \dots \times \{0, \dots, C_K\}$$

Events :  $e_1, \dots, e_m$

$$\Phi(x, event) = next\ state$$



# Iterated system of functions

$\mathcal{X}$  state space (size  $n$ );  $\mathcal{U}$  set of external input values

Transition function  $\Phi$

$$\begin{aligned}\Phi : \mathcal{X} \times \mathcal{U} &\longrightarrow \mathcal{X} \\ (x, u) &\longmapsto \Phi(x, u)\end{aligned}$$

If  $\{U_n\}_{n \in \mathbb{Z}}$  is IID then

$$X_0 = x_0, \quad X_{n+1} = \Phi(X_n, U_{n+1})$$

is a Markov chain (stochastic recursive sequence).

$$\{\Phi(\cdot, u)\}_{u \in [0,1[}$$

is an iterated system of function.

Reciprocally, given a transition matrix  $Q$  it is possible to build a family of function  $\Phi(\cdot, u)$  such that the associated process is a markov chain with transition matrix  $Q$ .

$\implies$  Simulation kernel



# Backward coupling simulation

## Idea :

Propp & Wilson(1996)

- reverse time
- run  $N$  parallel trajectories
- wait for coupling.

$$\mathcal{Z}_n = \Phi(\Phi(\cdots(\Phi(\mathcal{X}, U_{-n+1}), \cdots), U_{-1}), U_0).$$

potential set of reachable states at step  $n$

```
for all  $x \in \mathcal{X}$  do
   $y(x) \leftarrow x$ 
end for
repeat
   $u \leftarrow \text{Random};$ 
  for all  $x \in \mathcal{X}$  do
     $y(x) \leftarrow y(\Phi(x, u));$ 
  end for
until All  $y(x)$  are equal
return  $y(x)$ 
```



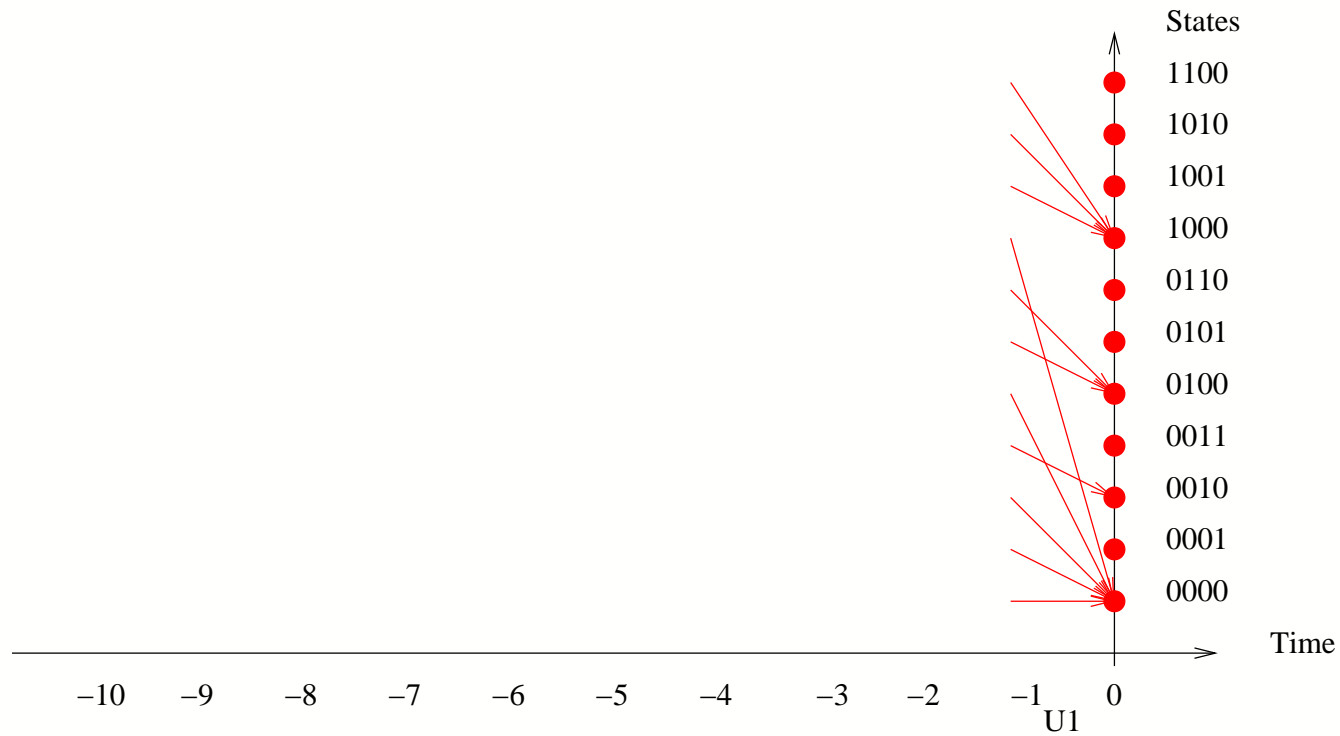
# Backward simulation example



$$Z_0 = \mathcal{X}$$

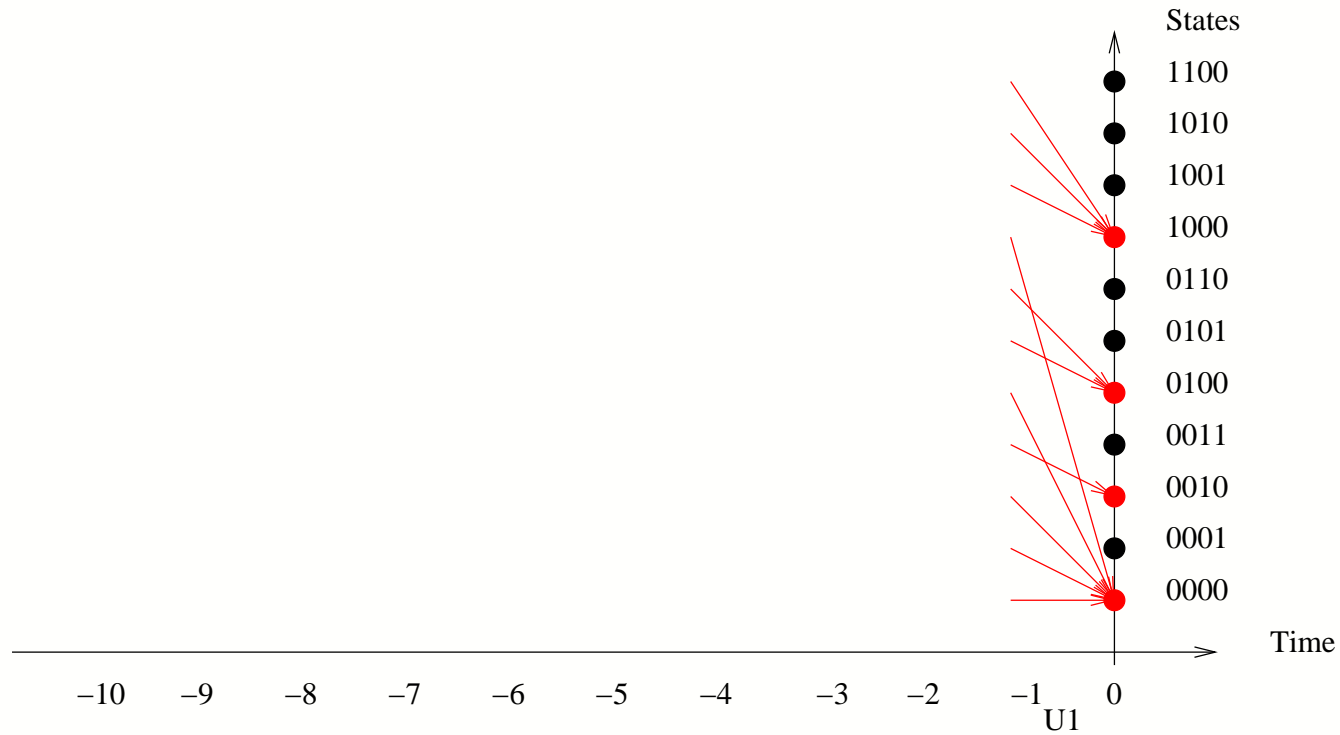


# Backward simulation example



$$Z_0 = \mathcal{X}$$

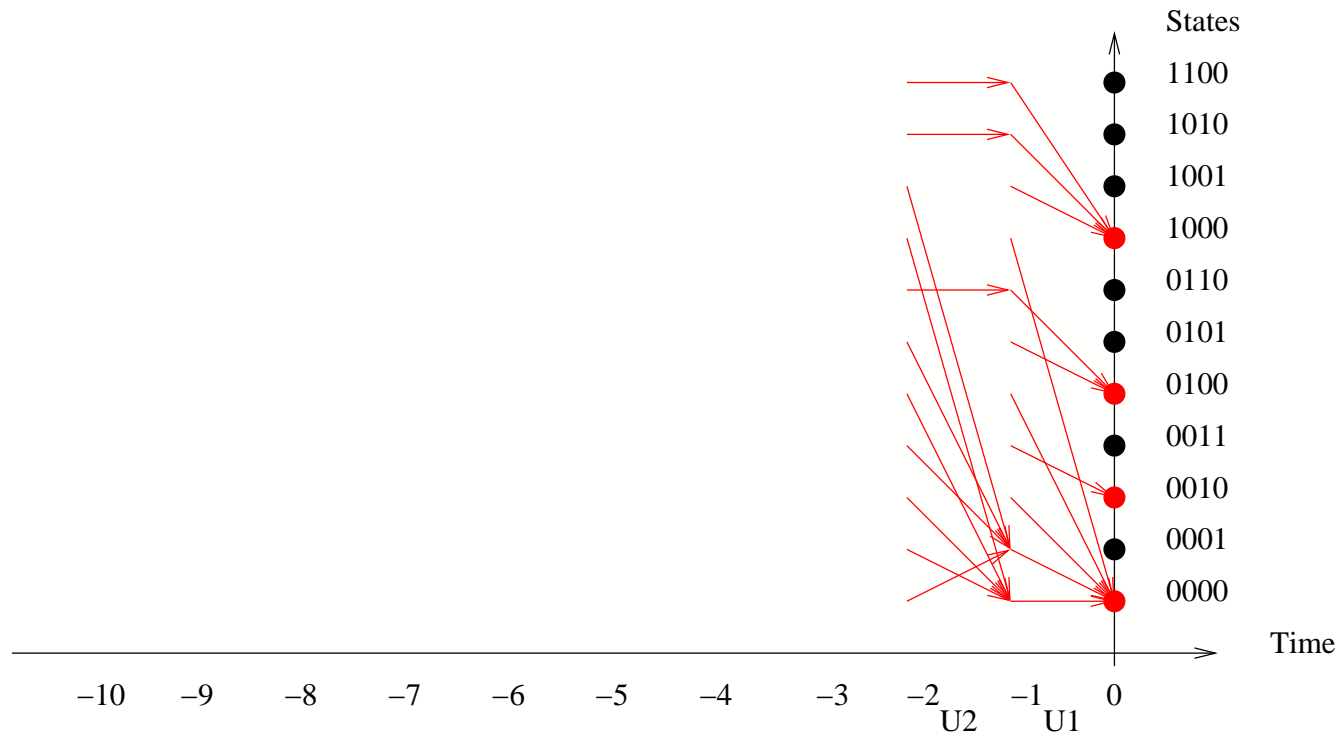
# Backward simulation example



$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

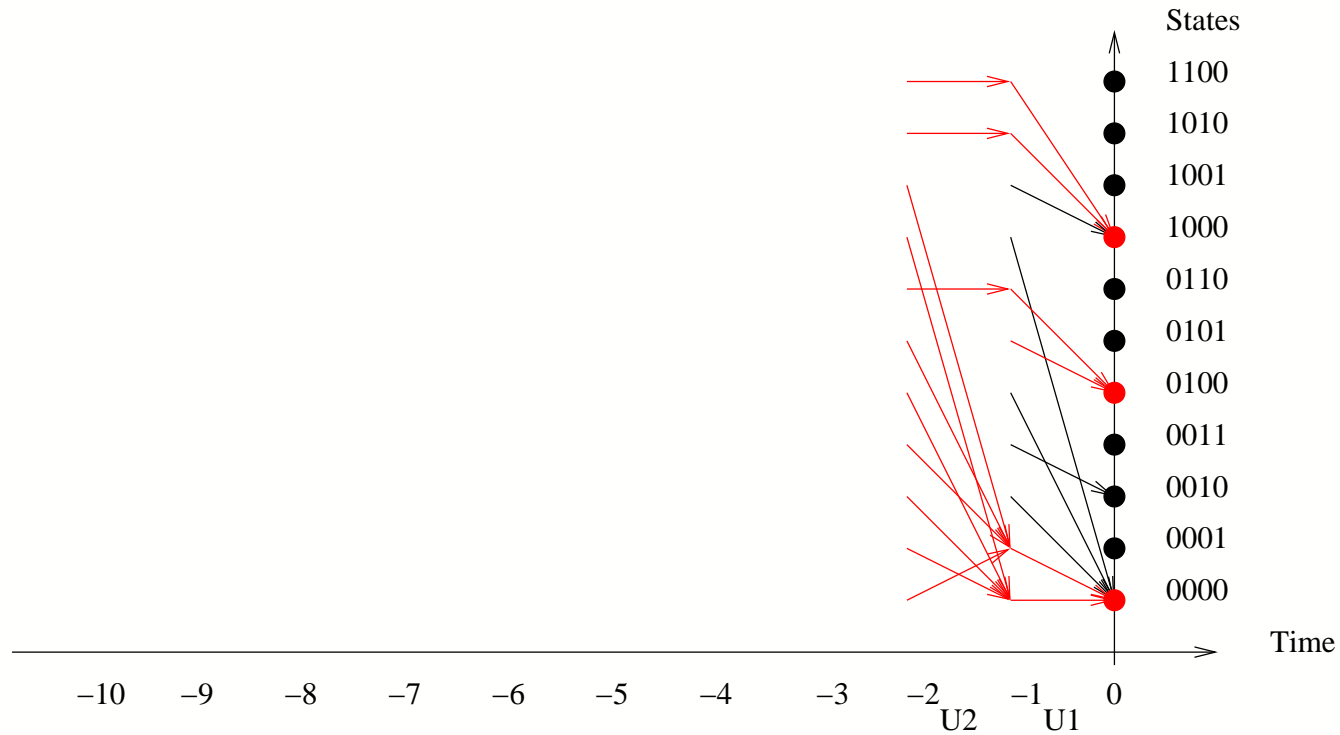
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# Backward simulation example



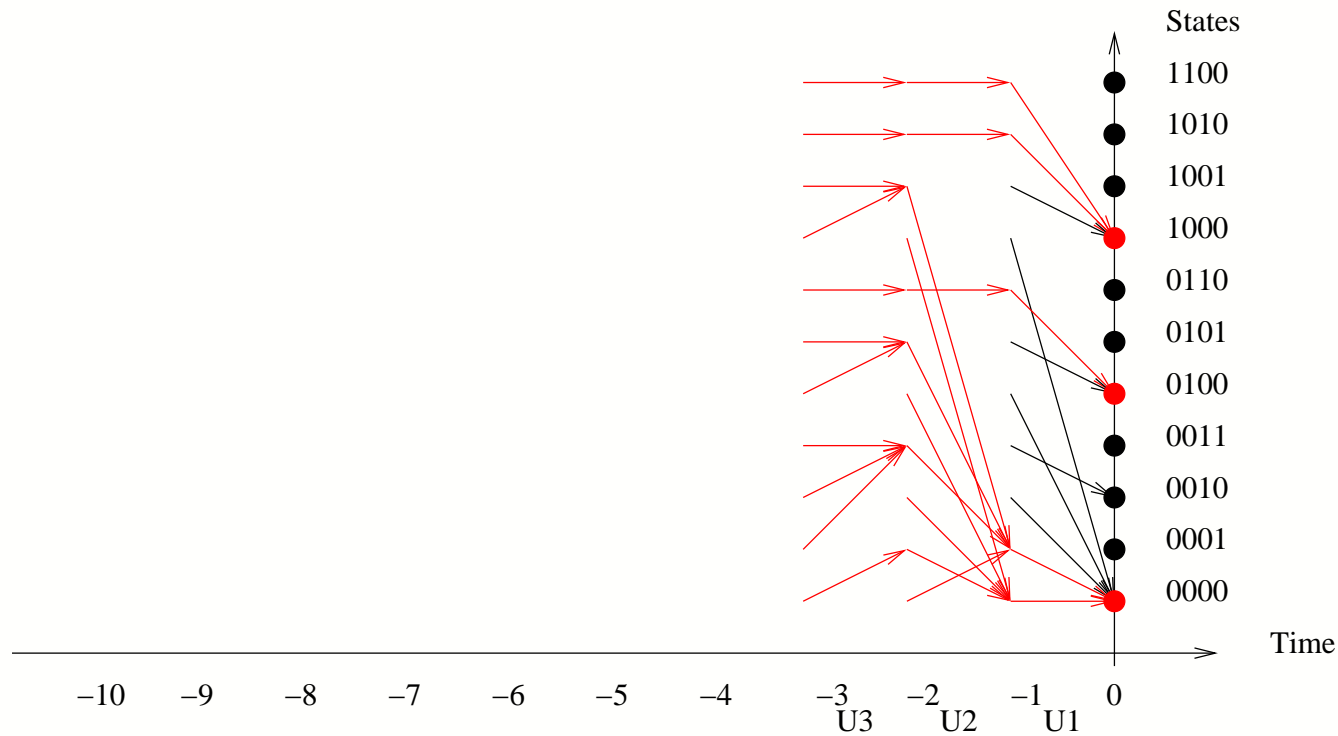
$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$



# Backward simulation example

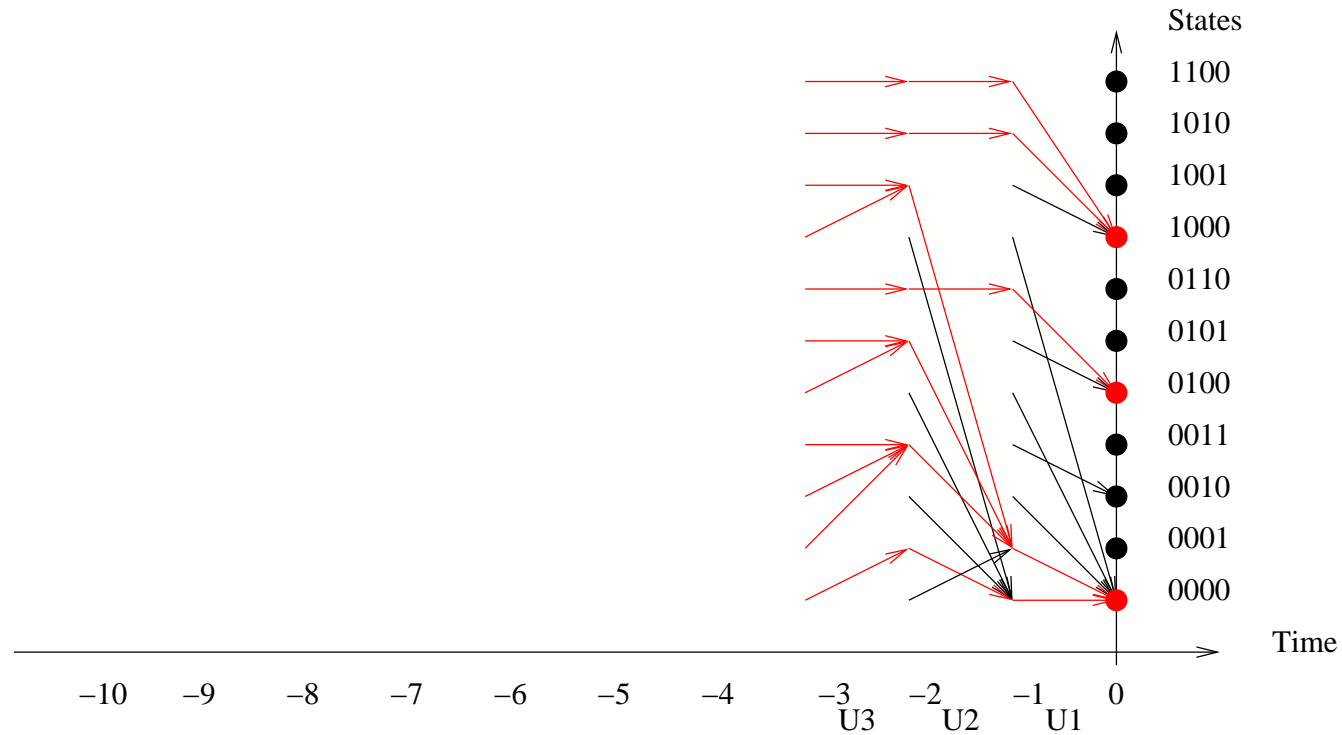


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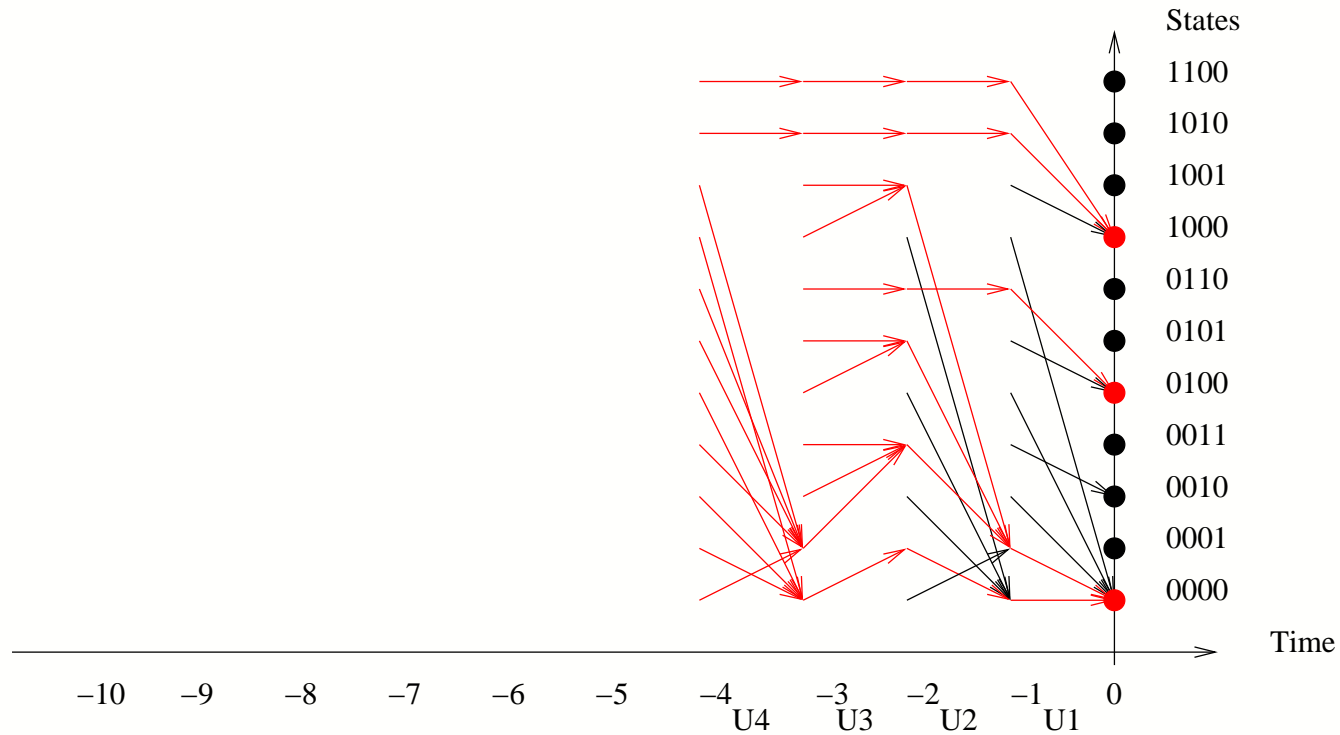
$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_3 = \{0000, 0100, 1000\}$$

# Backward simulation example



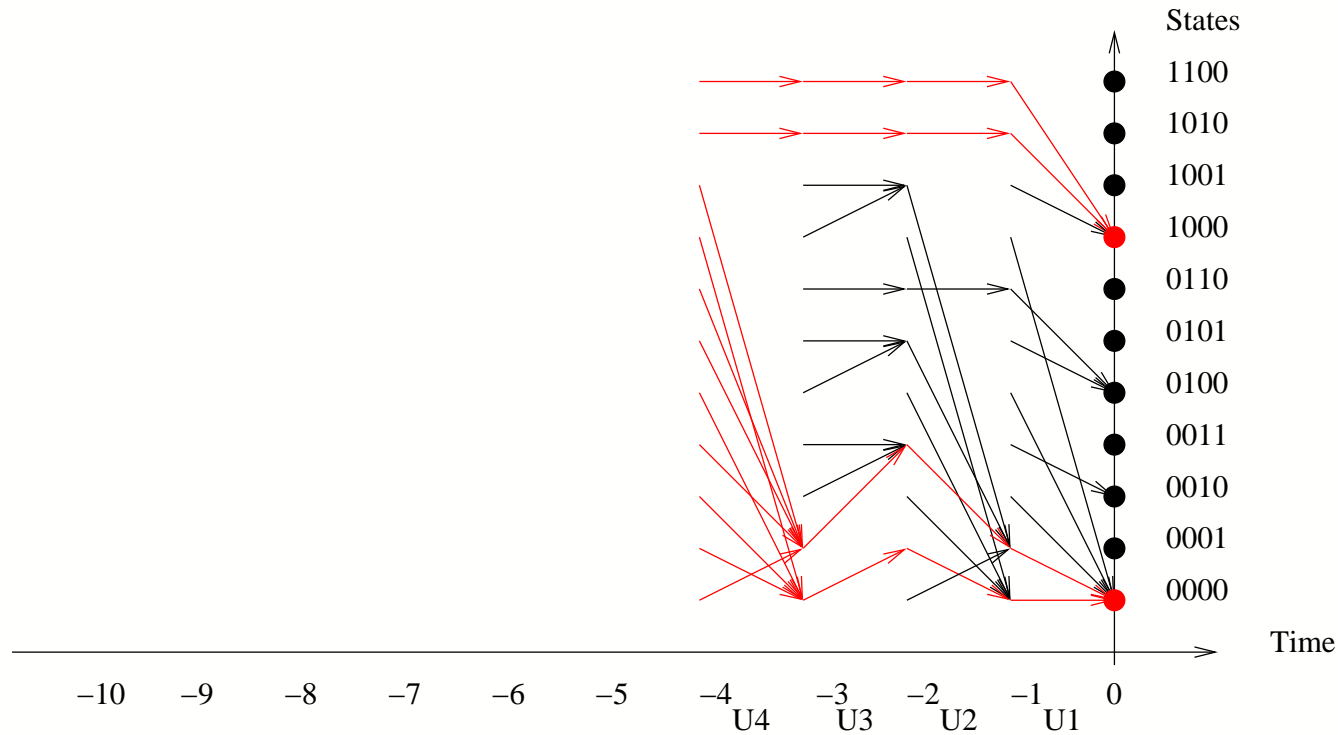
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# Backward simulation example



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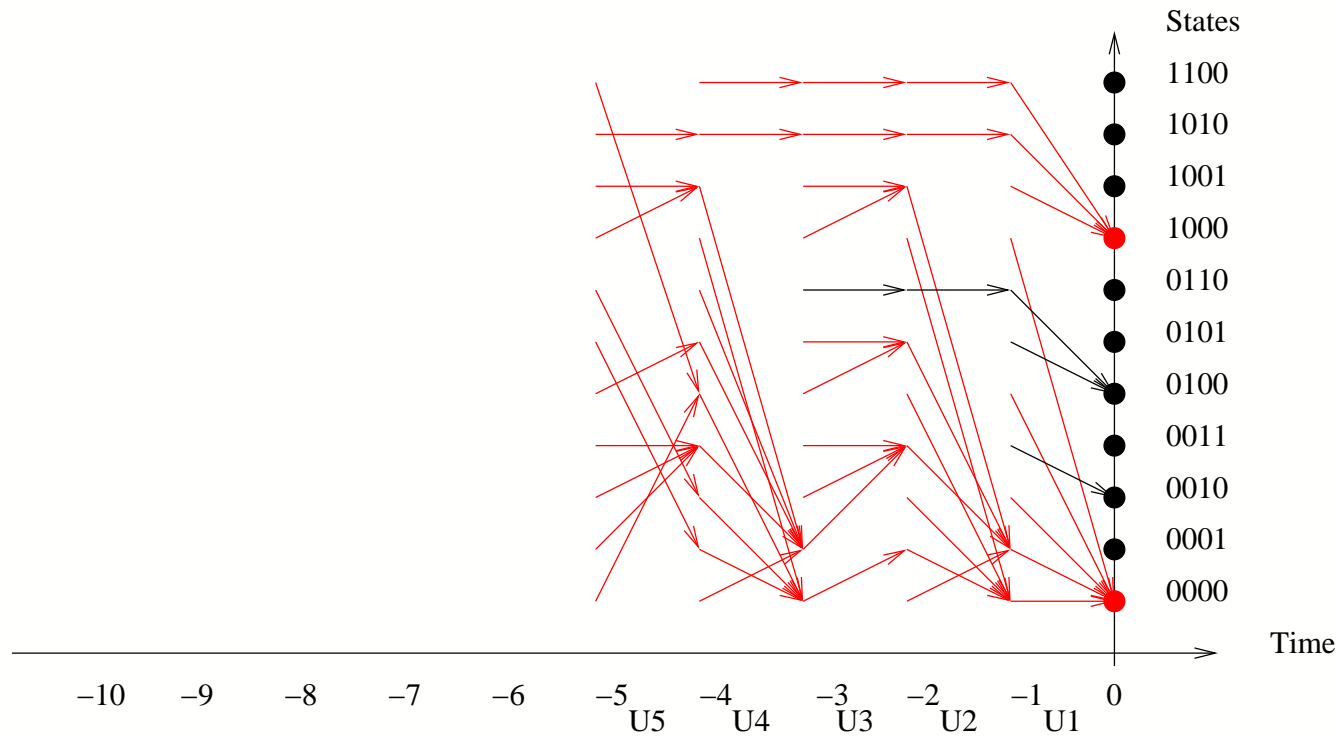
$$\mathcal{Z}_3 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

$$\mathcal{Z}_4 = \{0000, 1000\}$$

$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$

# Backward simulation example



$$\mathcal{Z}_0 = \mathcal{X}$$

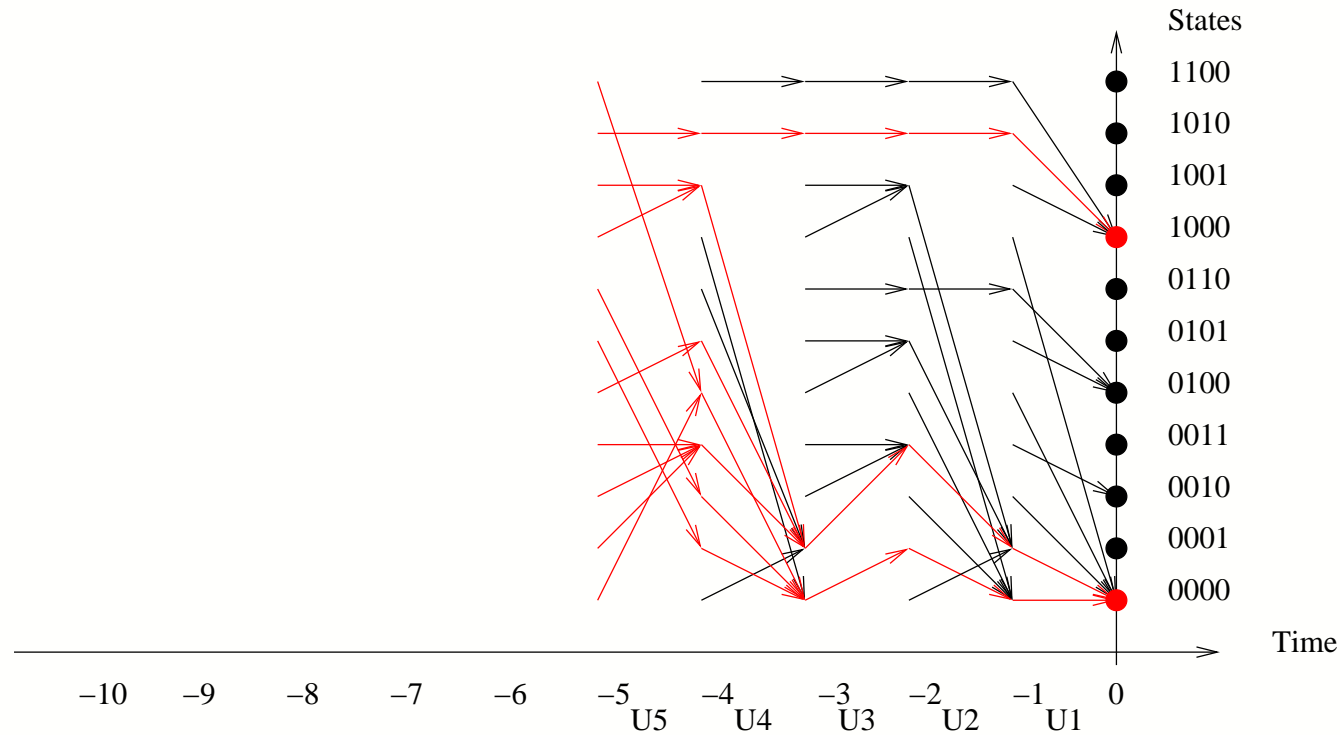
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# Backward simulation example



$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

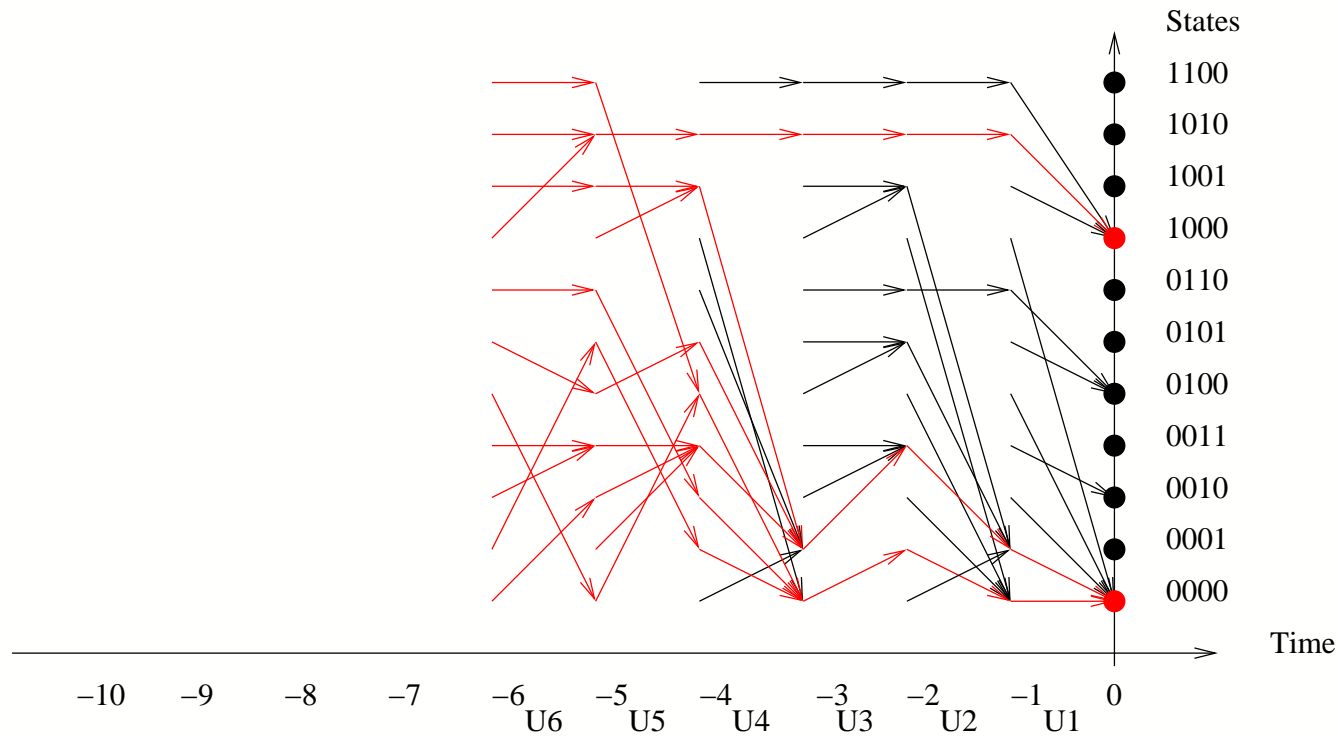
$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_3 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_4 = \{0000, 1000\}$$

$$\mathcal{Z}_5 = \{0000, 1000\}$$

# Backward simulation example



$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$

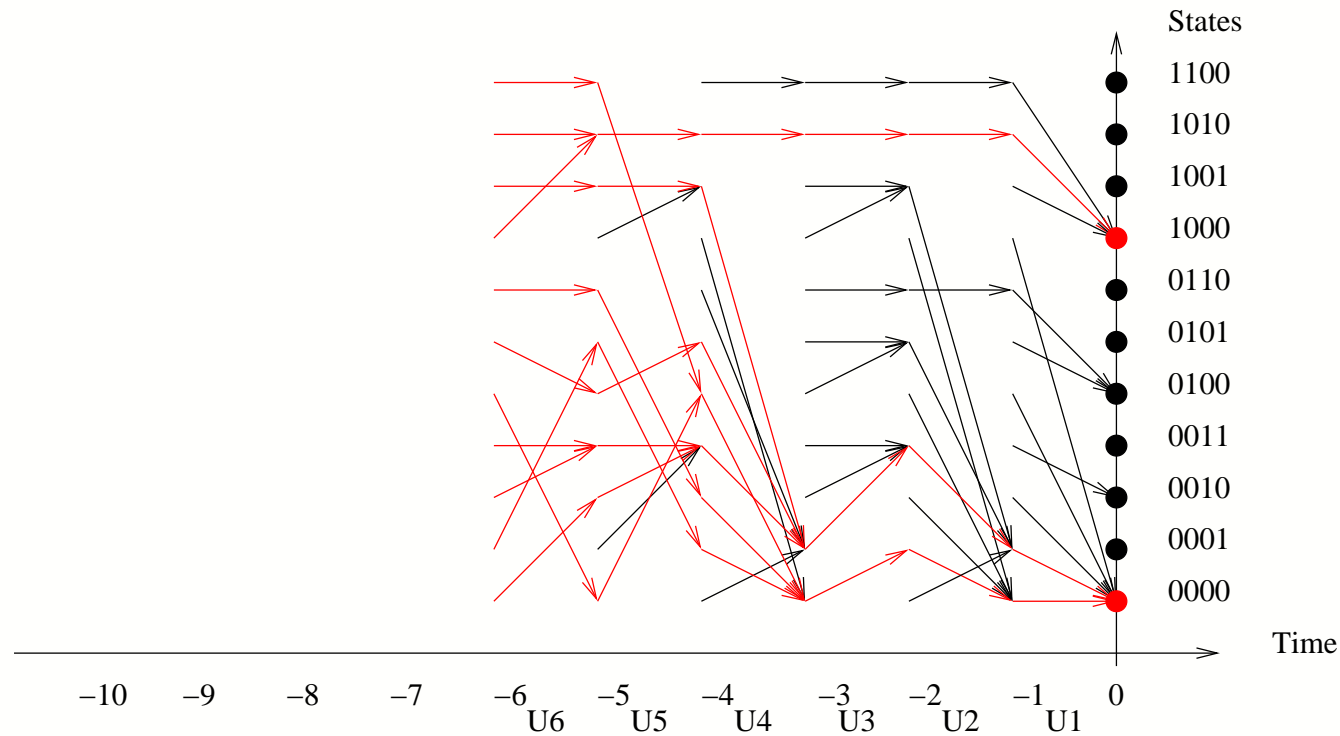
$$\mathcal{Z}_3 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_4 = \{0000, 1000\}$$

$$\mathcal{Z}_5 = \{0000, 1000\}$$



# Backward simulation example



$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_3 = \{0000, 0100, 1000\}$$

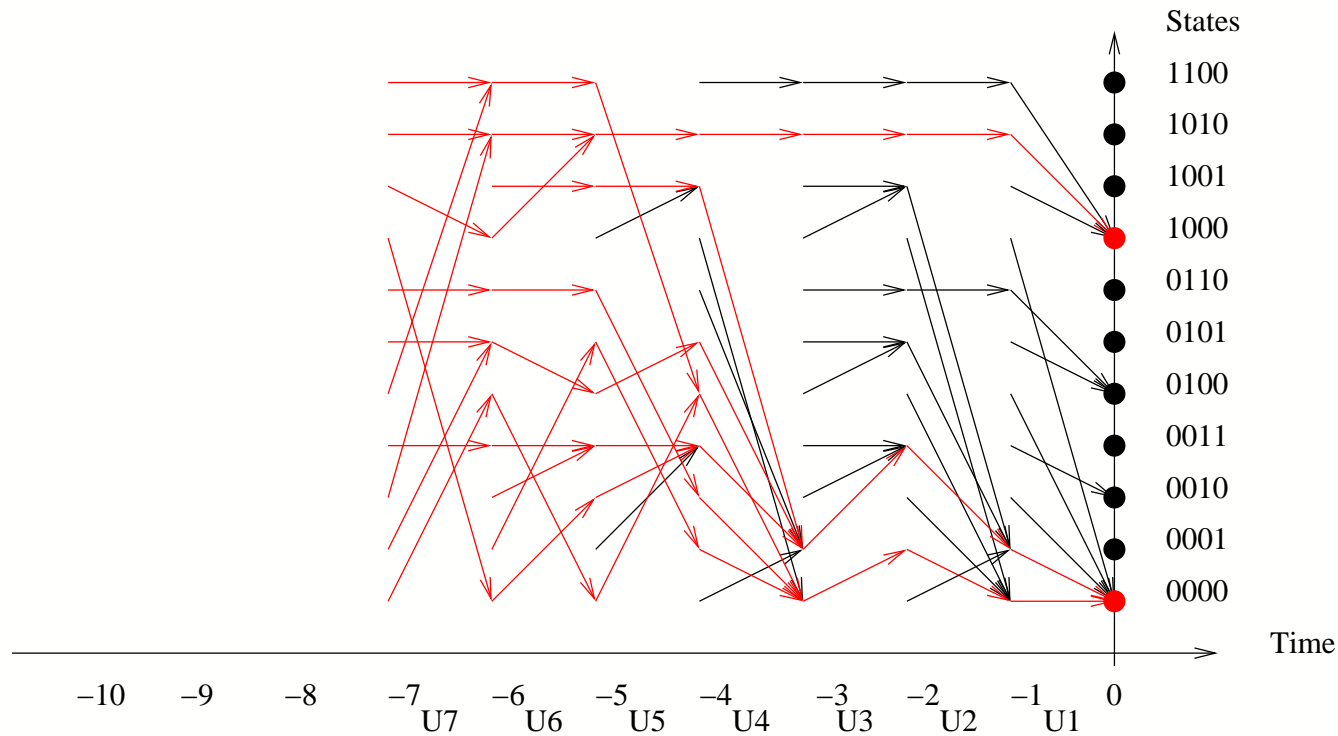
$$\mathcal{Z}_4 = \{0000, 1000\}$$

$$\mathcal{Z}_5 = \{0000, 1000\}$$

$$\mathcal{Z}_6 = \{0000, 1000\}$$



# Backward simulation example



$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$

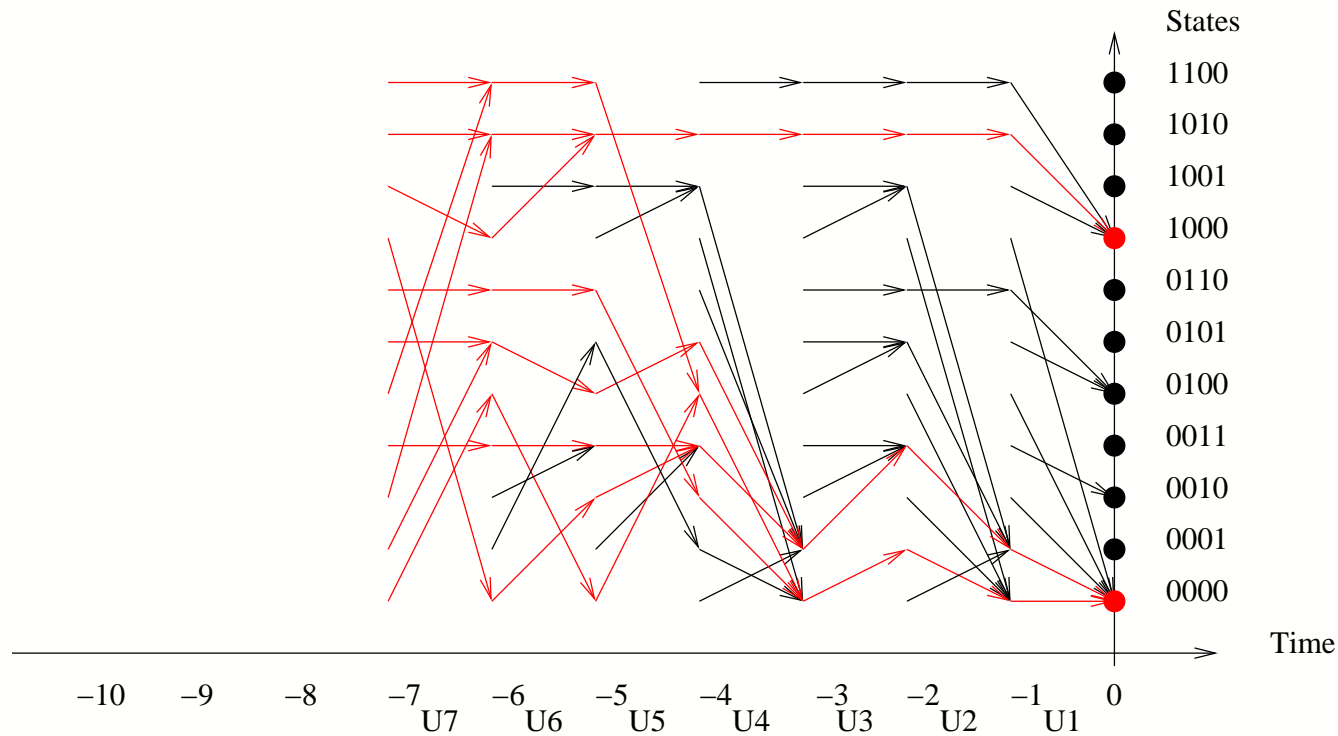
$$\mathcal{Z}_3 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_4 = \{0000, 1000\}$$

$$\mathcal{Z}_5 = \{0000, 1000\}$$

$$\mathcal{Z}_6 = \{0000, 1000\}$$

# Backward simulation example



$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_3 = \{0000, 0100, 1000\}$$

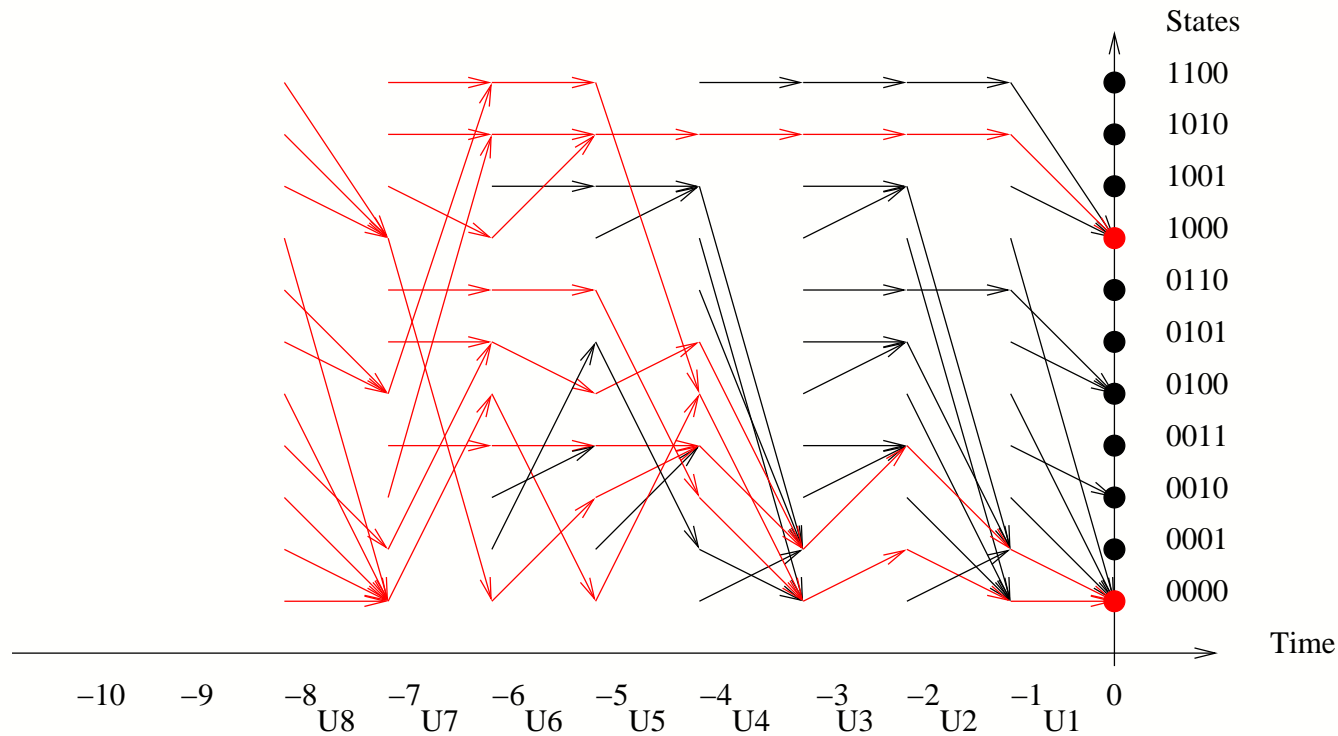
$$\mathcal{Z}_4 = \{0000, 1000\}$$

$$\mathcal{Z}_5 = \{0000, 1000\}$$

$$\mathcal{Z}_6 = \{0000, 1000\}$$

$$\mathcal{Z}_7 = \{0000, 1000\}$$

# Backward simulation example



$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_3 = \{0000, 0100, 1000\}$$

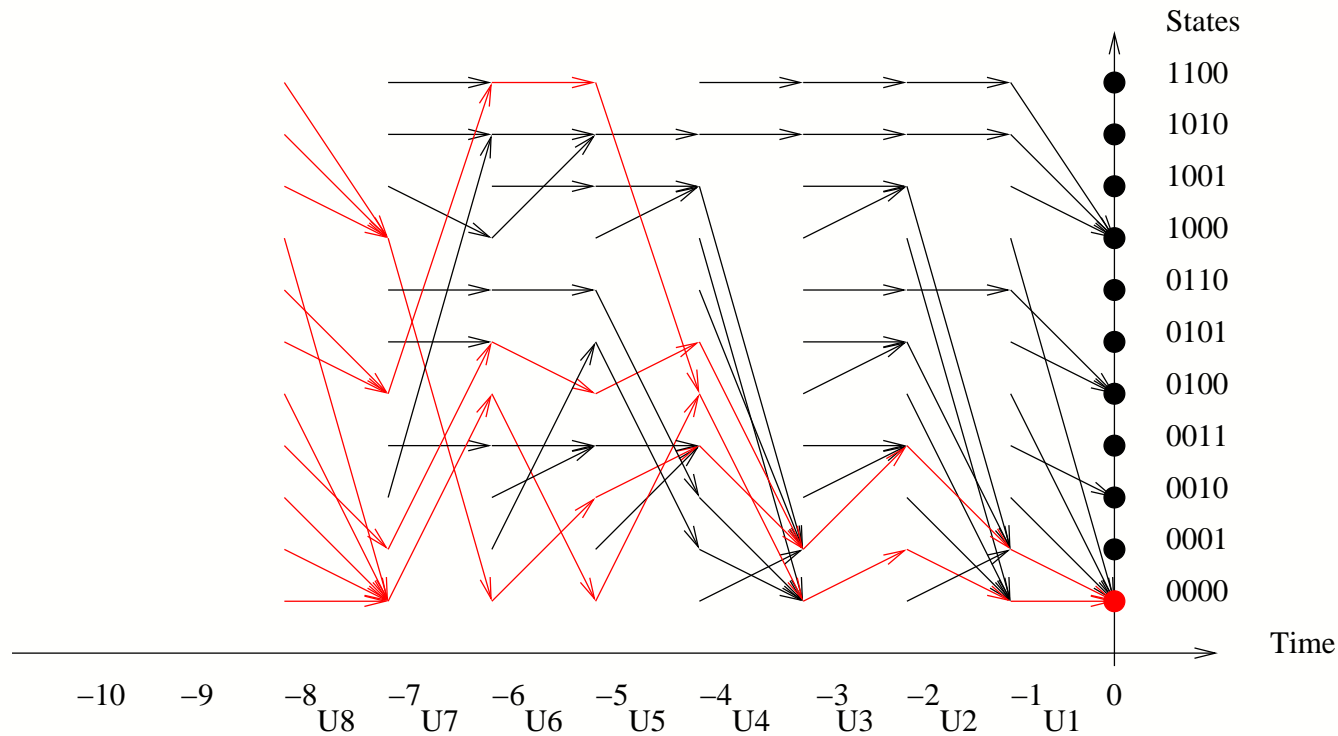
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# Backward simulation example



$$\mathcal{Z}_0 = \mathcal{X}$$

$$\mathcal{Z}_1 = \{0000, 0010, 0100, 1000\}$$

$$\mathcal{Z}_2 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_3 = \{0000, 0100, 1000\}$$

$$\mathcal{Z}_4 = \{0000, 1000\}$$

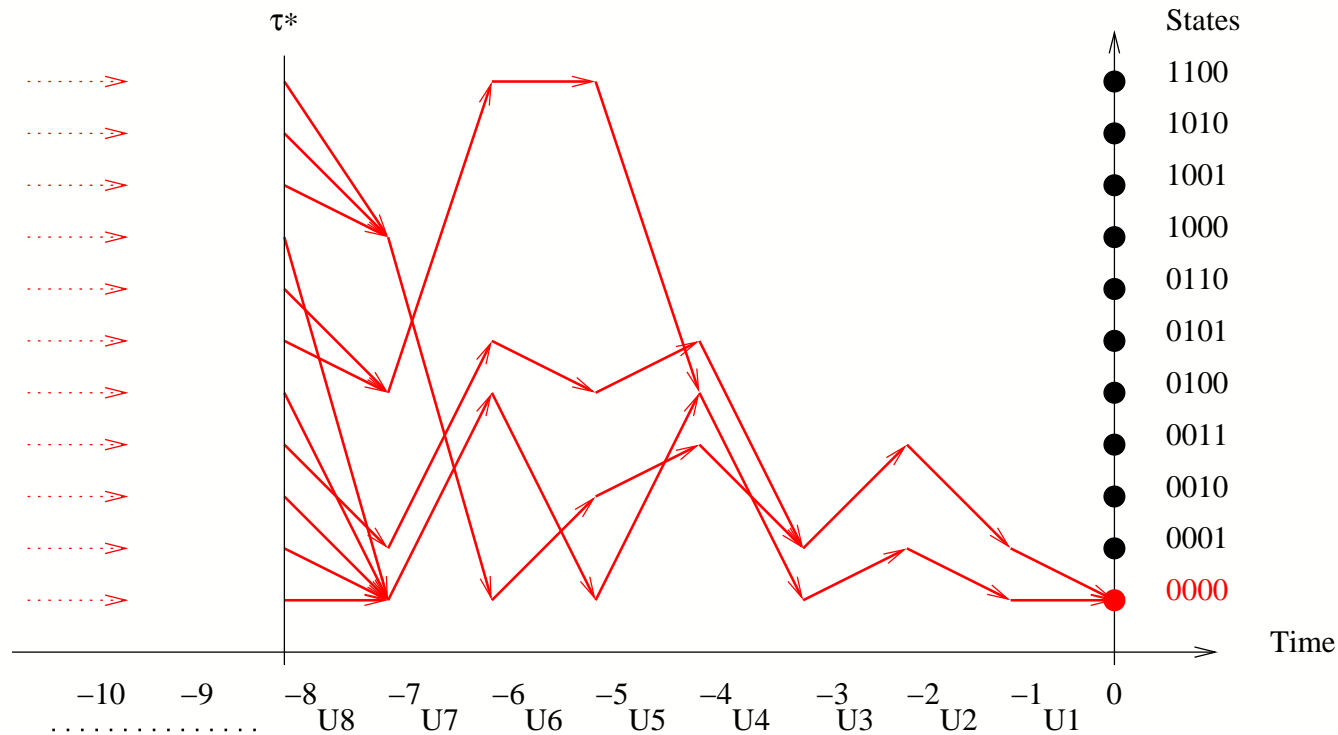
$$\mathcal{Z}_5 = \{0000, 1000\}$$

$$\mathcal{Z}_6 = \{0000, 1000\}$$

$$\mathcal{Z}_7 = \{0000, 1000\}$$

$$\mathcal{Z}_8 = \{0000\}$$

# Backward simulation example



Process stops when  $|\mathcal{Z}_n| = 1$

Stopping time  $\tau^* = 8$

Number of computation of  $\Phi$  (complexity) :  $n \cdot \tau^*$

# Backward coupling simulation

**Proposition 1 (Propp & Wilson)** *If  $\tau^* < +\infty$  a.s. then the returned value is stationary distributed.*

Proof :

$\{Z_n\}_{n \in \mathbb{N}}$  is non-increasing and constant for  $n$  sufficiently large

$$\Phi(\Phi(\dots(\Phi(\mathcal{X}, U_{-n+1}), \dots), U_{-1}), U_0) \stackrel{\mathcal{L}}{\sim} \Phi(\Phi(\dots(\Phi(\mathcal{X}, U_1), \dots), U_{n-1}), U_n)$$

Remarks :

Stopping times  $\tau$  and  $\tau^*$  have the same law,

$\tau$  depends on  $\Phi$  coding (not only on the transition matrix !)

$\Rightarrow$  optimization problem

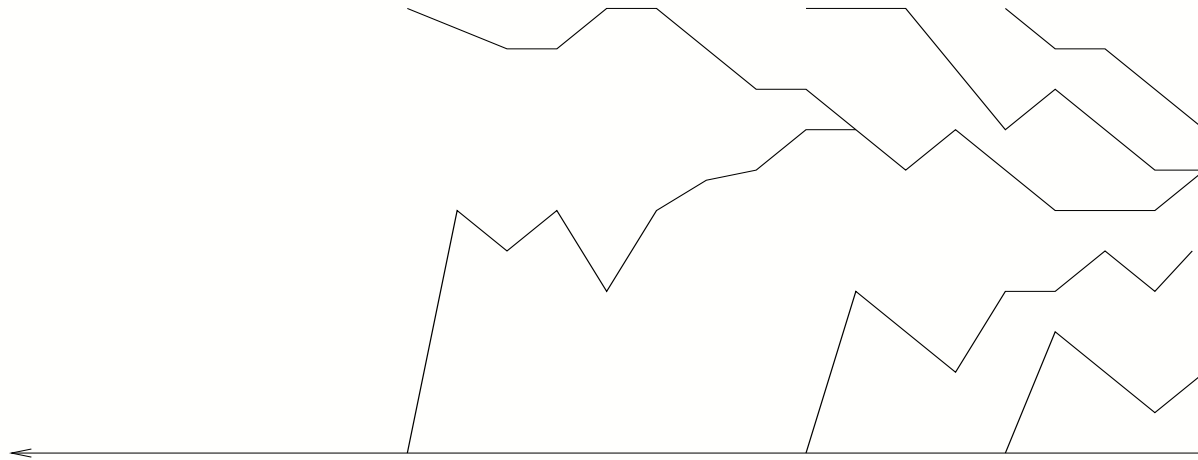


# Partially ordered state space

$\mathcal{X}$  is partially ordered :

- set  $M$  of maximum
- set  $m$  of minimum
- example : vector and componentwise comparison

The global scheme is



Propp & Wilson : double period of simulation

Fill : interruptible forward algorithm



# Monotonous Perfect

## MONOTONICITY:

Propp & Wilson(1996)

- reverse time
- run parallel trajectories  $|M| + |m|$
- wait for coupling.

$$\mathcal{Z}_n = \Phi(\Phi(\dots(\Phi(M \cup m, U_{-n+1}), \dots), U_{-1}), U_0).$$

potential set of reachable states from maxima  
and minima at time  $n$ .

```
n=0;
repeat
  n=n+1;
  R[n]=Random;
  for all  $x \in M \cup m$  do
     $y(x) \leftarrow x$ 
  end for
  for i=n downto 1 do
    for all  $x \in M \cup m$  do
       $y(x) \leftarrow \Phi(y(x), R[i])$ 
    end for
  end for
until All  $y(x)$  are equal
return  $y(x)$ 
```





# Monotonous Perfect

## MONOTONICITY:

Propp & Wilson(1996)

- reverse time
- run parallel trajectories  $|M| + |m|$
- wait for coupling.

$$\mathcal{Z}_n = \Phi(\Phi(\dots(\Phi(M \cup m, U_{-n+1}), \dots), U_{-1}), U_0).$$

potential set of reachable states from maxima and minima at time  $n$ .

Optimal coupling time : complexity  $\mathcal{O}(\mathbb{E}\tau)$

Storage of random sequence

```
n=1
repeat
  n=2n;
  for all  $x \in M \cup m$  do
     $y(x) \leftarrow x$ 
  end for
  for i=n downto n/2 do
    R[i]=Random;
  end for
  for i=n downto 0 do
    for all  $x \in M \cup m$  do
       $y(x) \leftarrow \Phi(y(x), R[i])$ 
    end for
  end for
until All  $y(x)$  are equal
return  $y(x)$ 
```



# Monotone routing (1)

Queueing network

- $K$  queues, capacity  $C_i$  for queue  $i$
- State space  $\mathcal{X} = \{0, \dots, C_1\} \times \dots \times \{0, \dots, C_K\}$
- event  $e$

Routing strategy (overflow):

- origin queue  $i$
- destination list  $j_1, \dots, j_k$
- rate  $\lambda_e$

according to a state  $x$  event  $e$  route a customer from queue  $i$  to the first non-full queue in the list of destinations. If all destinations are full the customer is routed out of the network (rejection).

**Proposition 2** *The event  $e$  routing with rejection is monotonous.*

$$x \leq y \Rightarrow \Phi(x, e) \leq \Phi(y, e)$$

- If queue  $i$  is empty, nothing is done
- Arrival is a routing with overflow



# Monotone routing (2)

Routing strategy (blocking):

- origin queue  $i$
- destination list  $j_1, \dots, j_k$
- rate  $\lambda_e$

according to a state  $x$  event  $e$  route a customer from queue  $i$  to the first non-full queue in the list of destinations. If all destinations are full the customer stay in its queue and run its service again.

**Proposition 3** *The event  $e$  routing with blocking is monotonous.*

$$x \leq y \Rightarrow \Phi(x, e) \leq \Phi(y, e)$$

- If queue  $i$  is empty, nothing is done
- destination list  $j_1, \dots, j_k, i$



# Monotone routing (3)

Routing strategy (JSQ):

- origin queue  $i$
- destination list  $j_1, \dots, j_k$
- rate  $\lambda_e$

according to a state  $x$  event  $e$  route a customer from queue  $i$  to the lowest non-full queue in the list of destinations. If all destinations are full the customer stay in its queue and run its service again or is rejected from the network.

**Proposition 4** *The event  $e$  routing with JSQ policy is monotonous.*

$$x \leq y \Rightarrow \Phi(x, e) \leq \Phi(y, e)$$

- If queue  $i$  is empty, nothing is done



# Uniformization

Each event is driven by a Poisson process

-  $\lambda_j$  rate of event  $e_j$

Define  $\Lambda = \sum \lambda_i$  the uniformized process with rate  $\Lambda$

With probability  $\frac{\lambda_i}{\Lambda}$  event  $e_i$  occurs, if  $e_i$  is not admissible the transition is skipped.

**Theorem 1** *Markovian queueing networks with monotone routing policies have an uniformized version which is monotonous.*

⇒ Monotonous Perfect Simulation



# General architecture of $\Psi 2$

Model description specification : textual file

Coupling function : C file (optional)

Simulation parameters : textual file

Unix-like command : `psi2_unix -i model.txt -o result.txt -p param.txt -c couplage.c`



# Example queue : model

```
#nombre_de_files
1
#capacite_des_files
10
#etat_initial_mini_des_files
0
#etat_initial_maxi_des_files
10
#nombre_evenements
2
#tableau_des_evenements
#evt_id-evt_typ-taux-nb_fi_evt-origine-destin1-destin2-destina3
0      0      0.4    3     -1      0      -1
1      0      1.7    2      0      -1
```



# Example queue : param

```
#nombre_echantillons  
10000  
#taille_trajectoire_maxi  
100000000  
#germe_generateur  
5
```





# Example queue : result

```
# numero d'echantillon ECHN
# nombre d'iterations NBITER
# etat final sup de la file de numero n EFSFN
# etat final inf de la file de numero n EFIFN?
#ECHN:  NBITER:  EFSFN:  0  EFIFN:  0
#=====
      0      5      1      1
      1      5      1      1
      2      5      0      0
      3      4      0      0
      4      5      0      0
# taille      5  duree d'un tirage : 136.200000 micro-secondes
# valeur initiale du randomize 5
```



# Validation

Analytical models :

- single queue
- erlang models

⇒ adequation statistical tests  $\chi^2$

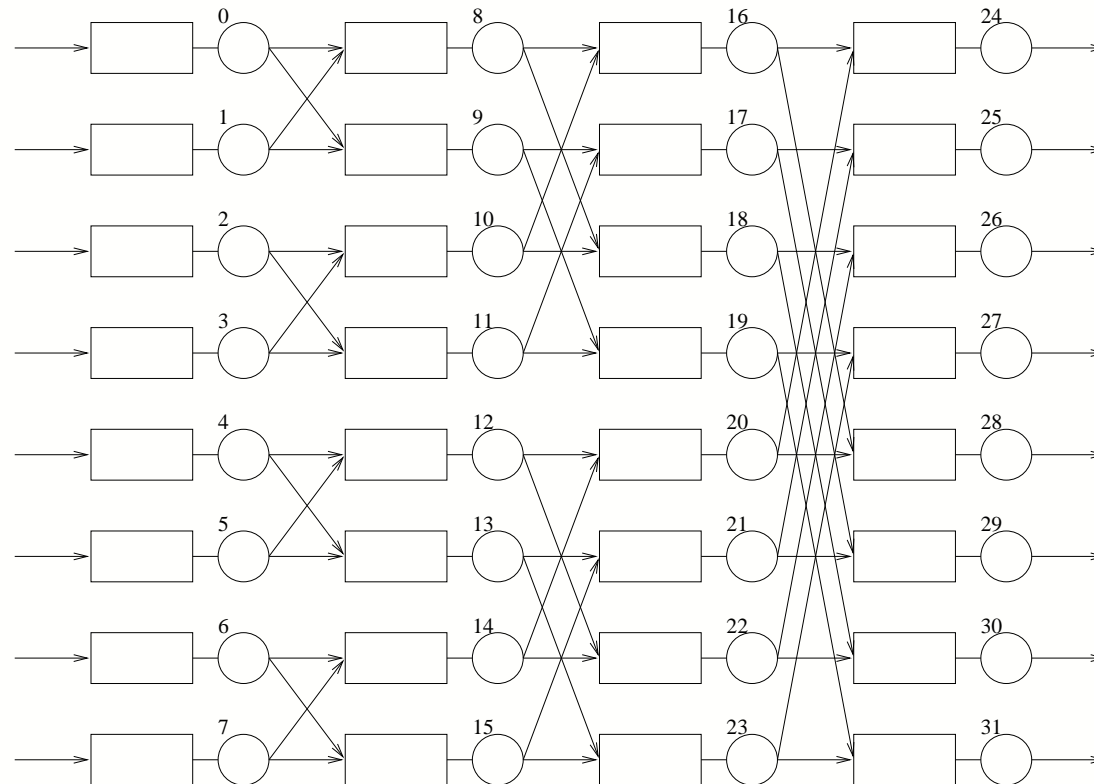


# Demo



# Interconnexion networks

## Delta $8 \times 8$ network, homogeneous



# Interconnexion networks

## Results

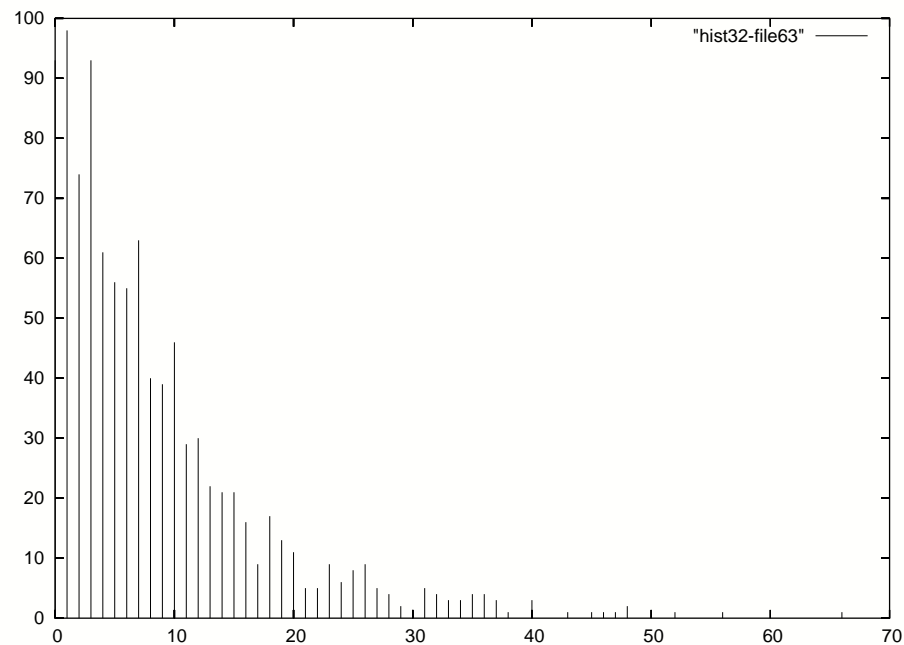
# taille 1000 duree d'un tirage : 135911.287000 micro-secondes

# valeur initiale du randomize 5

Coupling time :  $2^{17}$  with probability 0.45 and  $2^{18}$  with probability 0.55

Marginal distributions

Queue 63



# Conclusion (1)

- Theoretical results :
  - reverse scheme + contracting operator
  - coupling condition
  - monotonous transitions
  - functional reduction
- Algorithmic results:
  - generic representation of QN
  - guaranteed coupling algorithm
  - compact representation
- Experimental results
  - complexity reduction
  - significant results (depending on the diameter and the coding of the network and capacities)



# Conclusion (2)

Software tool: **PSI 2 : Perfect Simulator**

<http://www-id.imag.fr/Software/PSI2/>

unix command / with a simple interface

```
psi2_unix -i example.txt -p param.txt -c cost.c -o example.out
```

example.cost associates to each state its cost

generates samples of costs stationary distributed (example.sample)



# Future works

- Theoretical improvements:
  - deeper understanding of  $\Phi$  properties and the spectrum of the transition matrix
  - evaluation or bounds on the coupling time
- Algorithmic perspectives:
  - storage of random sequences,
  - memory utilization,
  - parallelization
- Model based approach :
  - generalization to other modelling frameworks (QN, SAN, GSPN, PA,...)
  - non-monotone events
  - model properties : monotonicity, reversibility,...
- Experimental results
  - find limit models
  - significant results (estimation of rare events probability)
  - more examples

