

Réseaux de Petri T-temporisés pour la résolution des problèmes d'ordonnancement

Control design with T-TPN models for scheduling problems

Dimitri Lefebvre

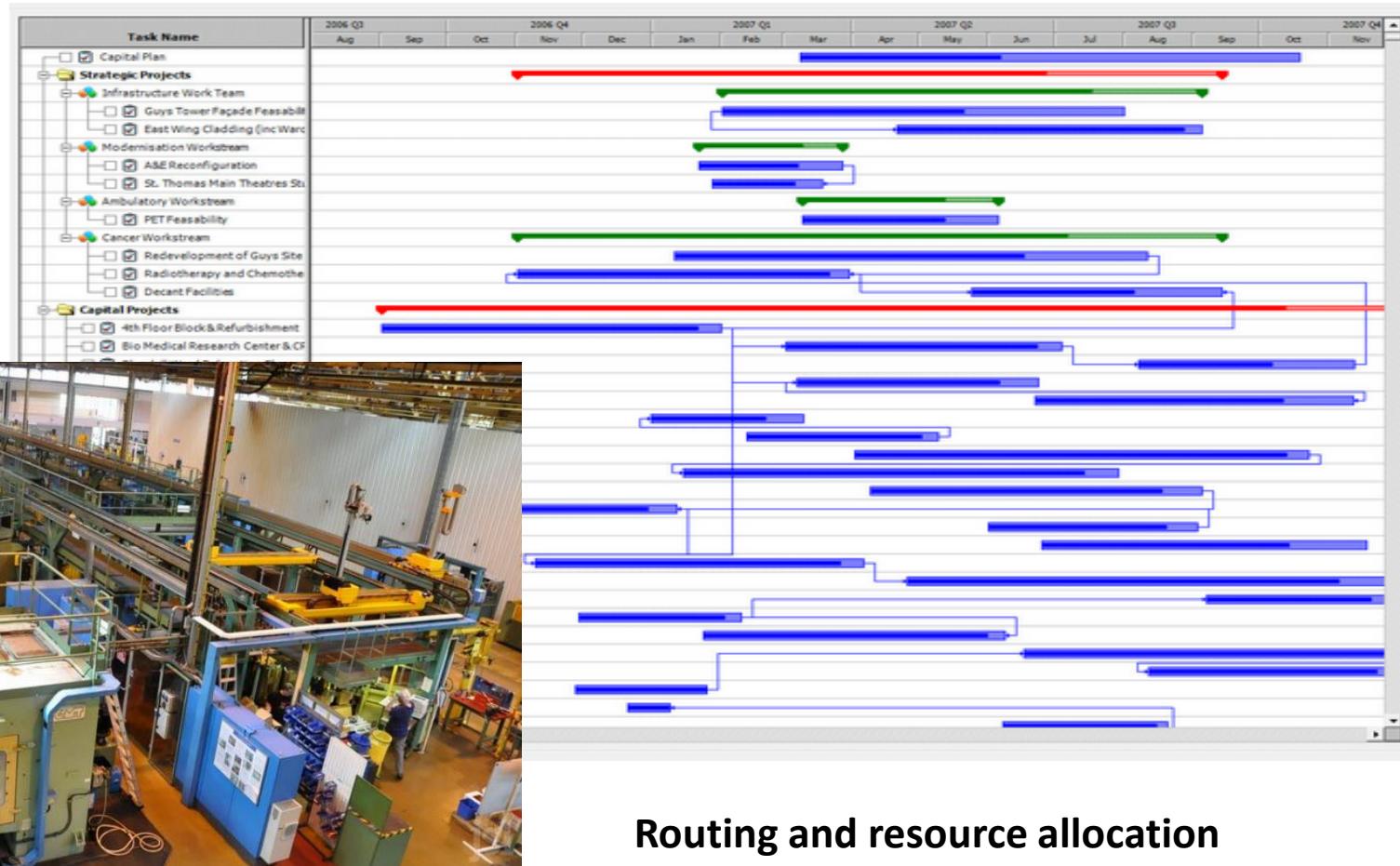
Normandie Université UNIHAVRE, GREAH, 76600 Le Havre, France.



Outline

- 1. Problems and objectives**
- 2. Modeling scheduling problems with T-TPN**
- 3. Timed Extended Reachability Graph**
- 4. Approximated Timed Extended Reachability Graph**
- 5. Beam Search**
- 6. Conclusion and future works**

Scheduling problems



Problems and objectives

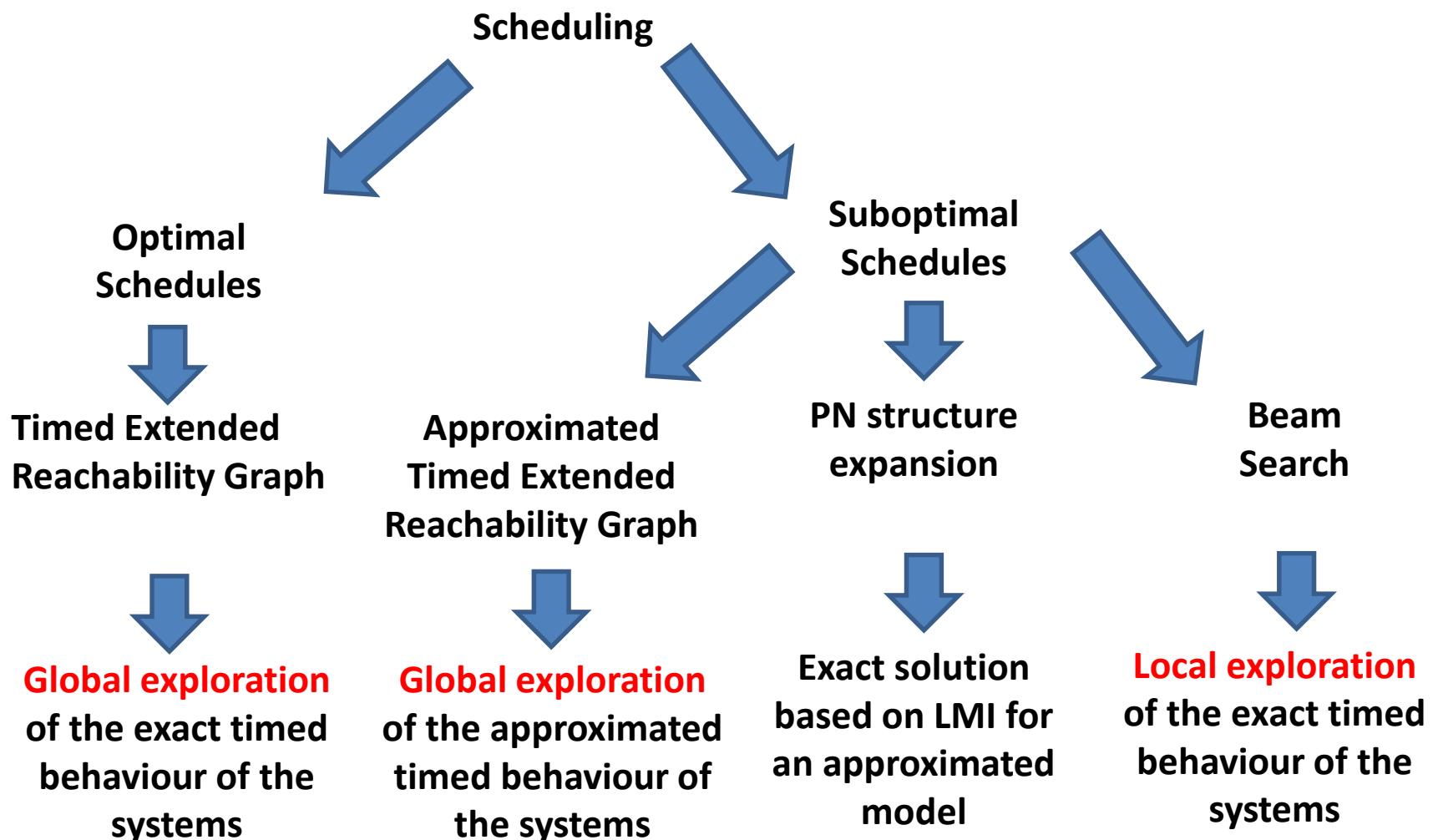
To compute control sequences that reach a reference state from an initial one for FMS in minimal time with :

- Limited resources
- State constraints
- Temporal constraints



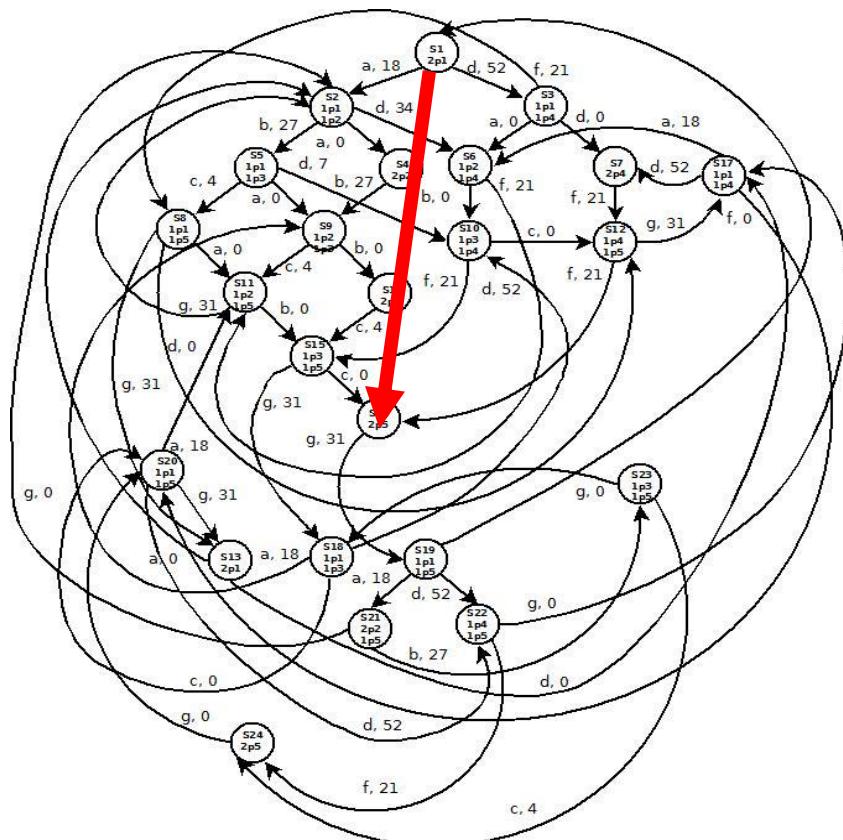
To compute firing sequences that reach a reference marking from an initial one for Transition – Timed Petri net models in minimal time with:

- Structural constraints (precedence)
- Marking constraints (deadlocks and dead branches avoidance,...)
- Minimal firing duration specifications

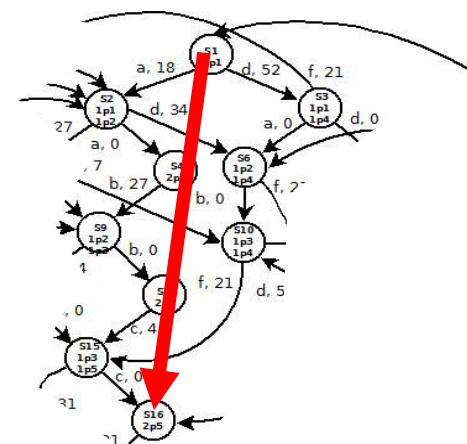


Scheduling problems

Global exploration



Local exploration



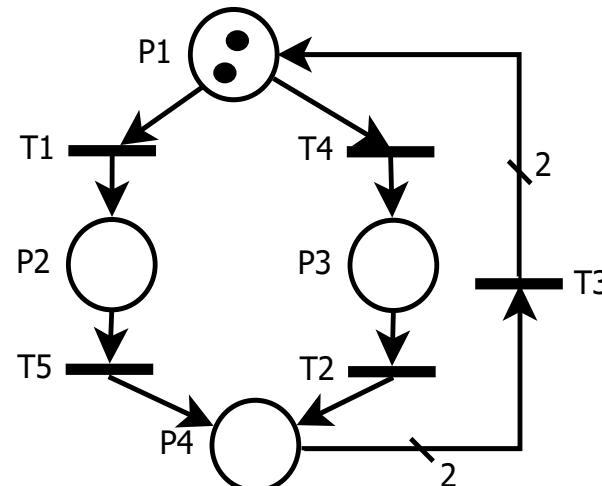
Outline

1. Problems and objectives
2. **Modeling scheduling problems with T-TPN**
3. Timed Extended Reachability Graph
4. Approximated Timed Extended Reachability Graph
5. Beam Search
6. Conclusion and future works

Petri nets

PN = $\langle P, T, W, M_i \rangle$

- **P : set of places**
- **T : set of transitions**
- **W: model structure**
- **M_i : initial marking**



State and dynamics

$$\begin{pmatrix} P_1 \\ P_2 \\ P_3 \\ P_4 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \\ 0 \\ 0 \end{pmatrix} + W \cdot \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{pmatrix} \begin{matrix} T_1 \\ T_2 \\ T_3 \\ T_4 \\ T_5 \end{matrix}$$

Incidence matrix

$$W_{PR} = \begin{pmatrix} 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 \end{pmatrix}$$

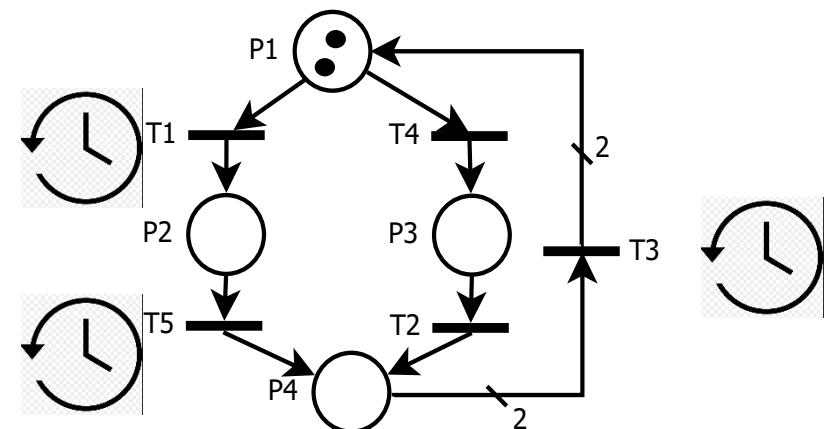
$$W_{PO} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 \end{pmatrix}$$

$$W = W_{PO} - W_{PR} = \begin{pmatrix} -1 & 0 & 2 & -1 & 0 \\ 1 & 0 & 0 & 0 & -1 \\ 0 & -1 & 0 & 1 & 0 \\ 0 & 1 & -2 & 0 & 1 \end{pmatrix}$$

Transition -Timed Petri nets (T-TPN)

TPN = $\langle P, T, W, d, M_0 \rangle$

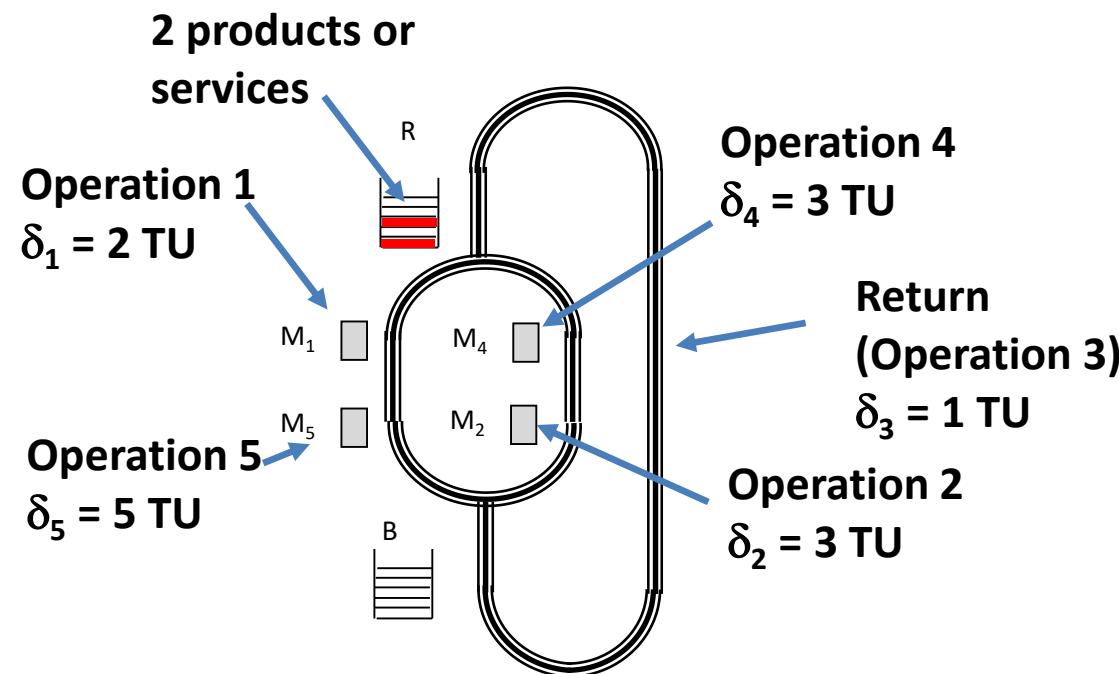
- **P : set of places**
- **T : set of transitions**
- **W: model structure**
- **M_0 : initial marking**
- **$\delta : T \rightarrow R^+$: minimal firing duration for transition T**
- **$\Delta : T \rightarrow R^+$: maximal firing duration for transition T**



Choice policy : preselection policy
Server policy : infinite or k -server policy
Memory policy : enabling memory policy

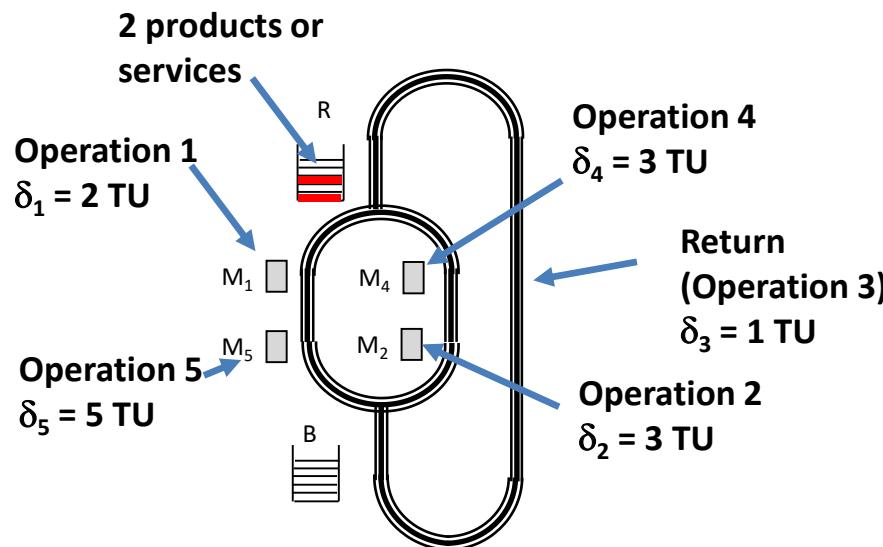
The firings occurs according to the earliest firing policy

Modeling manufacturing systems with MATLAB



Définir le modèle T-TPN avec MATLAB
>> [Wpr,Wpo,delta,MI] = Exemple1_MACS

Modeling manufacturing systems with MATLAB



Wpr =

$$\begin{matrix} 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 \end{matrix}$$

Wpo =

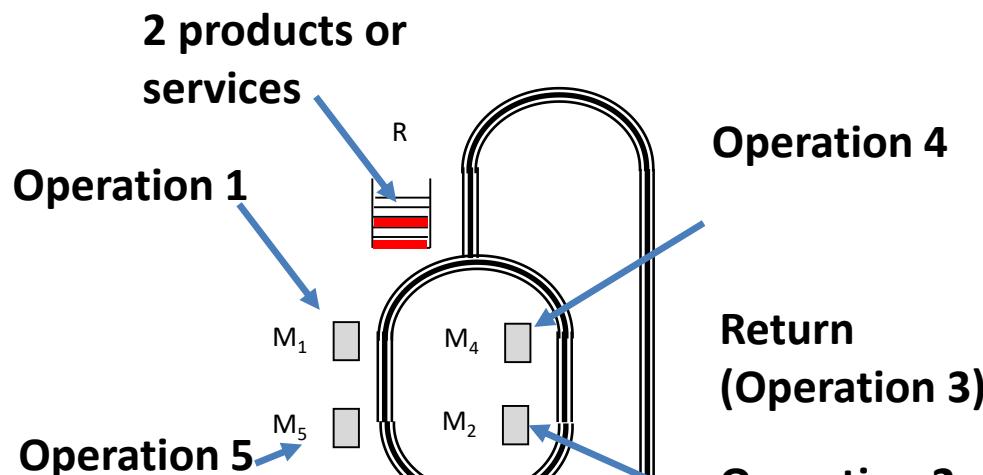
$$\begin{matrix} 0 & 0 & 2 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 \end{matrix}$$

delta = **MI =**

$$\begin{matrix} 2 & 2 \\ 3 & 0 \\ 1 & 0 \\ 3 & 0 \\ 5 & \end{matrix}$$

Définir le modèle T-TPN avec MATLAB

>> [Wpr,Wpo,delta,MI] = Exemple1_MACS

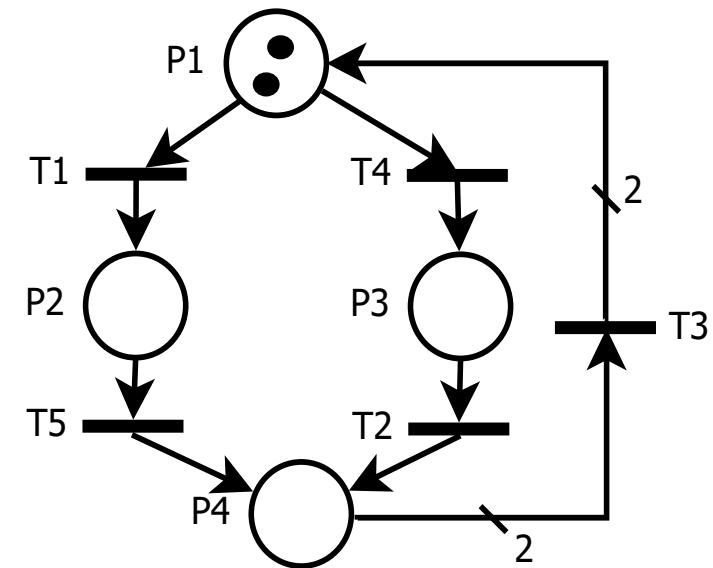


$$M_I = \begin{pmatrix} 2 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\delta = \begin{pmatrix} 1 \\ 2 \\ 1 \\ 2 \\ 2 \end{pmatrix}$$

Initial marking

Time parameters

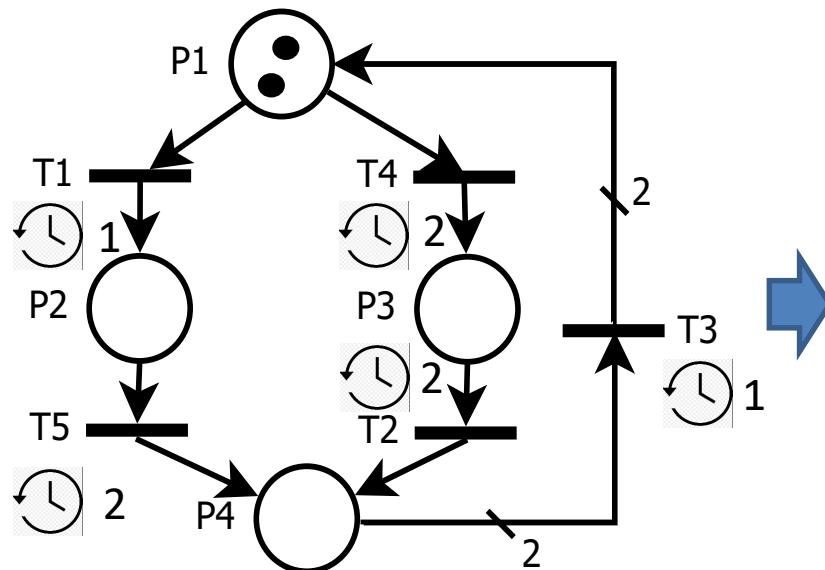


PN model

$$W = \begin{pmatrix} -1 & 0 & 2 & -1 & 0 \\ 1 & 0 & 0 & 0 & -1 \\ 0 & -1 & 0 & 1 & 0 \\ 0 & 1 & -2 & 0 & 1 \end{pmatrix}$$

Incidence matrix

T-TPN



Information about time



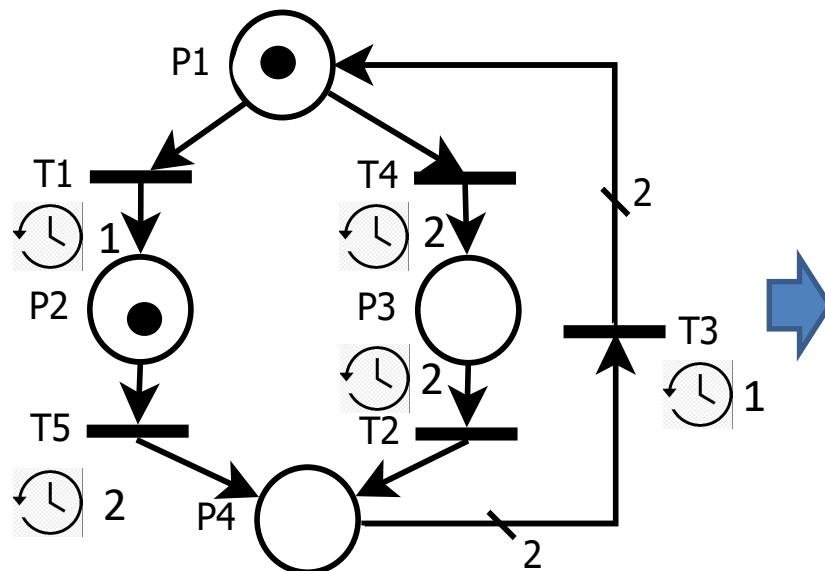
Agenda

Time: $t = 0$ TU

$T1 : t+1$ TU

$T4 : t+2$ TU

T-TPN



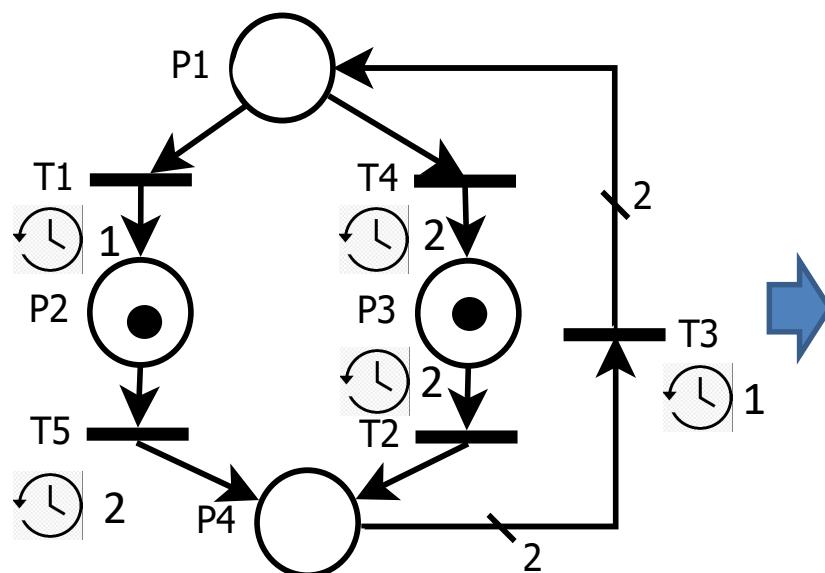
Information about time



$$\sigma = T_1(1)$$

- Agenda**
- Time:** $t = 1$ TU
- T1 :** $t+0$ TU
- T4 :** $t+1$ TU
- T5 :** $t+2$ TU

T-TPN



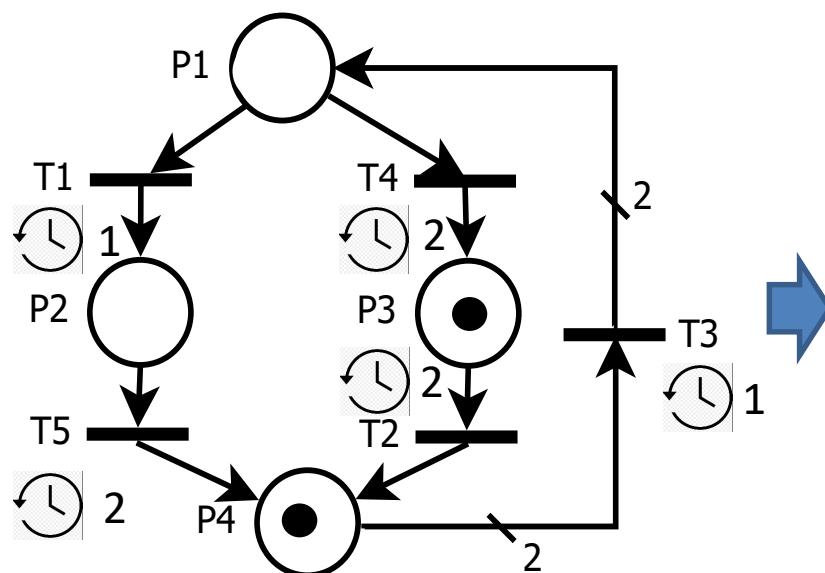
Information about time



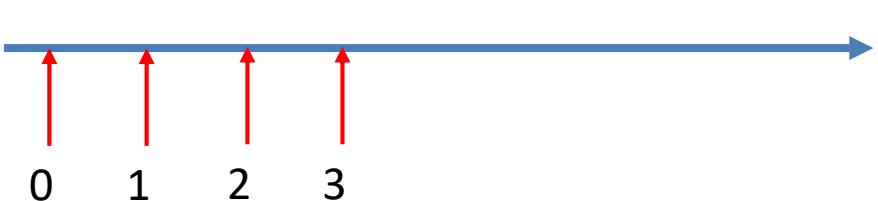
$$\sigma = T_1(1) \ T_4(2)$$

Agenda
Time: $t = 2$ TU
T5 : $t+1$ TU
T2 : $t+2$ TU

T-TPN



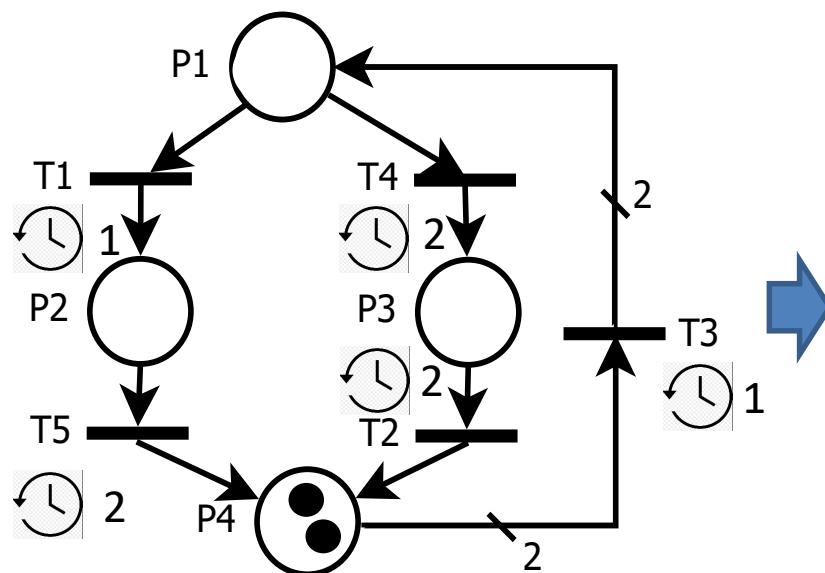
Information about time



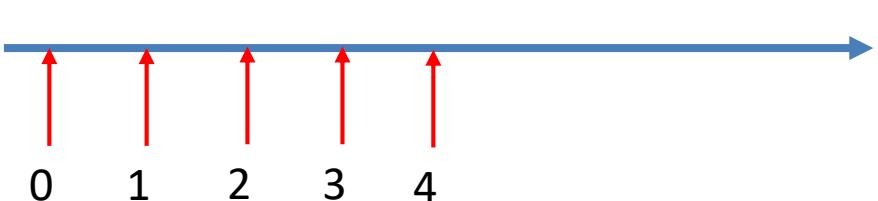
$$\sigma = T_1(1) T_4(2) T_5(3)$$

Agenda
Time: $t = 3$ TU
 $T2 : t+1$ TU

T-TPN



Information about time



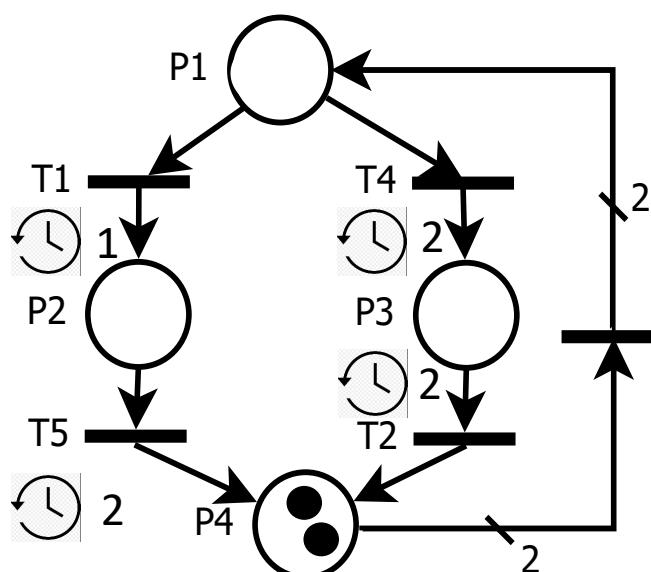
$$\sigma = T_1(1) T_4(2) T_5(3) T_2(4)$$

Agenda

Time: $t = 4$ TU

T3 : $t+1$ TU

T-TPN



Dater une séquence avec MATLAB

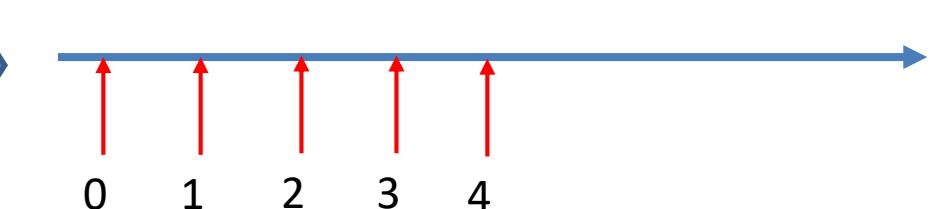
```
>> seq =[1 4 5 2]
```

```
>>[timed_seq]=dateur_MACS(Wpr,Wpo,delta,seq,MI)
```

```
>>timed_seq =
```

```
1 4 5 2
```

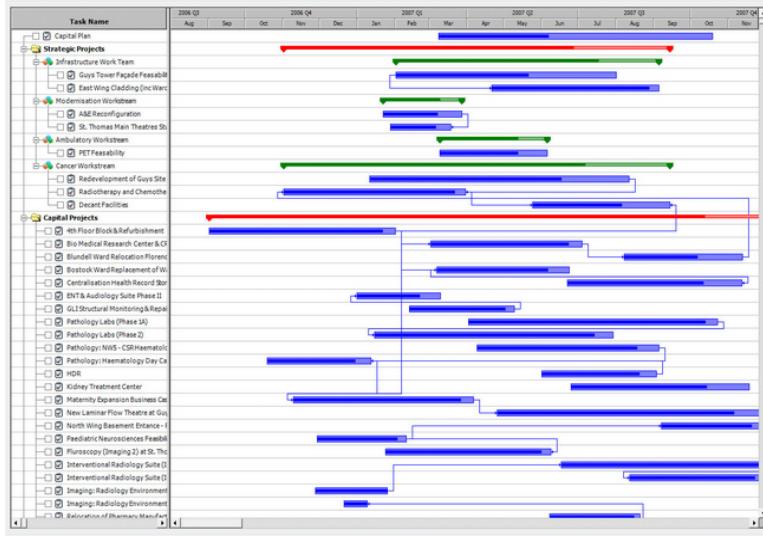
```
1 2 3 4
```



$$\sigma = T_1(1) T_4(2) T_5(3) T_2(4)$$

Modeling scheduling problem with T-TPN

Job Shop Scheduling problems

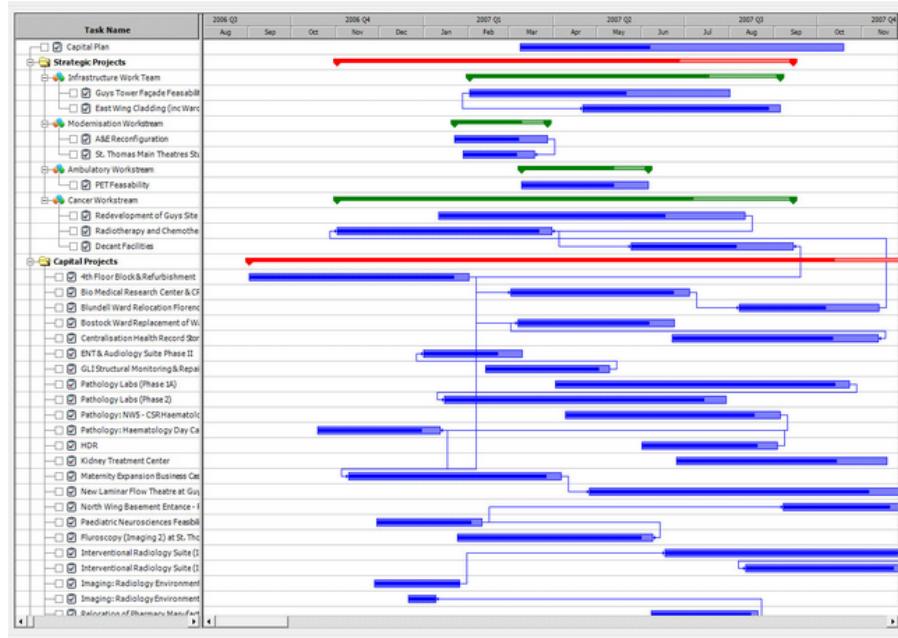


Problem specification

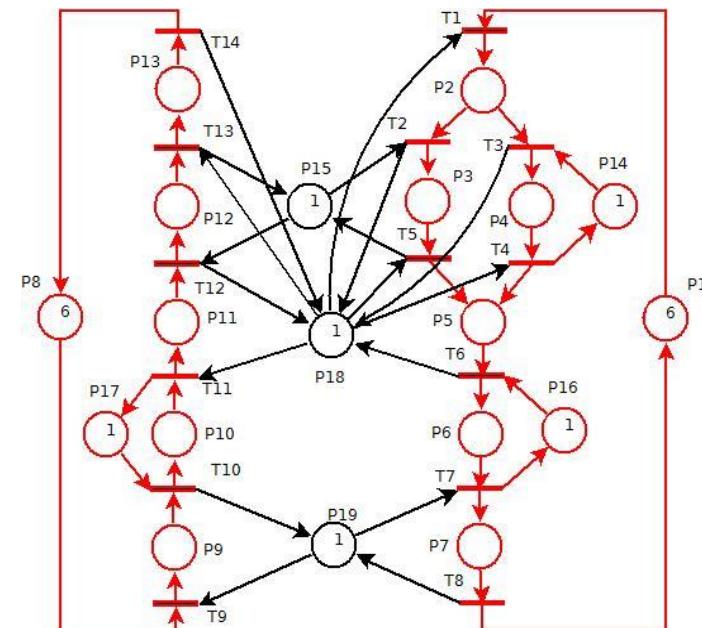
- Set of jobs J
- Each job J is composed by a set of operations o
- Each operation has a duration d and needs a set of resource R
- Resources are shared
- Set of services to be completed according to the jobs
- Each job can be single or multi server
- Each operation can be single or multi server

Modeling scheduling problem with T-TPN

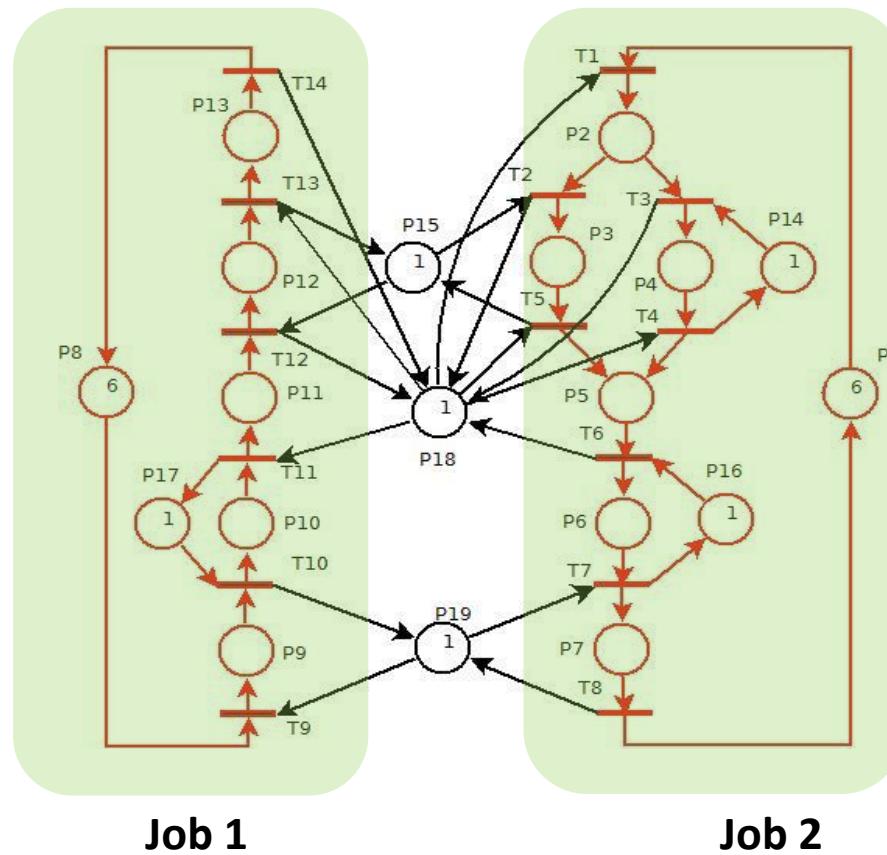
Job Shop Scheduling problems



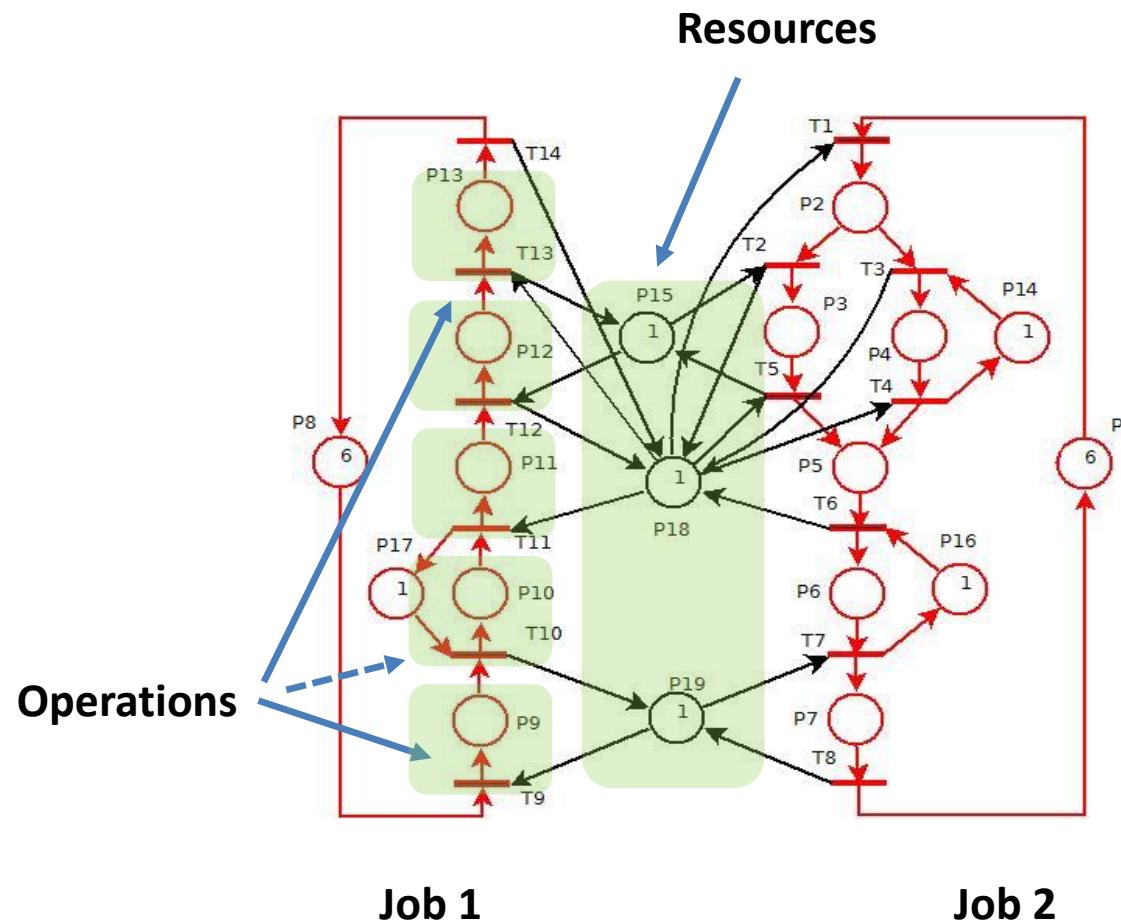
Petri net modeling



Scheduling problems



Scheduling problems



Exercice1

Construire le modèle T-TPN d'un problème d'ordonnancement avec:

2 jobs à réaliser chacun 1 fois

Capacité des jobs : 1

3 opérations o1, o2, o3 dans le job 1 de durée 5, 5 et 2

3 opérations o4, o5, o6 dans le job 2 de durée 5, 2 et 3

2 ressources partagée : pr1 partagée par o1, o2 et o5 et pr2 partagée par o2, o4 et o5

=> W_{PR} , W_{PO} , δ , M_i , ???

Exercice1

Construire le modèle T-TPN d'un problème d'ordonnancement avec:

2 jobs à réaliser chacun 1 fois

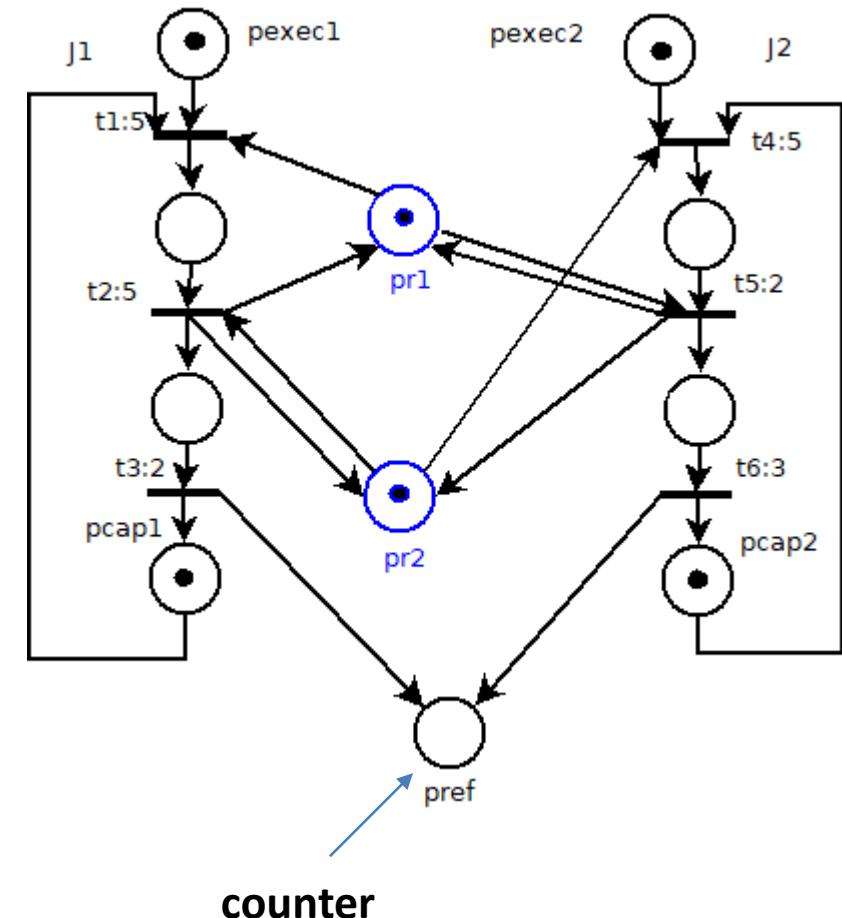
Capacité des jobs : 1

3 opérations o1, o2, o3 dans le job 1 de durée 5, 5 et 2

3 opérations o4, o5, o6 dans le job 2 de durée 5, 2 et 3

2 ressources partagée : pr1 partagée par o1, o2 et o5 et pr2 partagée par o2, o4 et o5

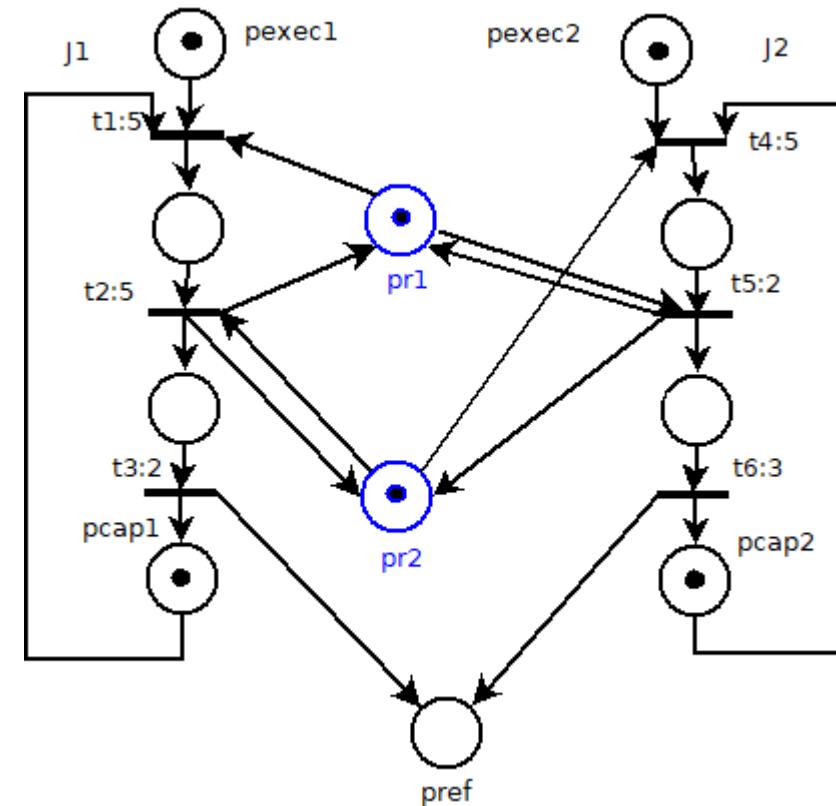
=> W_{PR} , W_{PO} , δ , M_i ???



MATLAB

```
>> [Wpr,Wpo,delta,MI,Pjob] = Exemple2_MACS(1,1)
```

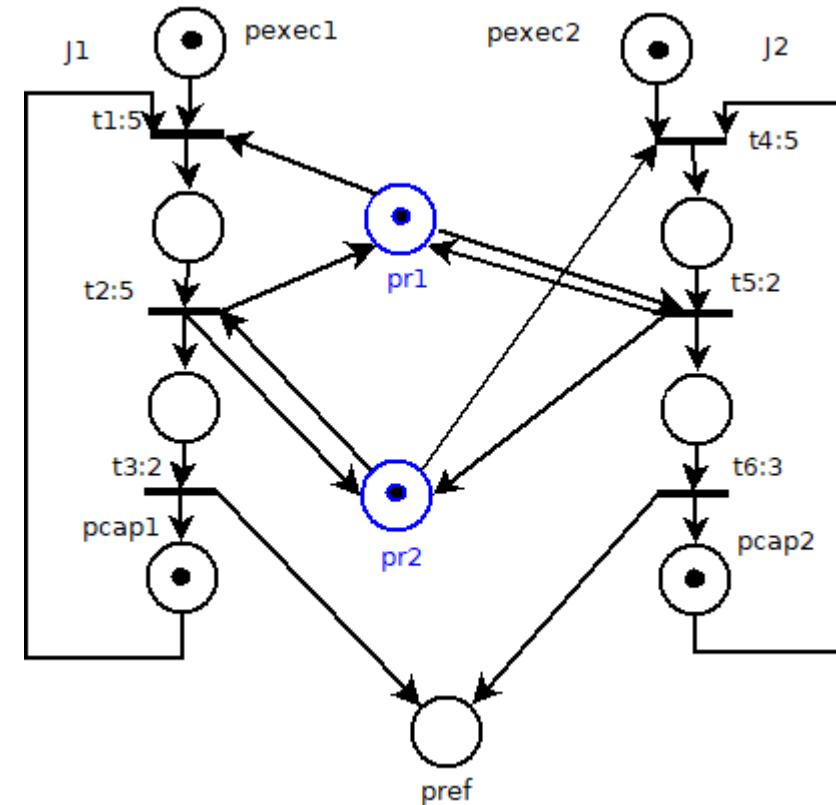
```
%1 2 3 4 5 6
Wpr=[ 1 0 0 0 0 0; %1
      1 0 0 0 0 0; %2
      0 1 0 0 0 0; %3
      0 0 1 0 0 0; %4
      0 0 0 1 0 0; %5
      0 0 0 1 0 0; %6
      0 0 0 0 1 0; %7
      0 0 0 0 0 1; %8
      1 0 0 0 1 0; %9
      0 1 0 1 0 0; %10
      0 0 0 0 0 0];%11
```



MATLAB

```
>> [Wpr,Wpo,delta,MI,Pjob] = Exemple2_MACS(1,1)
```

```
%1 2 3 4 5 6
Wpo=[ 0 0 0 0 0 0; %1
       0 0 1 0 0 0; %2
       1 0 0 0 0 0; %3
       0 1 0 0 0 0; %4
       0 0 0 0 0 0; %5
       0 0 0 0 0 1; %6
       0 0 0 1 0 0; %7
       0 0 0 0 1 0; %8
       0 1 0 0 1 0; %9
       0 1 0 0 1 0; %10
       0 0 1 0 0 1];%11
```

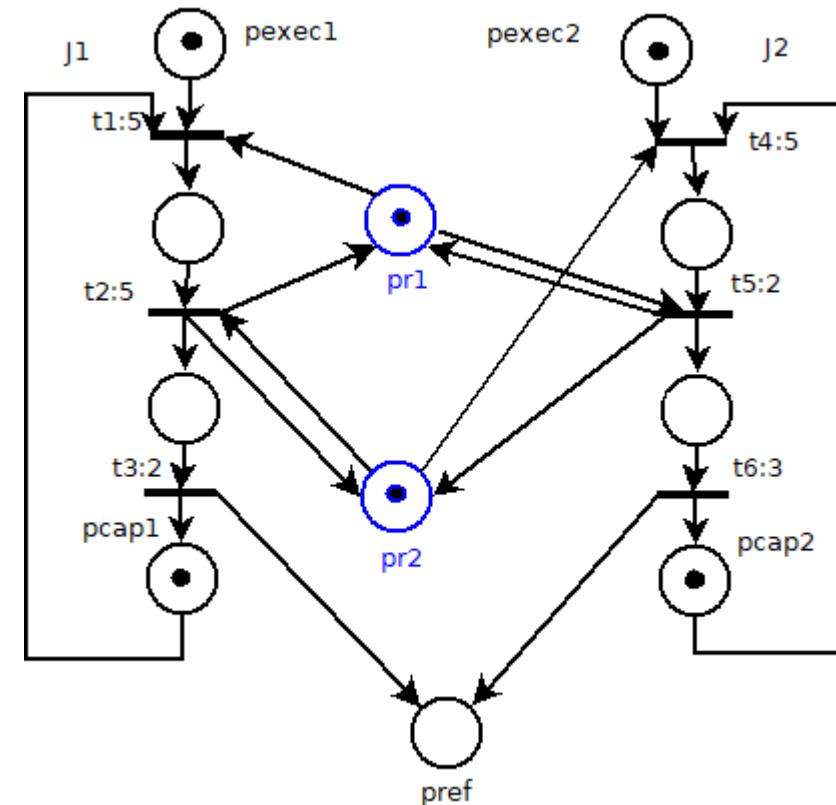


MATLAB

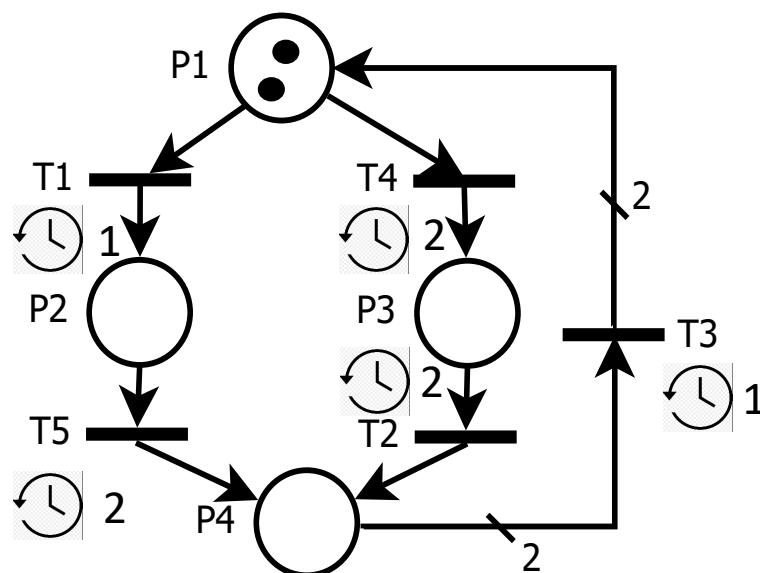
```
>> [Wpr,Wpo,delta,MI,Pjob] = Exemple2_MACS(1,1)
```

```
Pjob{1}.exec=1;
Pjob{1}.cap=2;
Pjob{1}.valcap=1;
Pjob{1}.job=[3 4];
Pjob{1}.ref=11;
Pjob{1}.Tc=[1:3];
Pjob{2}.exec=5;
Pjob{2}.cap=6;
Pjob{2}.valcap=1;
Pjob{2}.job=[7 8];
Pjob{2}.ref=11;
Pjob{2}.Tc=[4:6];
```

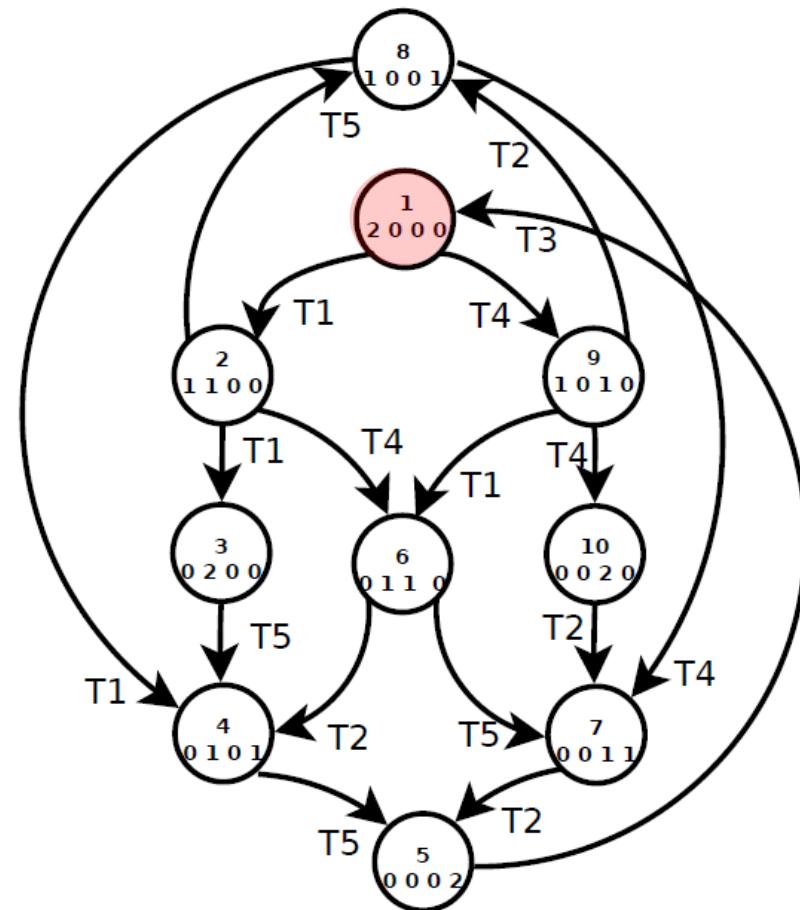
```
delta= [5 5 2 5 2 3]' ;
MI = [1 1 0 0 1 1 0 0 1 1 0]';
```



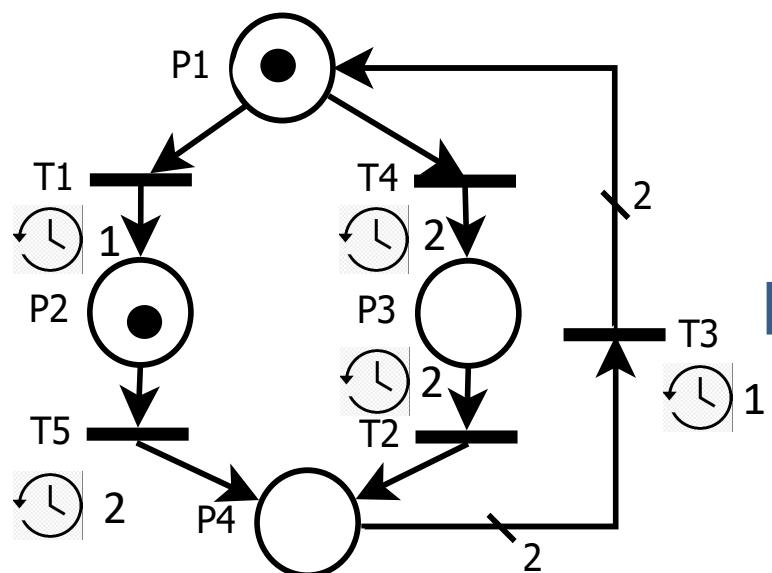
T-TPN



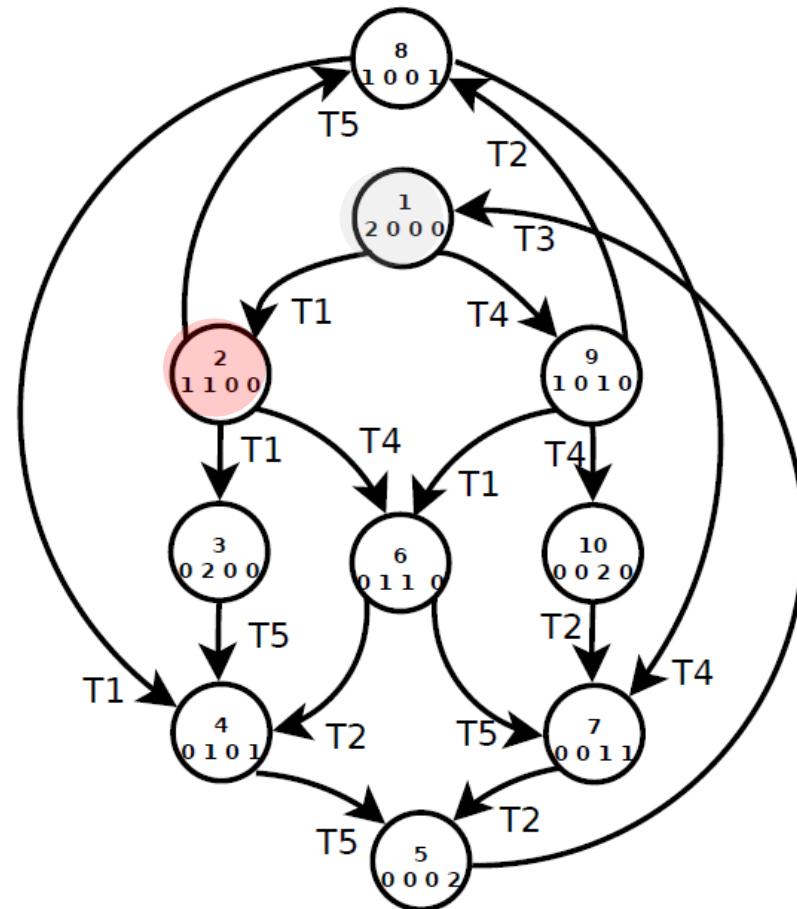
(untimed) reachability graph (RG)



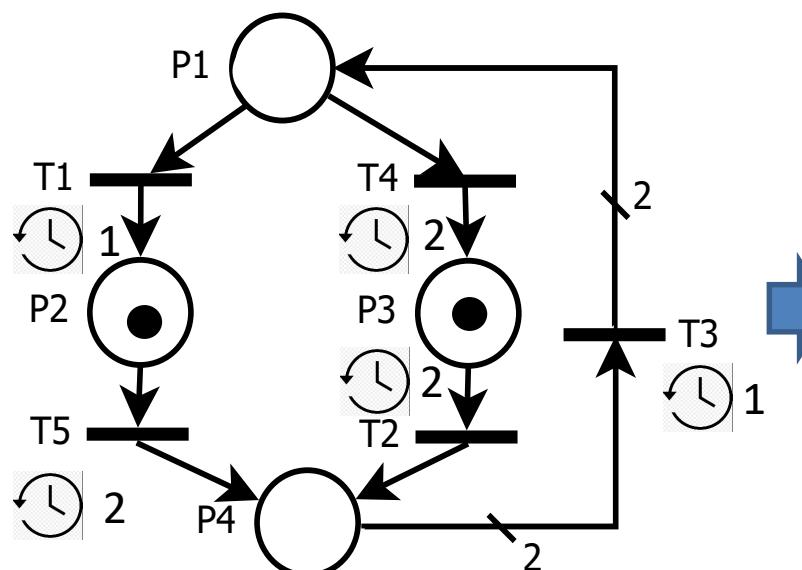
T-TPN



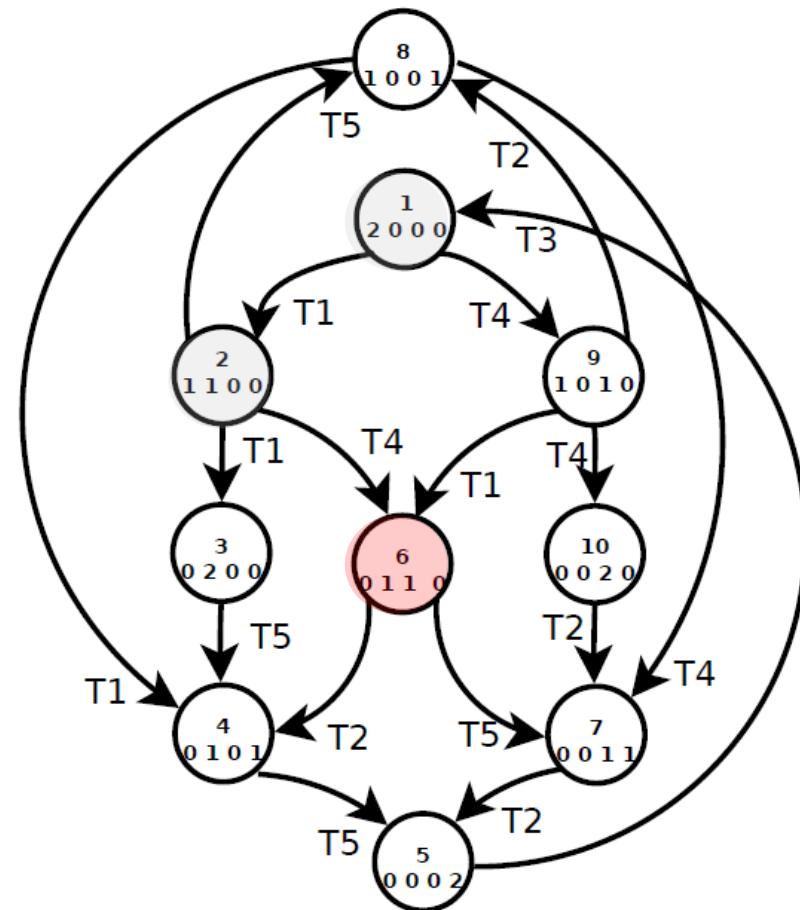
(untimed) reachability graph (RG)



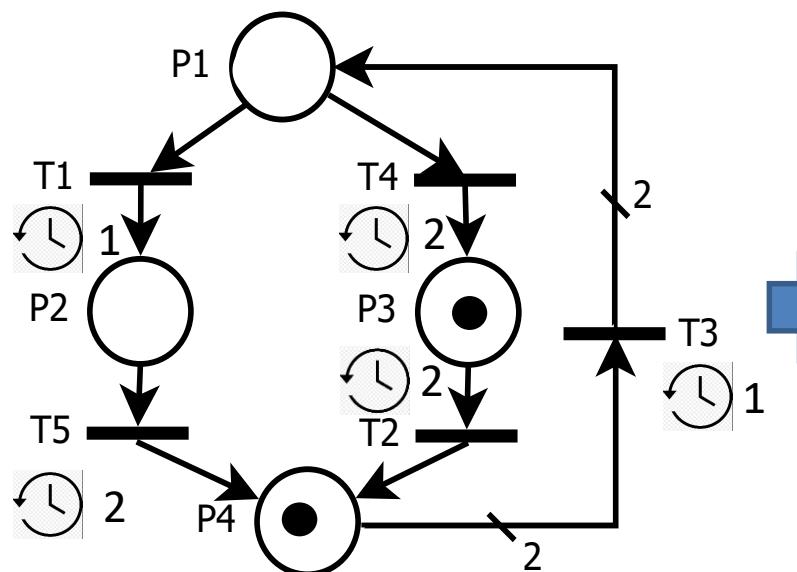
T-TPN



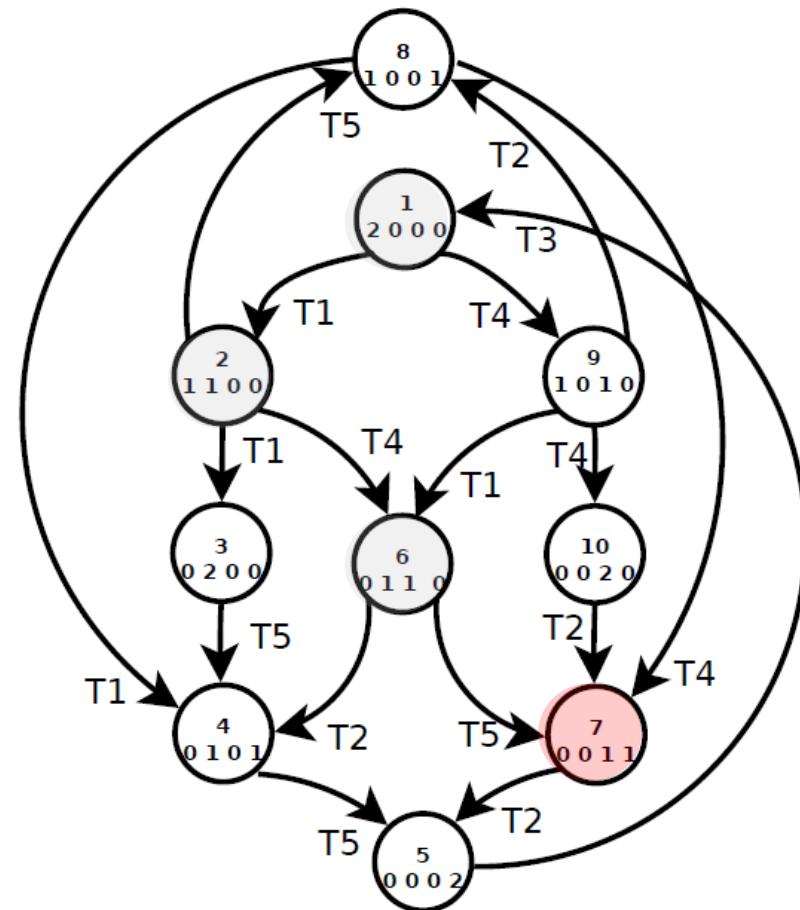
(untimed) reachability graph (RG)



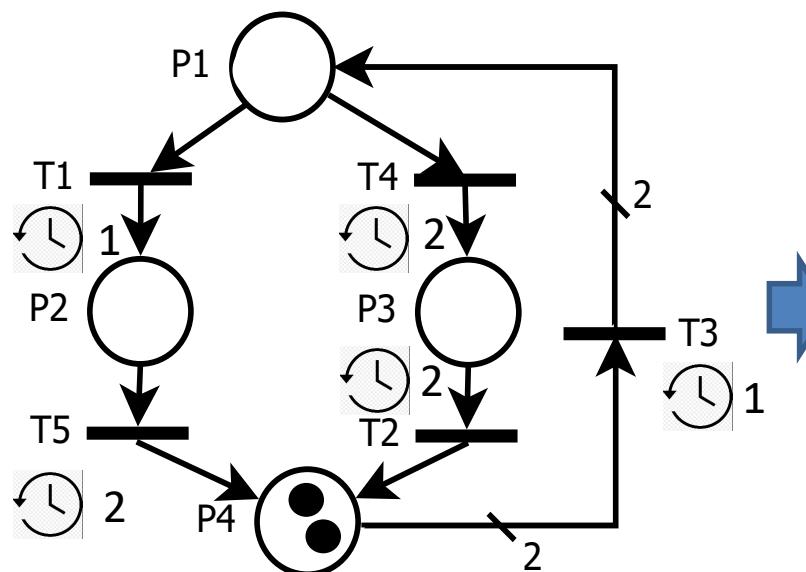
T-TPN



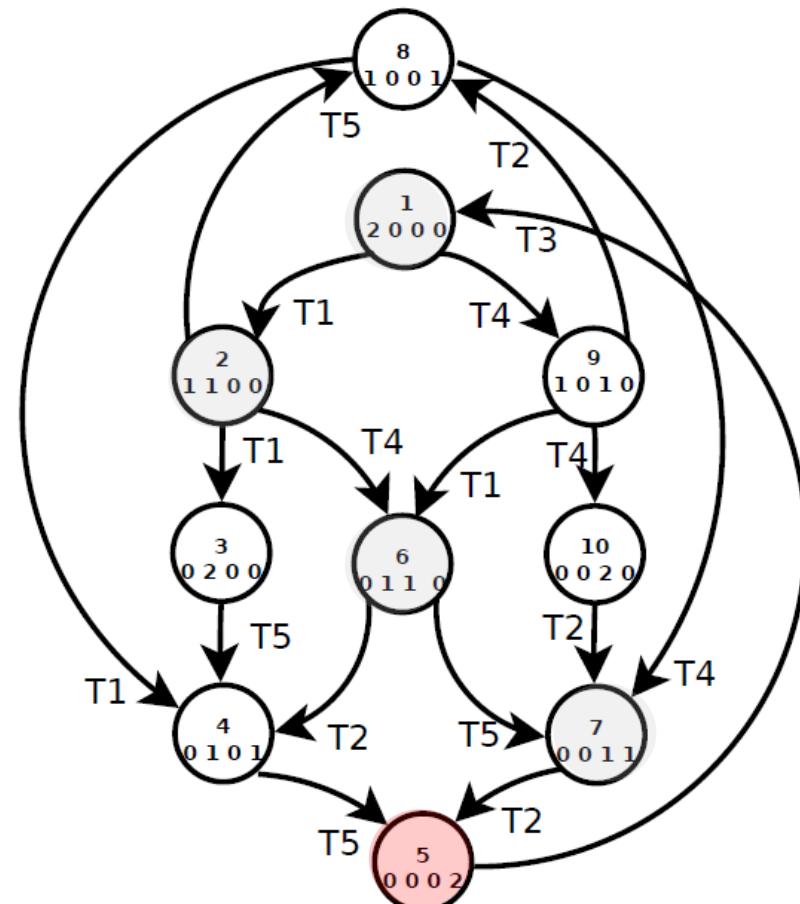
(untimed) reachability graph (RG)

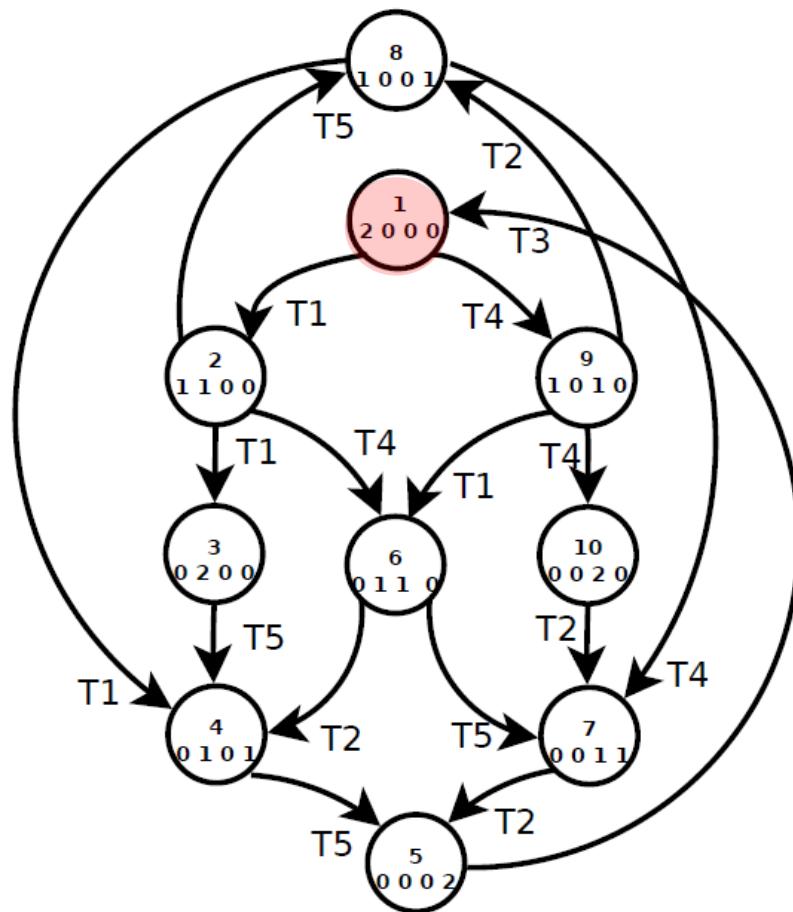


T-TPN

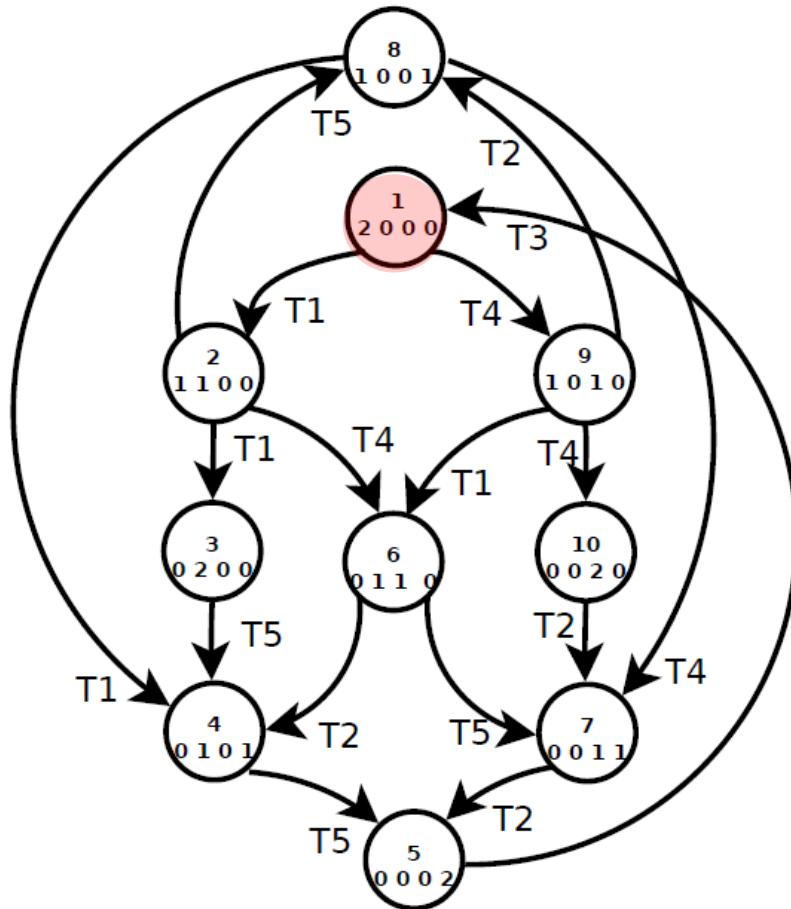


(untimed) reachability graph (RG)





Compute the reachability graph with MATLAB
`>>[G,S]=RG_MACS(Wpr,Wpo,delta,MI)`



RG =

0	1	4	Inf							
Inf	0	Inf	1	4	5	Inf	Inf	Inf	Inf	Inf
Inf	Inf	0	Inf	1	2	4	Inf	Inf	Inf	Inf
Inf	Inf	Inf	0	Inf	Inf	Inf	5	Inf	Inf	Inf
Inf	Inf	Inf	Inf	0	Inf	Inf	2	5	Inf	Inf
Inf	Inf	Inf	Inf	Inf	0	Inf	1	4	Inf	Inf
Inf	Inf	Inf	Inf	Inf	Inf	0	Inf	2	Inf	Inf
Inf	0	Inf	5	Inf						
Inf	0	Inf	2							
3	Inf	0								

S=1	2	3	4	5	6	7	8	9	10
2	1	1	0	0	1	0	0	0	0
0	1	0	2	1	0	0	1	0	0
0	0	1	0	1	0	2	0	1	0
0	0	0	0	0	1	0	1	1	2

Compute the reachability graph with MATLAB

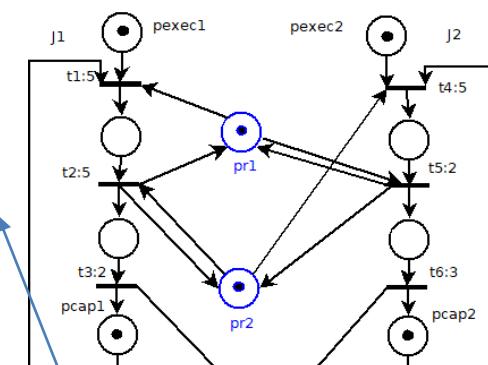
>>[RG,S]=RG_MACS(Wpr,Wpo,delta,MI)

MATLAB with example 2

>> [Wpr,Wpo,delta,MI,Pjob] = Exemple2 MACS(1,1)

>>[RG,S]=RG_MACS(Wpr,Wpo,delta,MI)

RG: 16 states



Deadlocks

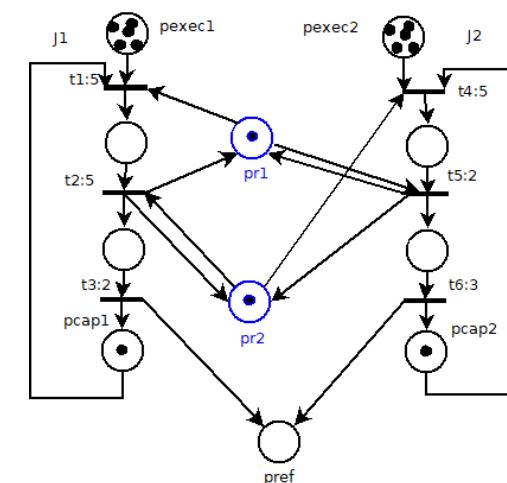
MATLAB with example 2

```
>> [Wpr,Wpo,delta,MI,Pjob] = Exemple2_MACS(5,5)
```

```
>>[RG,S]=RG_MACS(Wpr,Wpo,delta,MI)
```

RG: 256 states

Find the deadlocks in the problem ?



MATLAB with example 2

```
>> [Wpr,Wpo,delta,MI,Pjob] = Exemple2_MACS(5,5)
```

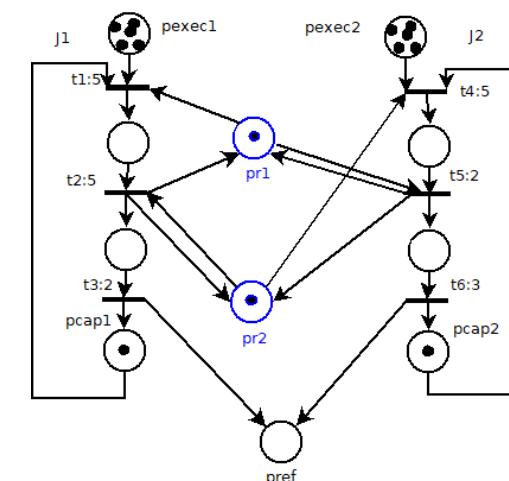
```
>>[RG,S]=RG_MACS(Wpr,Wpo,delta,MI)
```

RG: 256 states

Find the deadlocks in the problem ?

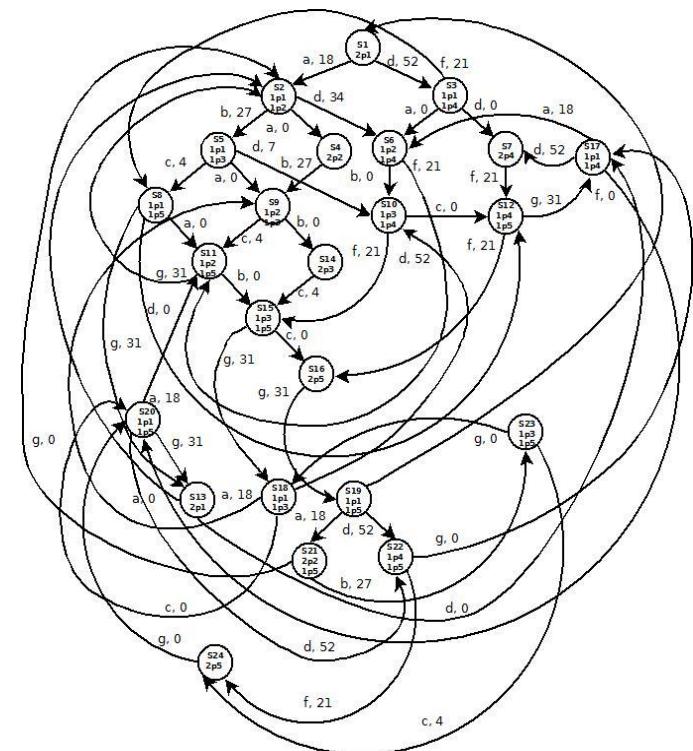
26 deadlocks:

```
[ 5  17  20  38  41  44  68  71  74  77
 107 110 113 116 119 154 157 160
 163 193 196 199 223 226 244 256]
```



Outline

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 - 3. Timed Extended Reachability Graph**
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 6. Conclusion and future works



How to compute an optimal schedule from the RG

Search for paths of minimal durations using

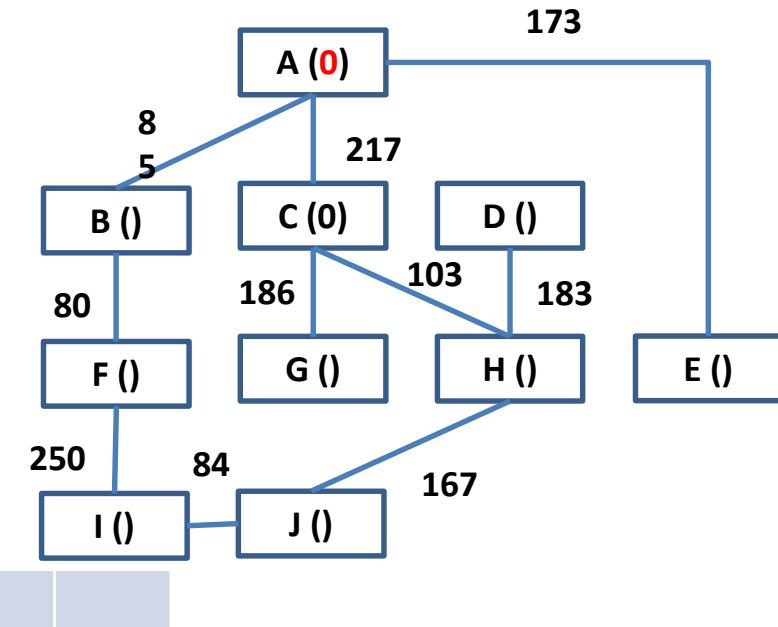
- 1) The generator matrix of the RG**
- 2) The time specifications δ**
- 3) Replace each transition T_j in RG by $\delta(T_j)$**
- 4) Dijkstra algorithm that searches for shortest path**

Cormen TH, Leiserson CE, Rivest RL, Stein C (2001) *Introduction to algorithms*.
MIT Press and McGraw-Hill

Dijkstra EW (1971) A short introduction to the art of programming

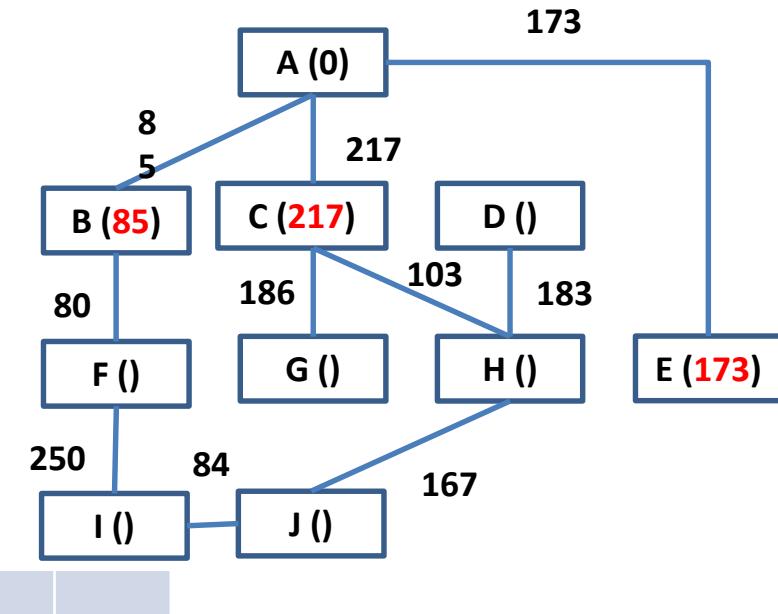
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2									
3									
4									
5									
6									
7									
8									
9									
10									
11									
pre	A								



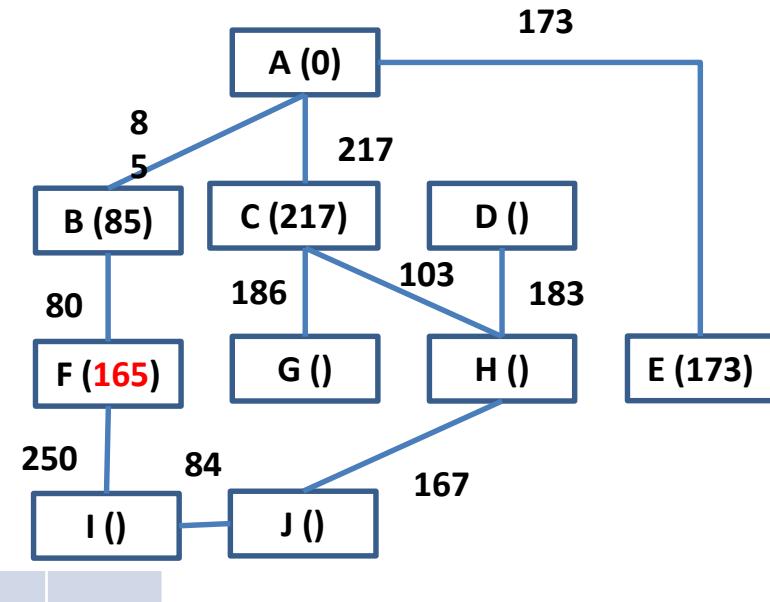
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3									
4									
5									
6									
7									
8									
9									
10									
11									
pre	A	A							



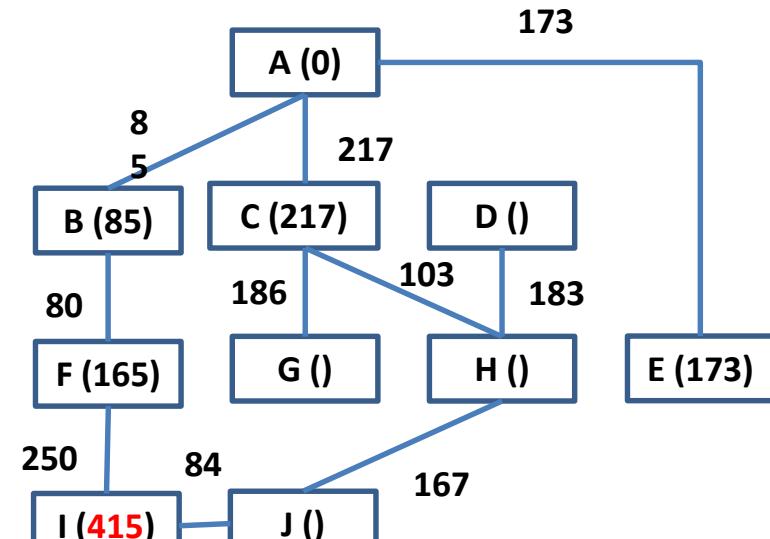
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3	X	X	217		173	165			
4									
5									
6									
7									
8									
9									
10									
11									
pre	A	A				B			



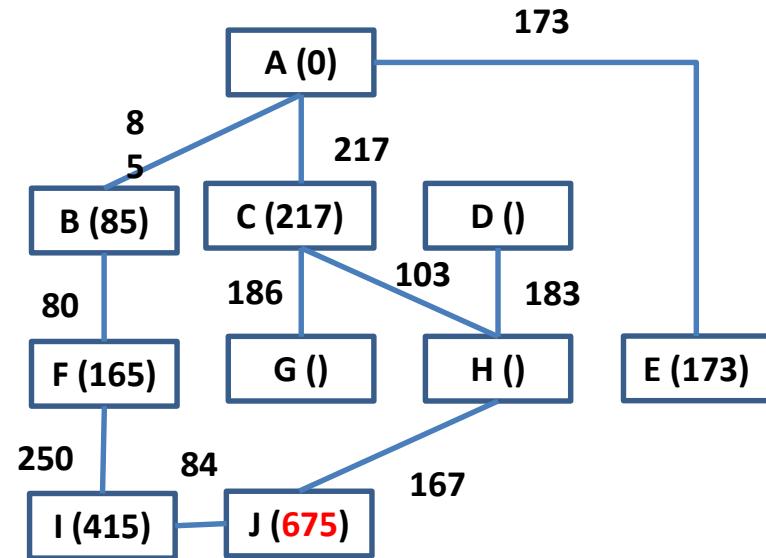
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3	X	X	217		173	165			
4	X	X	217		173	X			4
5									
6									
7									
8									
9									
10									
11									
pre	A	A			A	B			



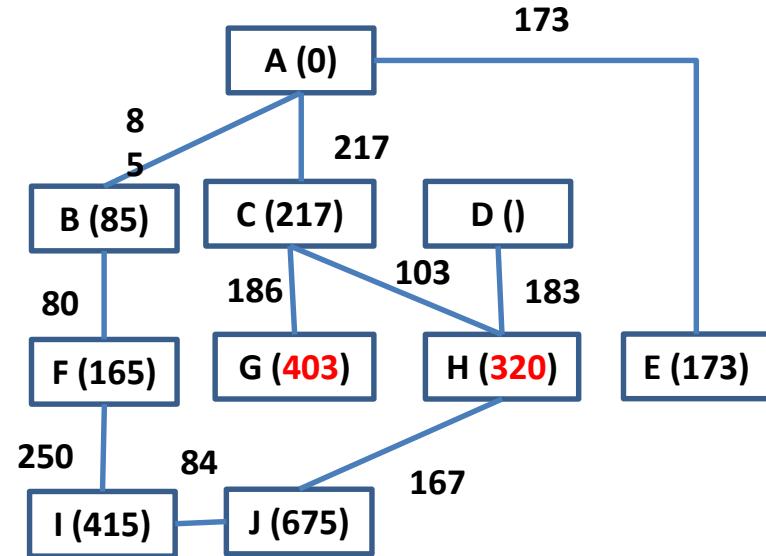
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3	X	X	217		173	165			
4	X	X	217		173	X			415
5	X	X	217		X	X			415 675
6									
7									
8									
9									
10									
11									
pre	A	A	A		A	B			



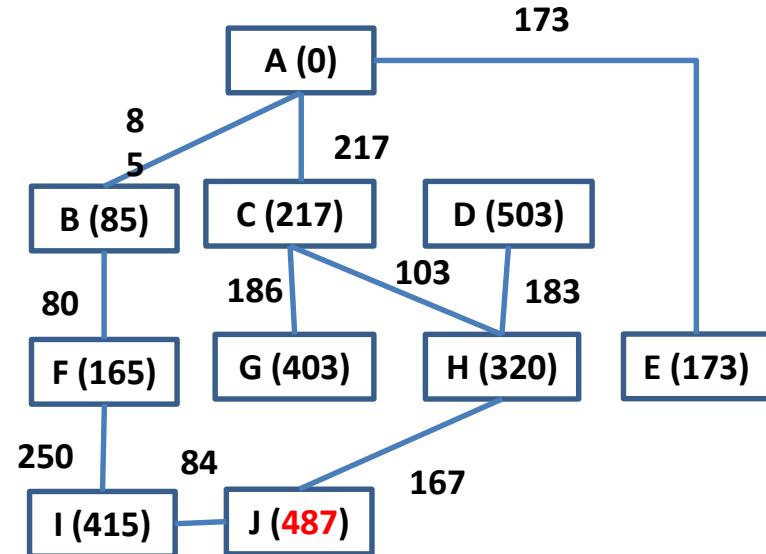
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3	X	X	217		173	165			
4	X	X	217		173	X			415
5	X	X	217		X	X			415 675
6	X	X	X		X	X	403	320	415 675
7									
8									
9									
10									
11									
pre	A	A	A		A	B		C	



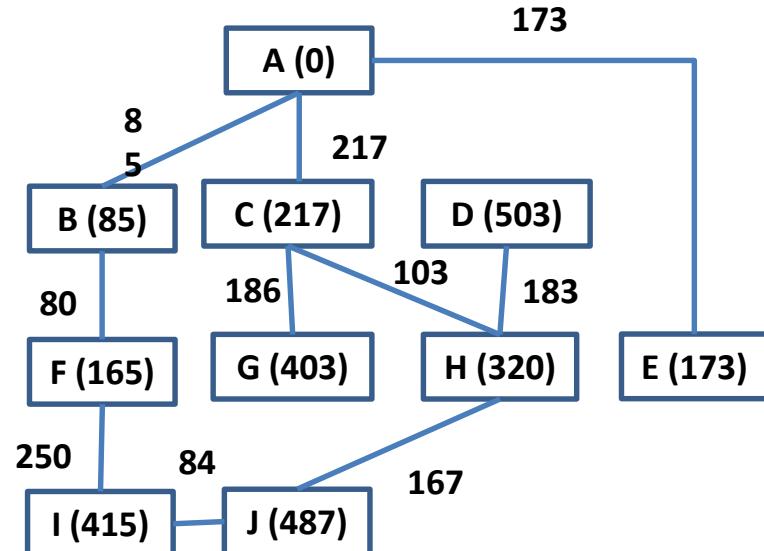
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3	X	X	217		173	165			
4	X	X	217		173	X			4
5	X	X	217		X	X			415 675
6	X	X	X		X	X	403	320	415 675
7	X	X	X	503	X	X	403	X	415 487
8									
9									
10									
11									
pre	A	A	A		A	B	C	C	



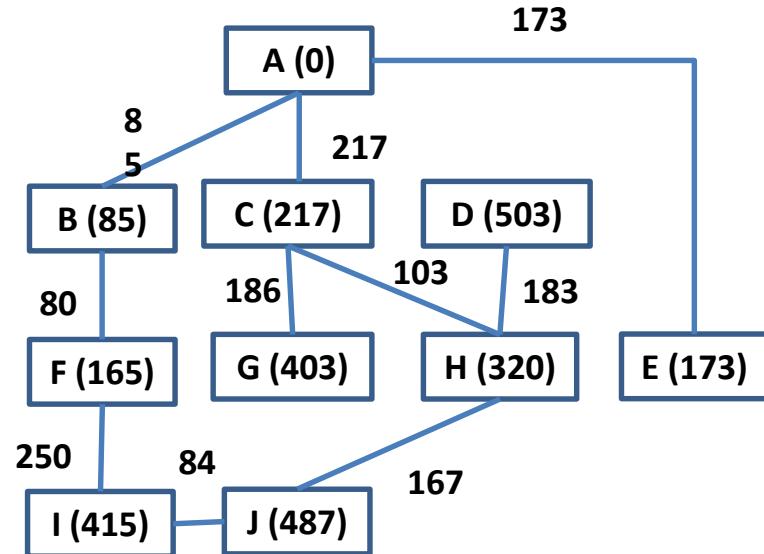
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3	X	X	217		173	165			
4	X	X	217		173	X			415
5	X	X	217		X	X			415 675
6	X	X	X		X	X	403	320	415 675
7	X	X	X	503	X	X	403	X	415 487
8	X	X	X	503	X	X	X	X	415 487
9									
10									
11									
pre	A	A	A		A	B	C	C	F



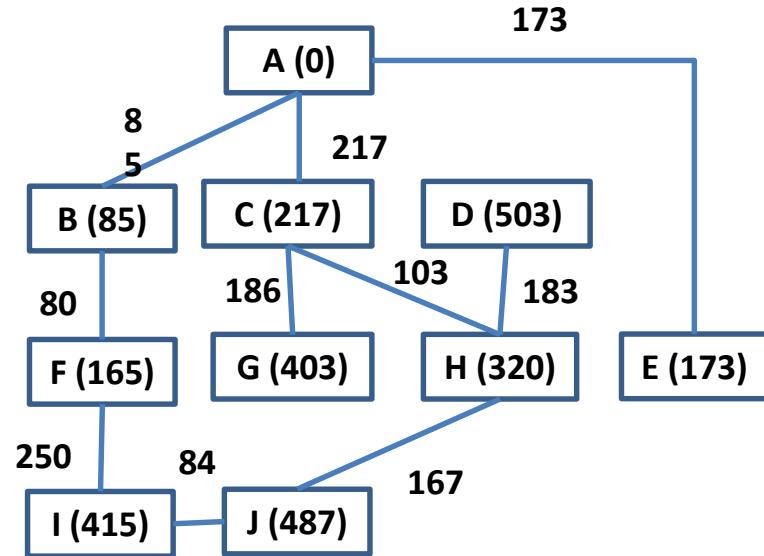
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3	X	X	217		173	165			
4	X	X	217		173	X			415
5	X	X	217		X	X			415 675
6	X	X	X		X	X	403	320	415 675
7	X	X	X	503	X	X	403	X	415 487
8	X	X	X	503	X	X	X	X	415 487
9	X	X	X	503	X	X	X	X	487
10									
11									
pre	A	A	A		A	B	C	C	F H



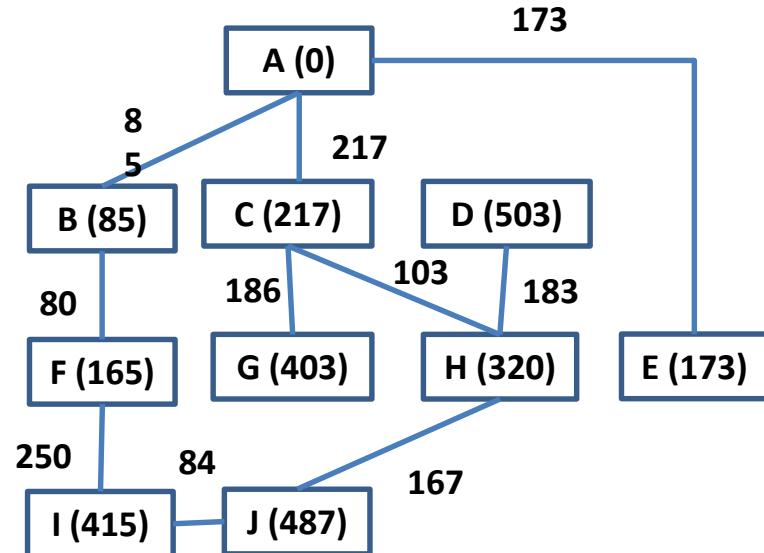
Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3	X	X	217		173	165			
4	X	X	217		173	X			415
5	X	X	217		X	X			415 675
6	X	X	X		X	X	403	320	415 675
7	X	X	X	503	X	X	403	X	415 487
8	X	X	X	503	X	X	X	X	415 487
9	X	X	X	503	X	X	X	X	487
10	X	X	X	503	X	X	X	X	X
11									
pre	A	A	A	H	A	B	C	C	F
									H



Dijkstra algorithm

	A	B	C	D	E	F	G	H	I
1	0								
2	X	85	217		173				
3	X	X	217		173	165			
4	X	X	217		173	X			415
5	X	X	217		X	X			415 675
6	X	X	X		X	X	403	320	415 675
7	X	X	X	503	X	X	403	X	415 487
8	X	X	X	503	X	X	X	X	415 487
9	X	X	X	503	X	X	X	X	487
10	X	X	X	503	X	X	X	X	X
11	X	X	X	X	X	X	X	X	X
pre	A	A	A	H	A	B	C	C	F
									H



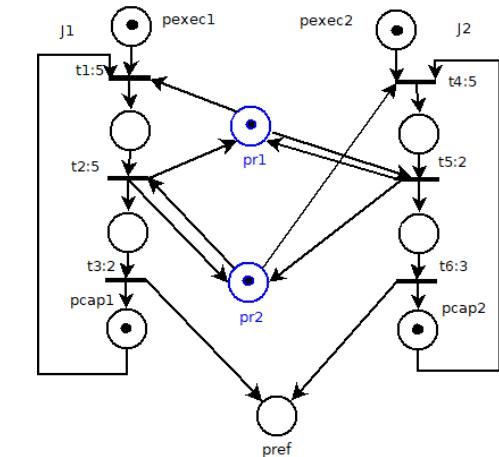
MATLAB with example 2

```
>> [Wpr,Wpo,delta,MI,Pjob] = Exemple2_MACS(1,1)
>>[RG,S,RG_time]=RG_MACS(Wpr,Wpo,delta,MI)
```

Transition labels

```
RG =  0  1  4  Inf  Inf
      Inf  0  Inf  2  4  Inf  Inf
      Inf  Inf  0  Inf  1  5  Inf  Inf
      Inf  Inf  Inf  0  Inf  Inf  3  4  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf
      Inf  Inf  Inf  Inf  0  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  0  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  1  6  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  Inf  4  Inf  Inf  Inf  Inf  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  3  5  Inf  Inf  Inf  Inf  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  2  6  Inf  Inf  Inf  Inf  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  1  Inf  Inf  Inf  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  5  Inf  Inf  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  3  6  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  0  Inf  2  Inf  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  0  Inf  6  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  0  3  Inf
      Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  0  Inf
```

```
delta= [5  5  2  5  2  3]' ;
```



Transition firing duration

```
0  5  5  Inf  Inf
Inf  0  Inf  5  5  Inf  Inf
Inf  Inf  0  Inf  5  2  Inf  Inf
Inf  Inf  Inf  0  Inf  Inf  2  5  Inf  Inf
Inf  Inf  Inf  Inf  0  Inf  Inf
Inf  Inf  Inf  Inf  Inf  0  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  5  3  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  5  Inf  Inf  Inf  Inf  Inf  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  2  2  Inf  Inf  Inf  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  5  3  Inf  Inf  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  5  Inf  Inf  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  Inf  2  Inf  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  2  3  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  0  Inf  5  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0  Inf  Inf  0  Inf  3
Inf  0  2
Inf  0  0
```

MATLAB

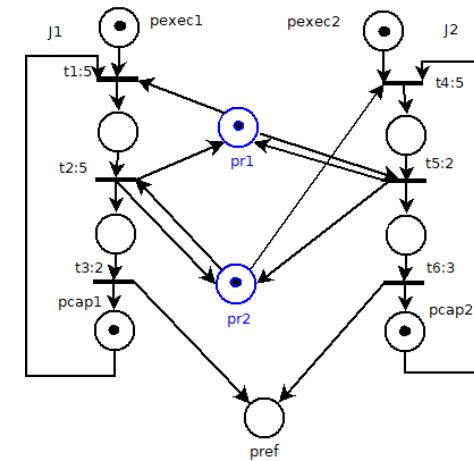
```
>>[precedesseur] = Dijkstra_predecessor_MACS([1:16],RG_time,1)
>>MS=16
>>[seq,cost] = Dijkstra_seq_MACS(TERG,TERG_time,precedesseur,1,MS)
```

seq =[1 2 3 4 5 6]

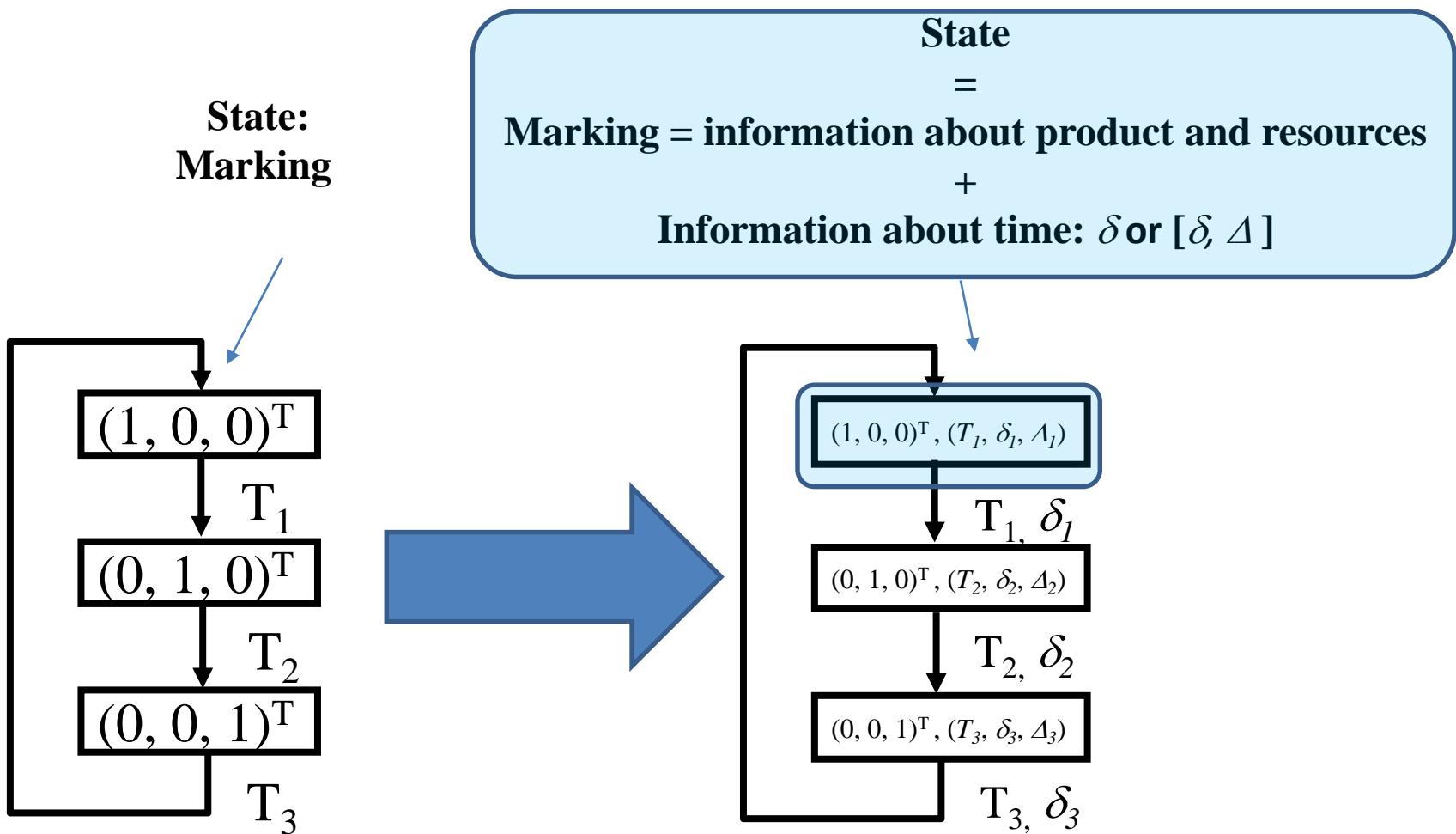
duration = 22 !!!

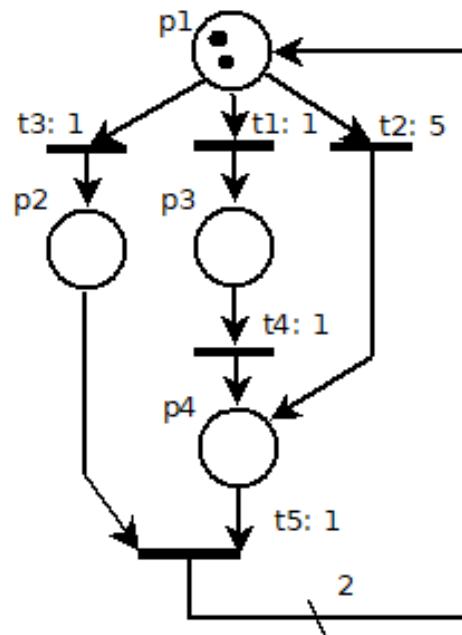


delta= [5 5 2 5 2 3]' ;

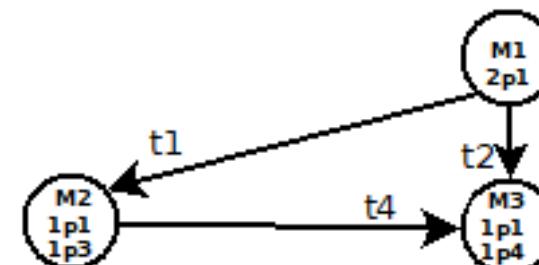


Timed Extended RG versus Untimed RG

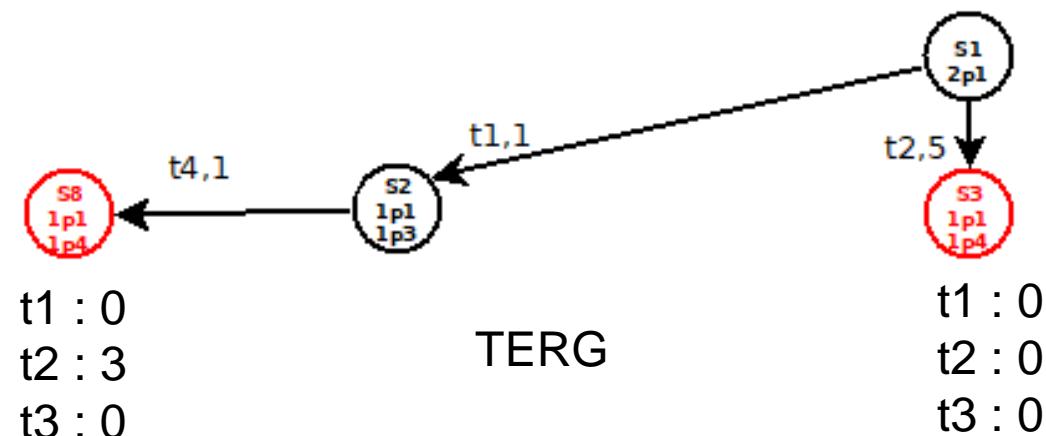




$$\begin{aligned}
 M_1 &= (2 \ 0 \ 0 \ 0)^T = 2p_1 \\
 M_{ref} &= (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4 \\
 \delta &= (1 \ 5 \ 1 \ 1 \ 1)^T \\
 \Delta &= (\infty \ \infty \ \infty \ \infty \ \infty)^T
 \end{aligned}$$



Untimed RG

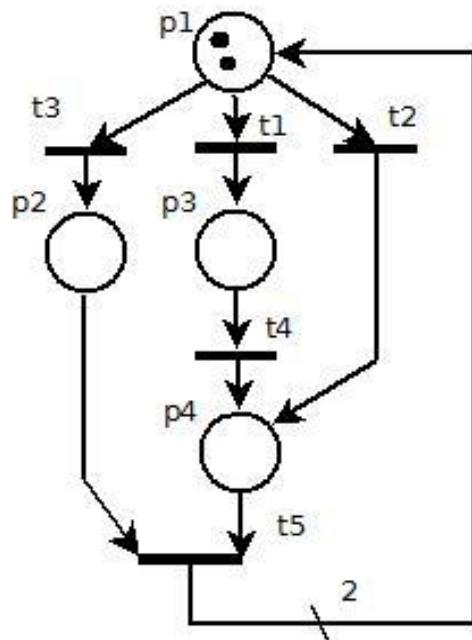


Timed Extended RG (TERG)

Proposition: Let $\langle G, STI, M_I, A \rangle$ be a controlled TPN system that behaves under earliest firing policy and its corresponding TERG $\langle S_E(M_I), \Omega_E, B_E, S_0 \rangle$.

- (a) If the timed trajectory $(\sigma, M_I) = M(0) [(t_{j_1}, \tau_1) > M(1) \dots > M(h-1) [(t_{j_h}, \tau_h) > M(h)]$ with $M(0) = M_I$ and $t_{j_k} \in T$, $k = 1, \dots, h$, is feasible in TPN system, then a path $S_0 S_1 \dots S_h$ exists in $\langle S_E(M_I), \Omega_E, B_E, S_0 \rangle$ st (1) $M(S_k) = M(k)$ for $k = 0, \dots, h$; (2) $\Omega_E(S_{k-1}, S_k) = t_{j_k}$ and $B_E(S_{k-1}, S_k).dt = \tau_k - \tau_{k-1}$ for $k = 1, \dots, h$.
- (b) If $S_0 S_1 \dots S_h$ is a path in TERG $\langle S_E(M_I), \Omega_E, B_E, S_0 \rangle$ with S_0 the root node of the TERG then the timed trajectory $(\sigma, M_I) = M(0) [(t_{j_1}, \tau_1) > M(1) \dots > M(h-1) [(t_{j_h}, \tau_h) > M(h)]$ with (1) $M(k) = M(S_k)$, $k = 0, \dots, h$; (2) $t_{j_k} = \Omega_E(S_{k-1}, S_k)$, $k = 1, \dots, h$; (3) $\tau_1 = B_E(S_0, S_1).dt$; (4) $\tau_k = \tau_{k-1} + B_E(S_{k-1}, S_k).dt$, $k = 1, \dots, h$; is feasible in TPN

All feasible timed trajectories in time PN starting from M_I and executed under earliest firing policy are EXACTLY encoded in Timed Extended Reachability Graph

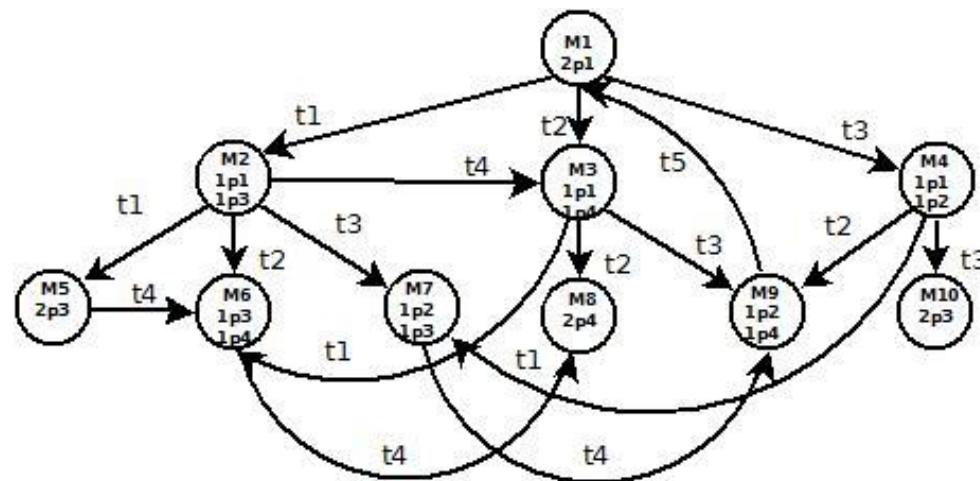


$$M_I = (2 \ 0 \ 0 \ 0)^T = 2p_1$$

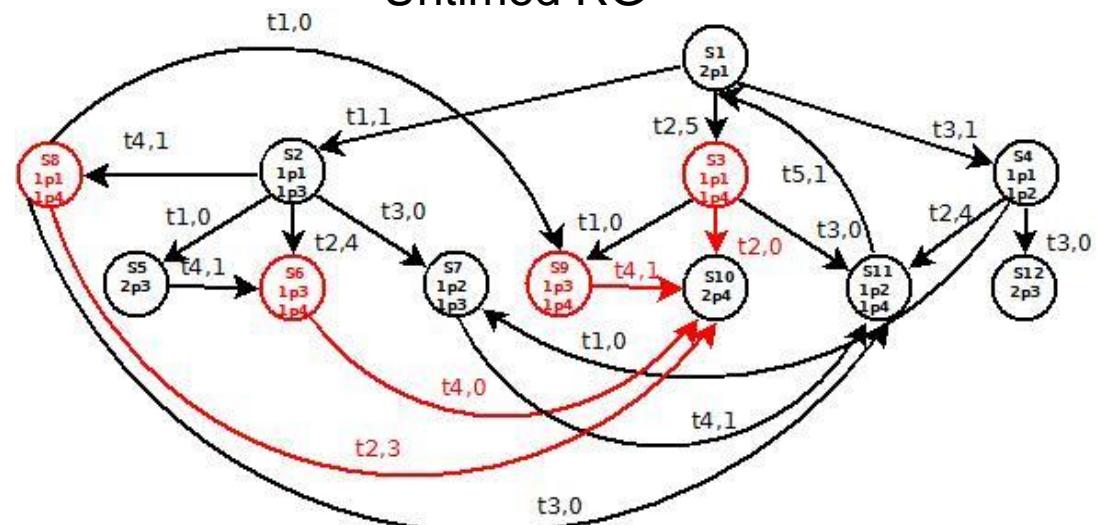
$$M_{ref} = (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4$$

$$\delta = (1 \ 5 \ 1 \ 1 \ 1)^T$$

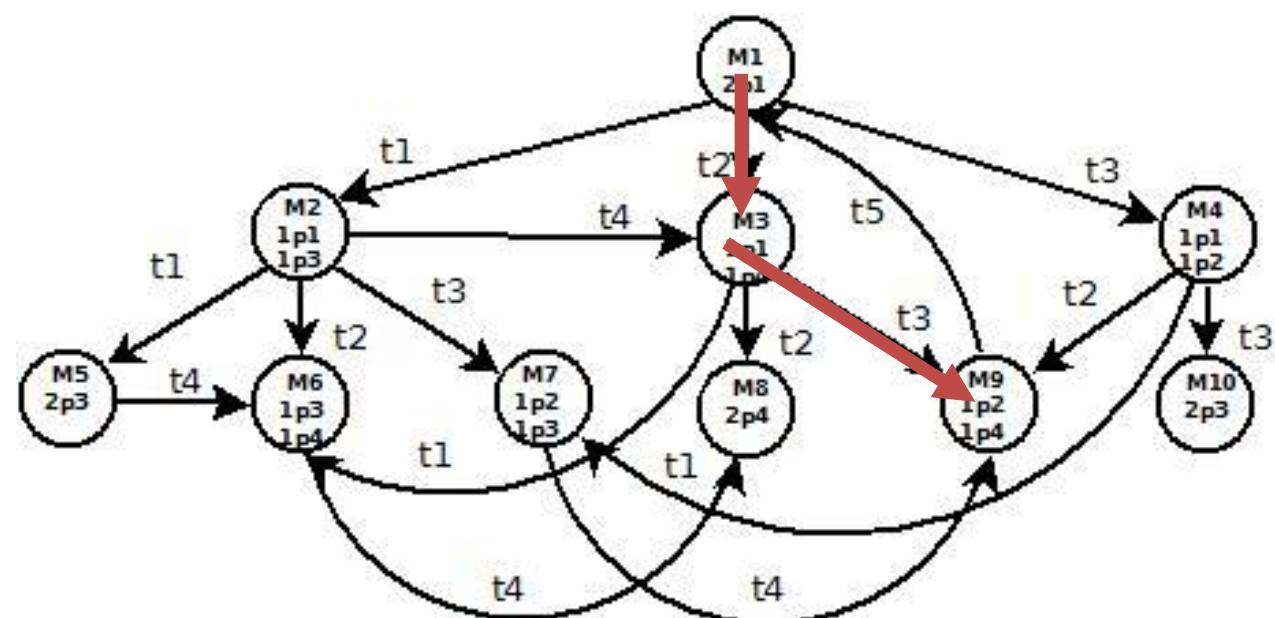
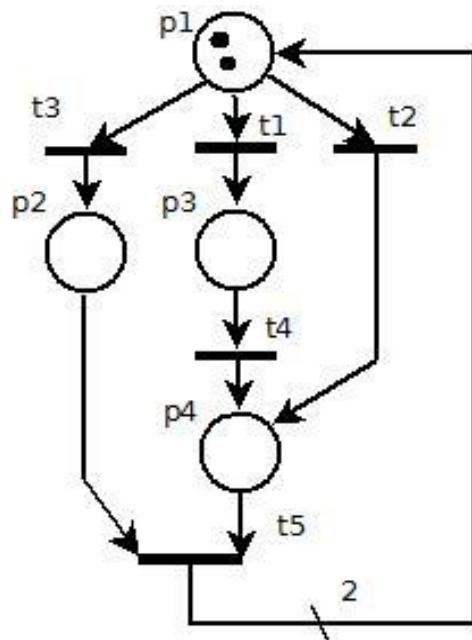
$$\Delta = (\infty \ \infty \ \infty \ \infty \ \infty)^T$$



Untimed RG



TERG



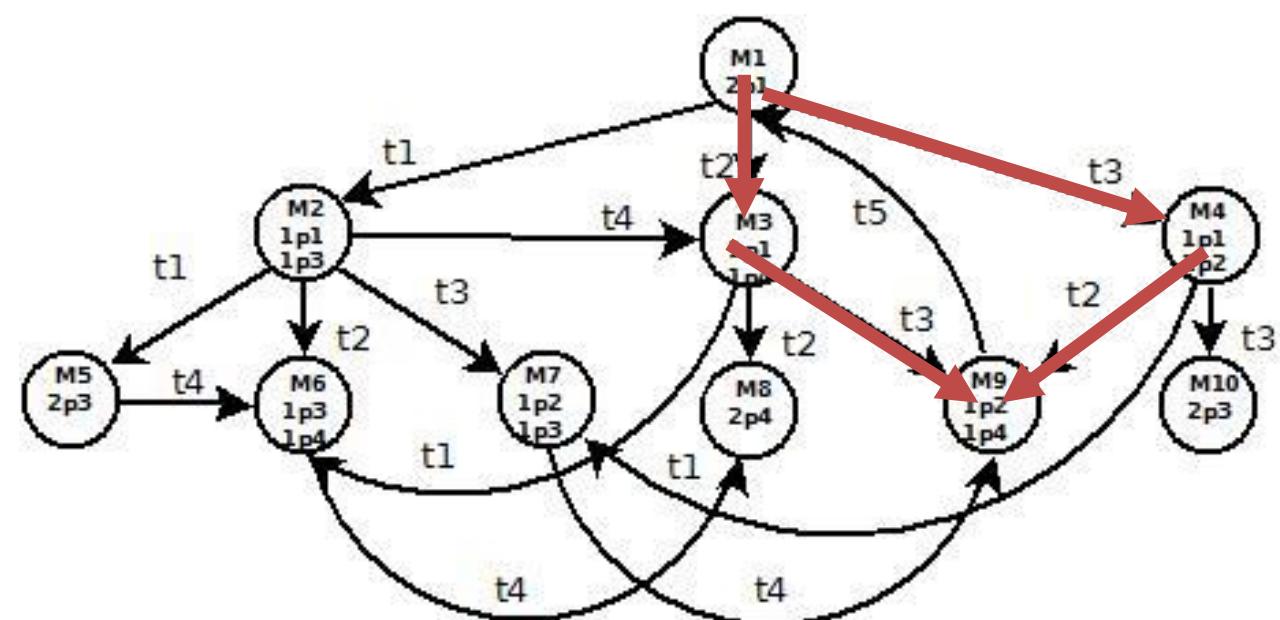
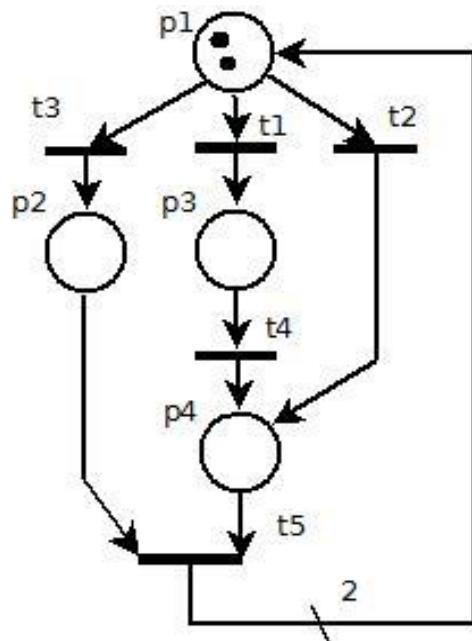
$$M_i = (2 \ 0 \ 0 \ 0)^T = 2p_1$$

$$M_{ref} = (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4$$

$$\delta = (1 \ 5 \ 1 \ 1 \ 1)^T$$

$$\Delta = (\infty \ \infty \ \infty \ \infty \ \infty)^T$$

Untimed RG



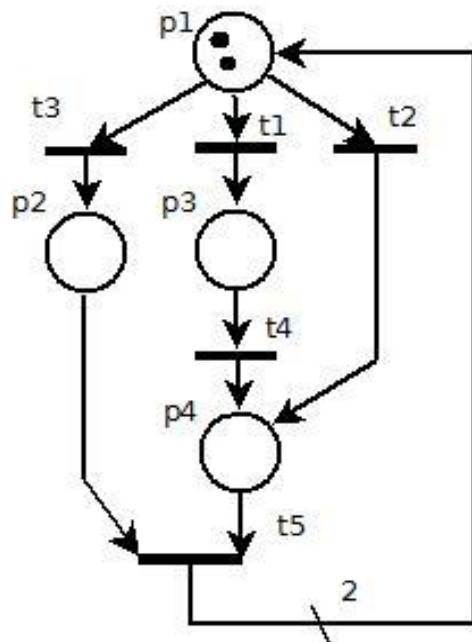
$$M_i = (2 \ 0 \ 0 \ 0)^T = 2p_1$$

$$M_{ref} = (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4$$

$$\delta = (1 \ 5 \ 1 \ 1 \ 1)^T$$

$$\Delta = (\infty \ \infty \ \infty \ \infty \ \infty)^T$$

Untimed RG

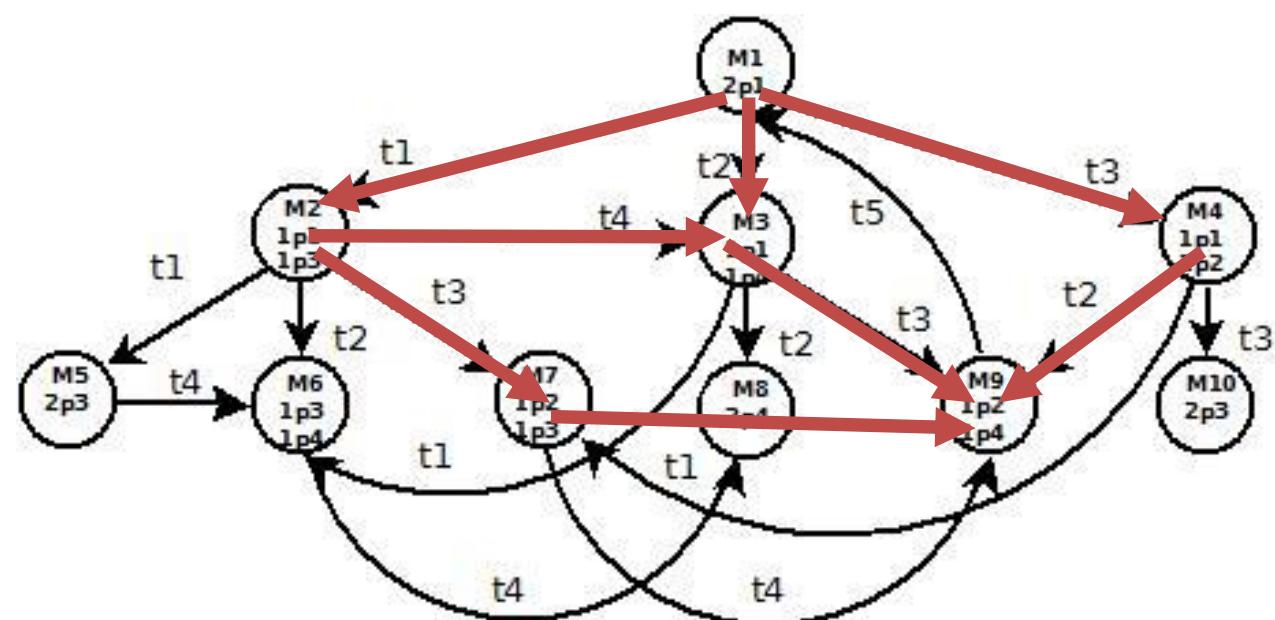


$$M_i = (2 \ 0 \ 0 \ 0)^T = 2p_1$$

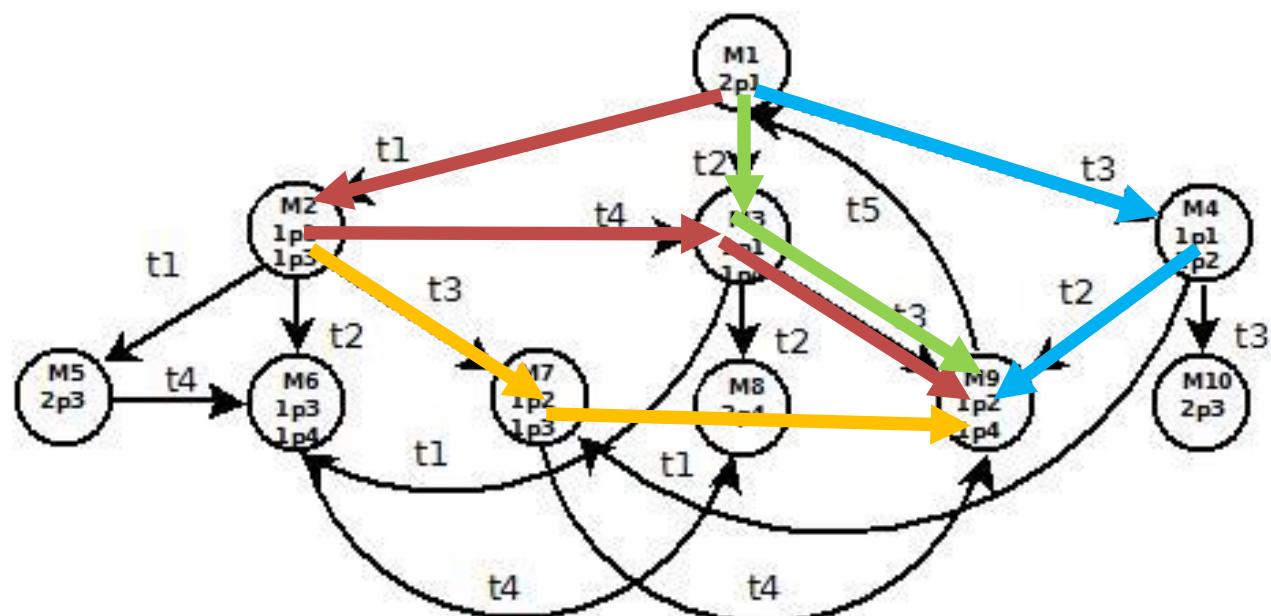
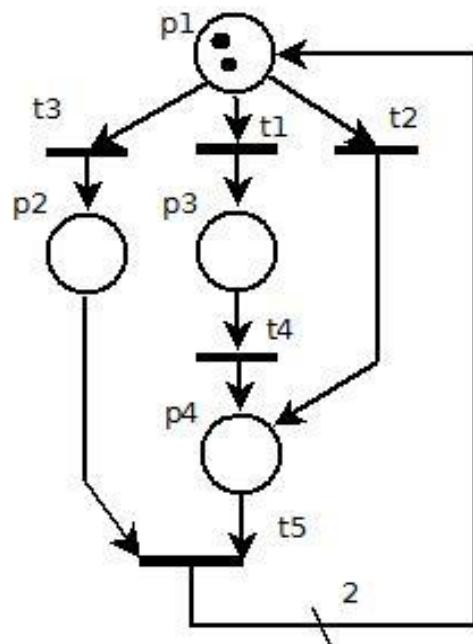
$$M_{ref} = (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4$$

$$\delta = (1 \ 5 \ 1 \ 1 \ 1)^T$$

$$\Delta = (\infty \ \infty \ \infty \ \infty \ \infty)^T$$



Untimed RG : what is the path of minimal duration ?
=> exhaustive path enumeration
+ path duration evaluation



Exercice

Path durations

Path 1 : ???

Path 2 : ???

Path 3 : ???

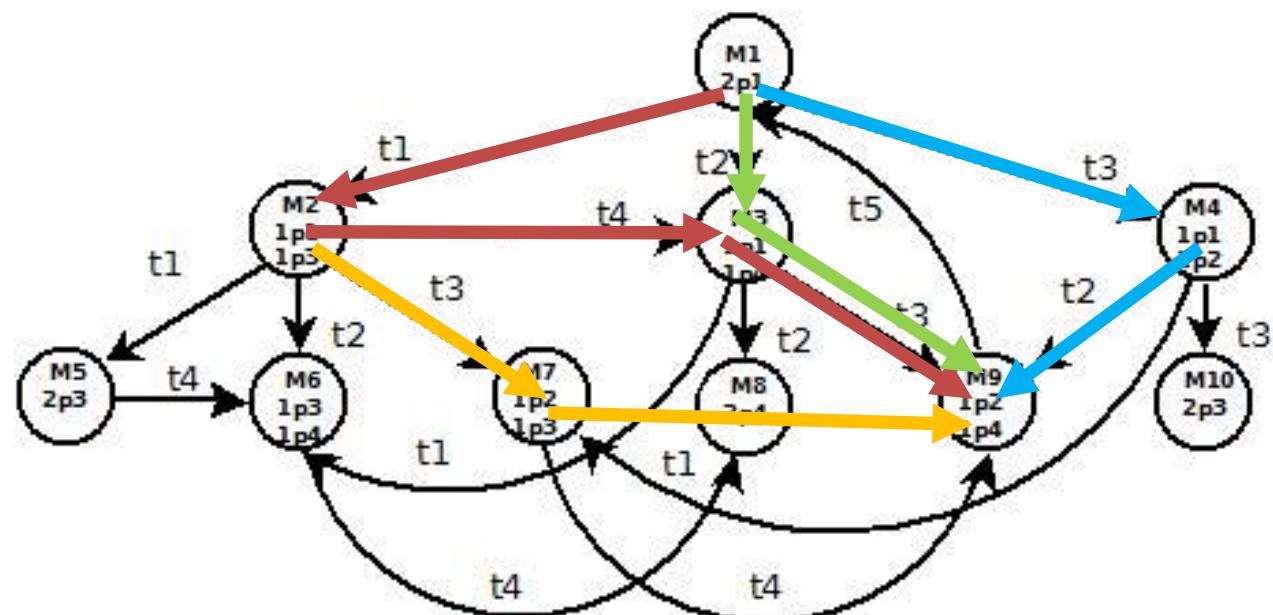
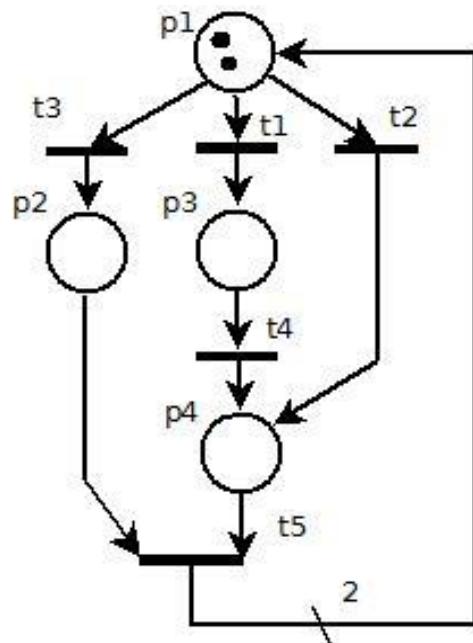
Path 4 : ???

$$M_1 = (2 \ 0 \ 0 \ 0)^T = 2p_1$$

$$M_{ref} = (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4$$

$$\delta = (1 \ 5 \ 1 \ 1 \ 1)^T$$

$$\Delta = (\infty \ \infty \ \infty \ \infty \ \infty)^T$$



Exercice

Path durations

Path 1 : T3(1)T2(5), duration = 5

Path 2 : ???

Path 3 : ???

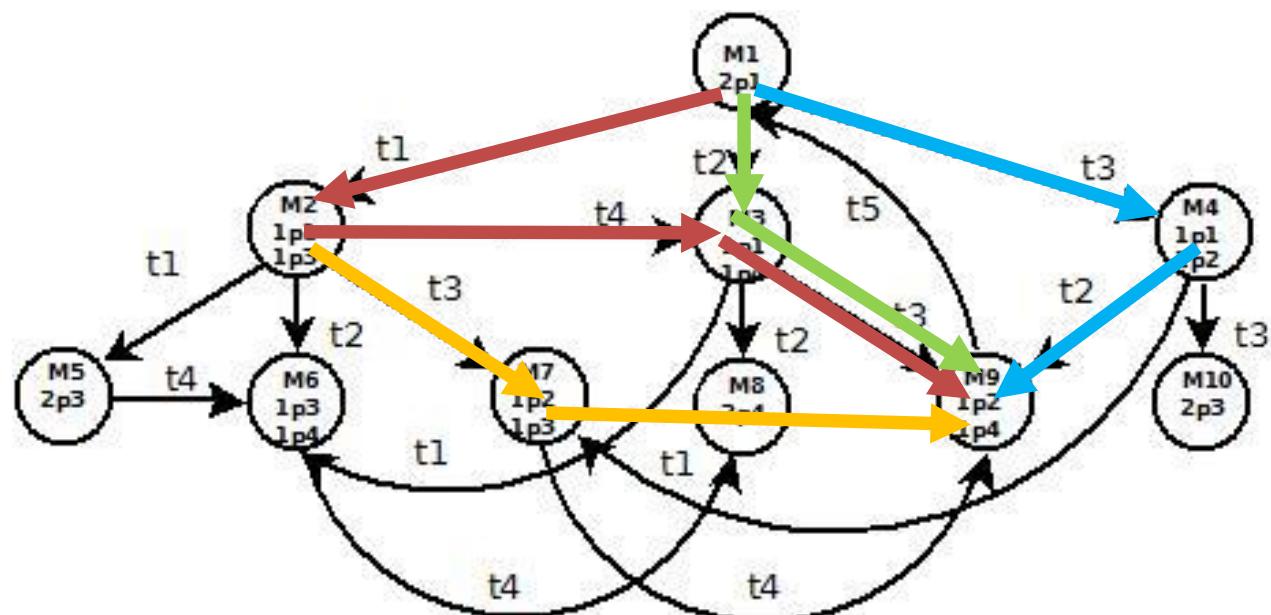
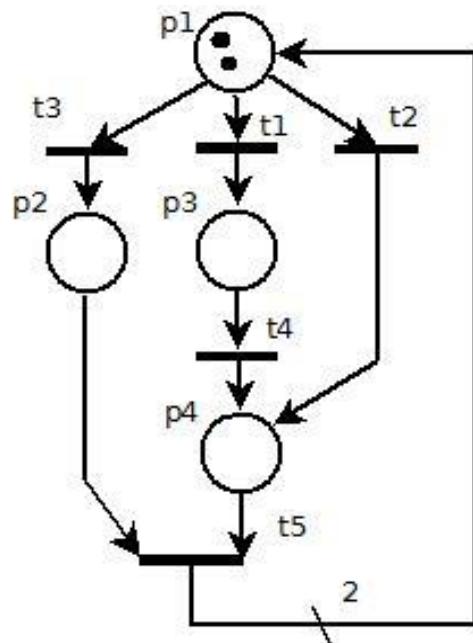
Path 4 : ???

$$M_1 = (2 \ 0 \ 0 \ 0)^T = 2p_1$$

$$M_{ref} = (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4$$

$$\delta = (1 \ 5 \ 1 \ 1 \ 1)^T$$

$$\Delta = (\infty \ \infty \ \infty \ \infty \ \infty)^T$$



Exercice

Path durations

Path 1 : T3(1)T2(5), duration = 5 TU

Path 2 : T2(5)T3(5), duration = 5 TU

Path 3 : T1(1)T4(2)T3(2), duration = 2 TU

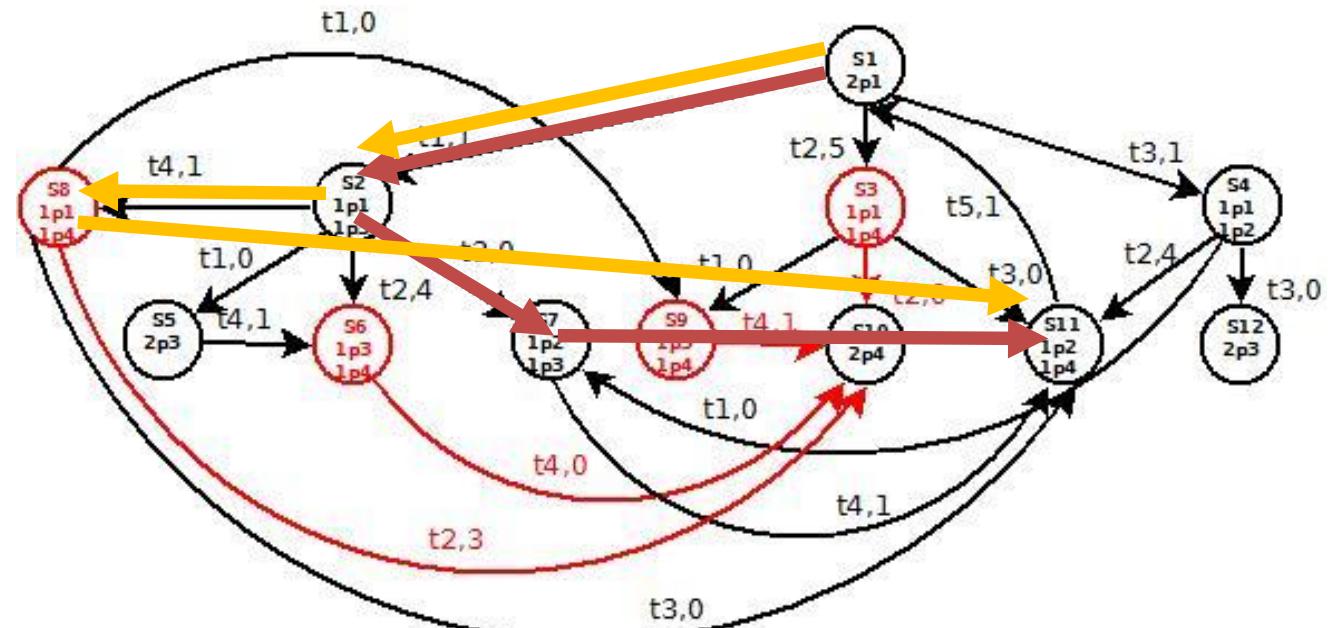
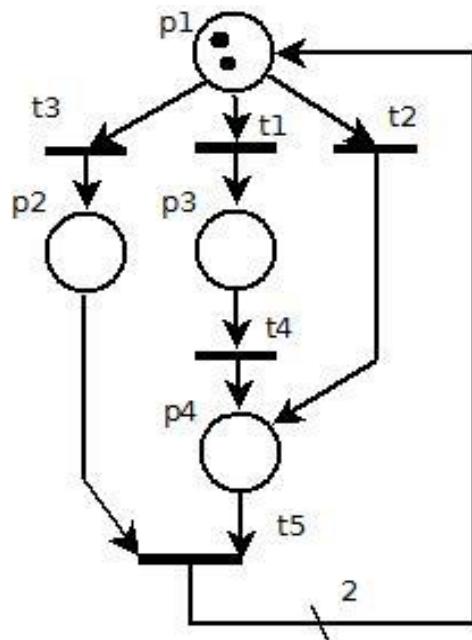
Path 4 : T1(1)T3(1)T4(2), duration = 2 TU

$$M_1 = (2 \ 0 \ 0 \ 0)^T = 2p_1$$

$$M_{ref} = (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4$$

$$\delta = (1 \ 5 \ 1 \ 1 \ 1)^T$$

$$\Delta = (\infty \ \infty \ \infty \ \infty \ \infty)^T$$



$$M_I = (2 \ 0 \ 0 \ 0)^T = 2p_1$$

$$M_{ref} = (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4$$

$$\delta = (1 \ 5 \ 1 \ 1 \ 1)^T$$

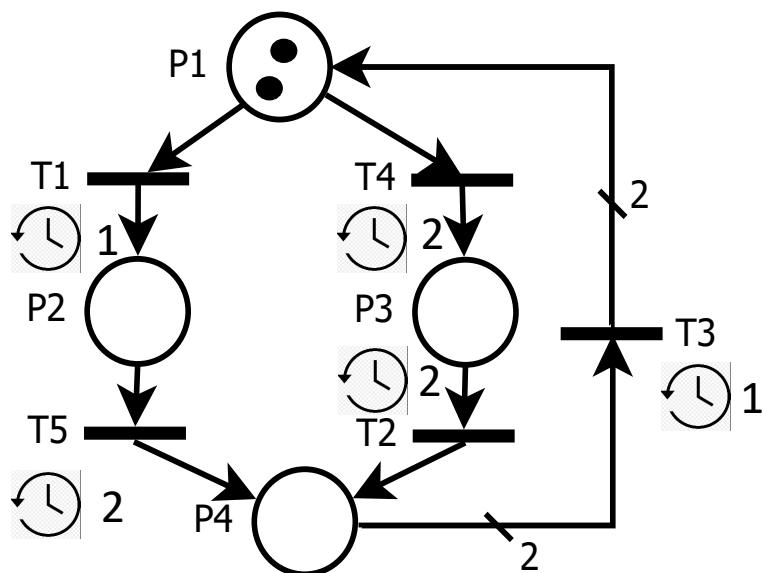
$$\Delta = (\infty \ \infty \ \infty \ \infty \ \infty)^T$$

TERG + Dijkstra algorithm
⇒ Paths of minimal duration

Path 3 : T1(1)T4(2)T3(2), duration = 2 TU

Path 4 : T1(1)T3(1)T4(2), duration = 2 TU

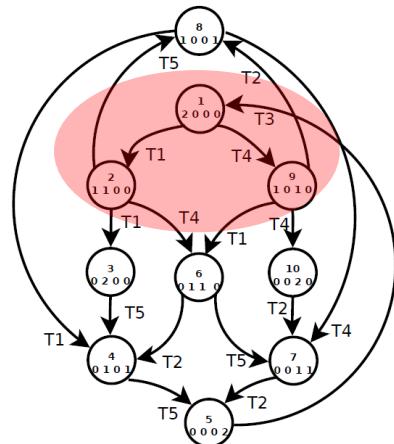
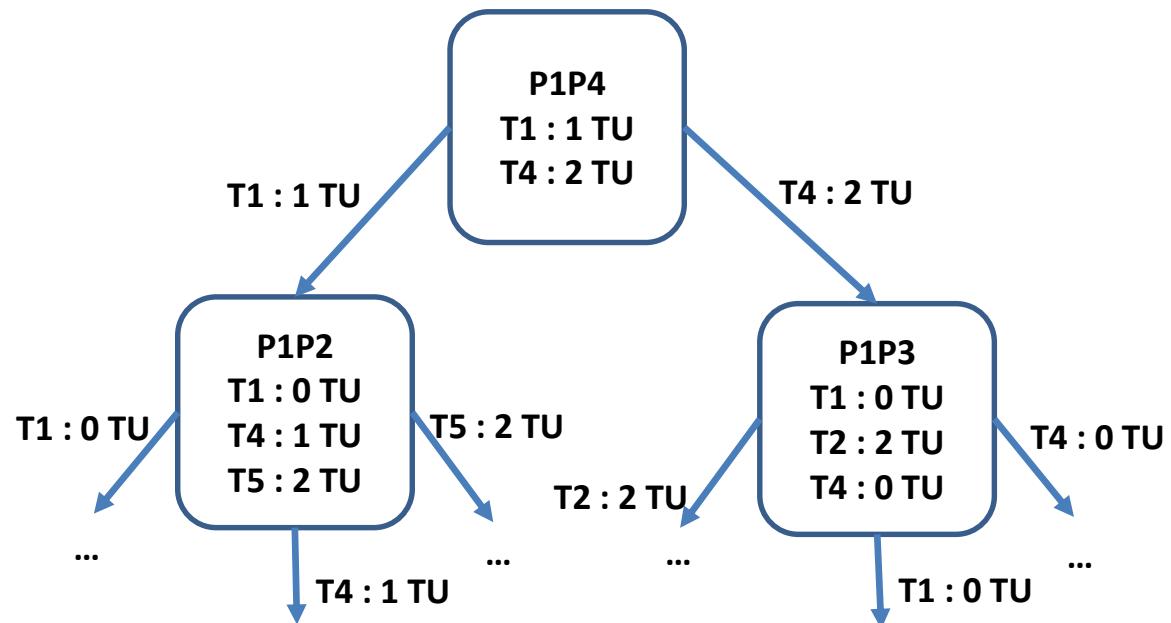
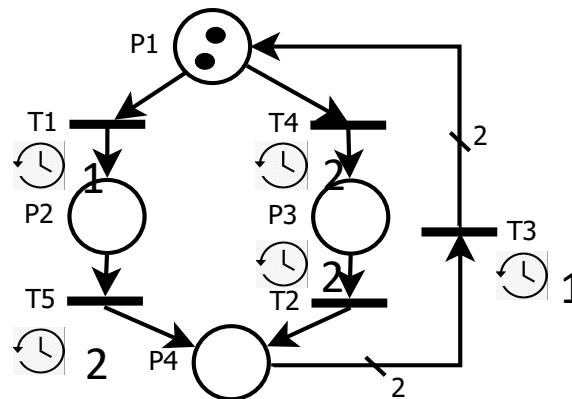
Example 1



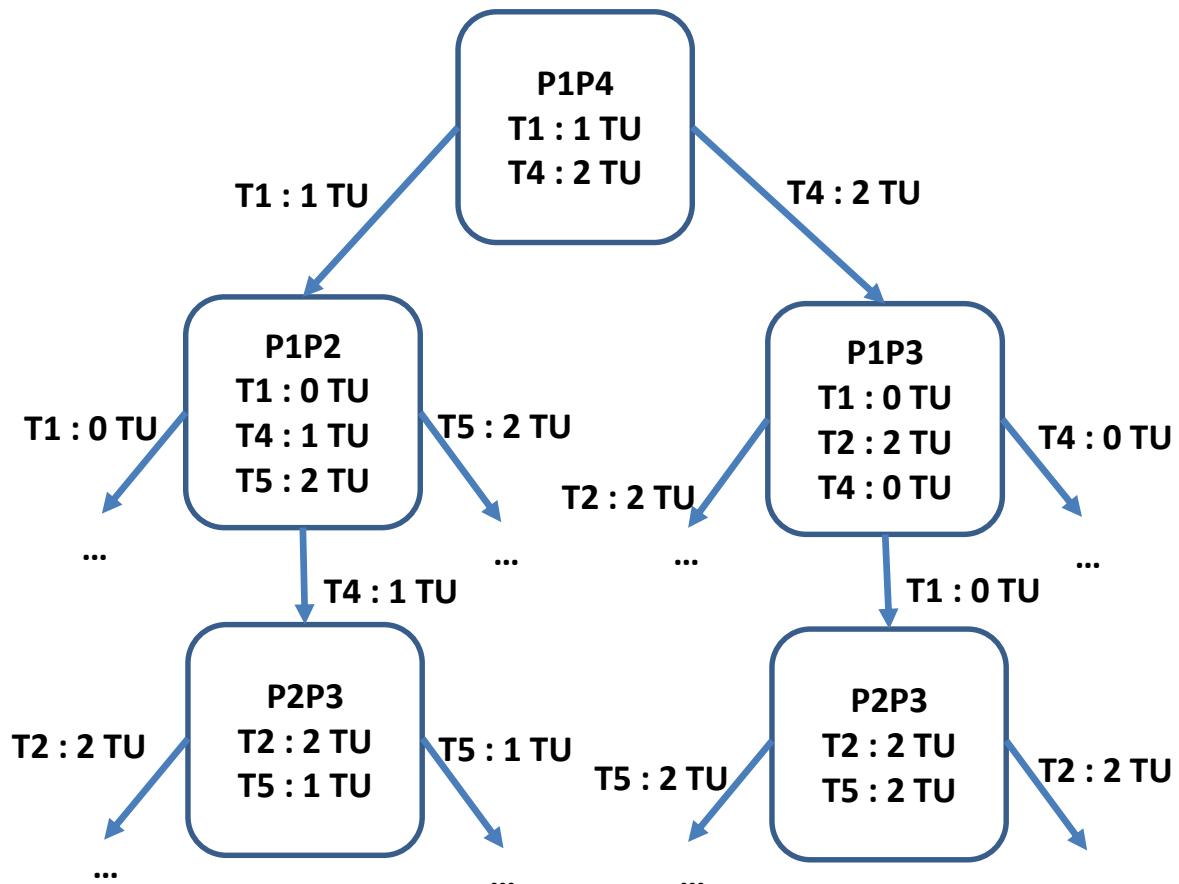
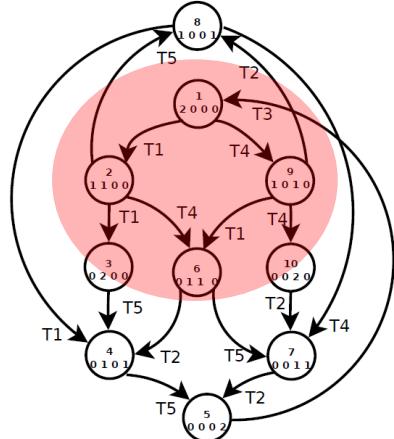
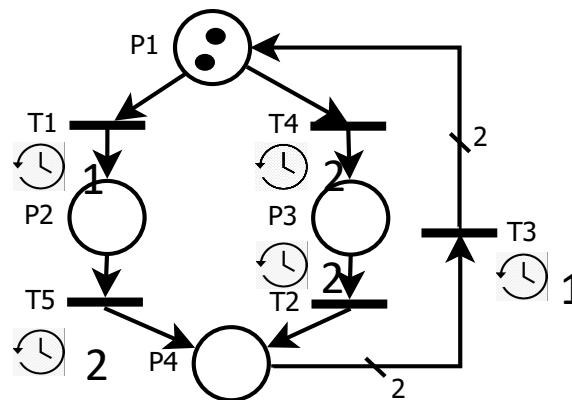
Exercice

TERG ???

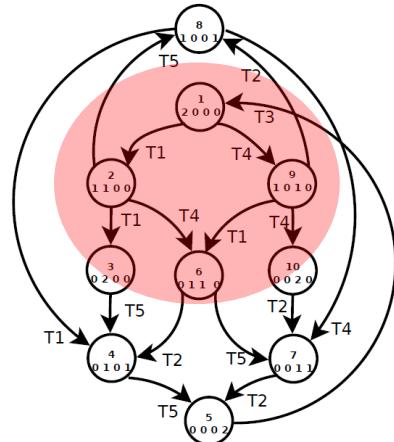
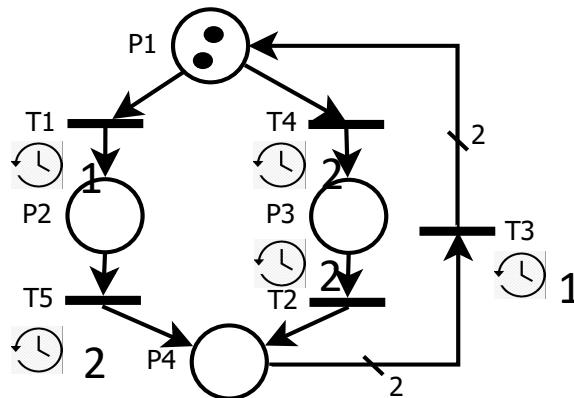
Exercice



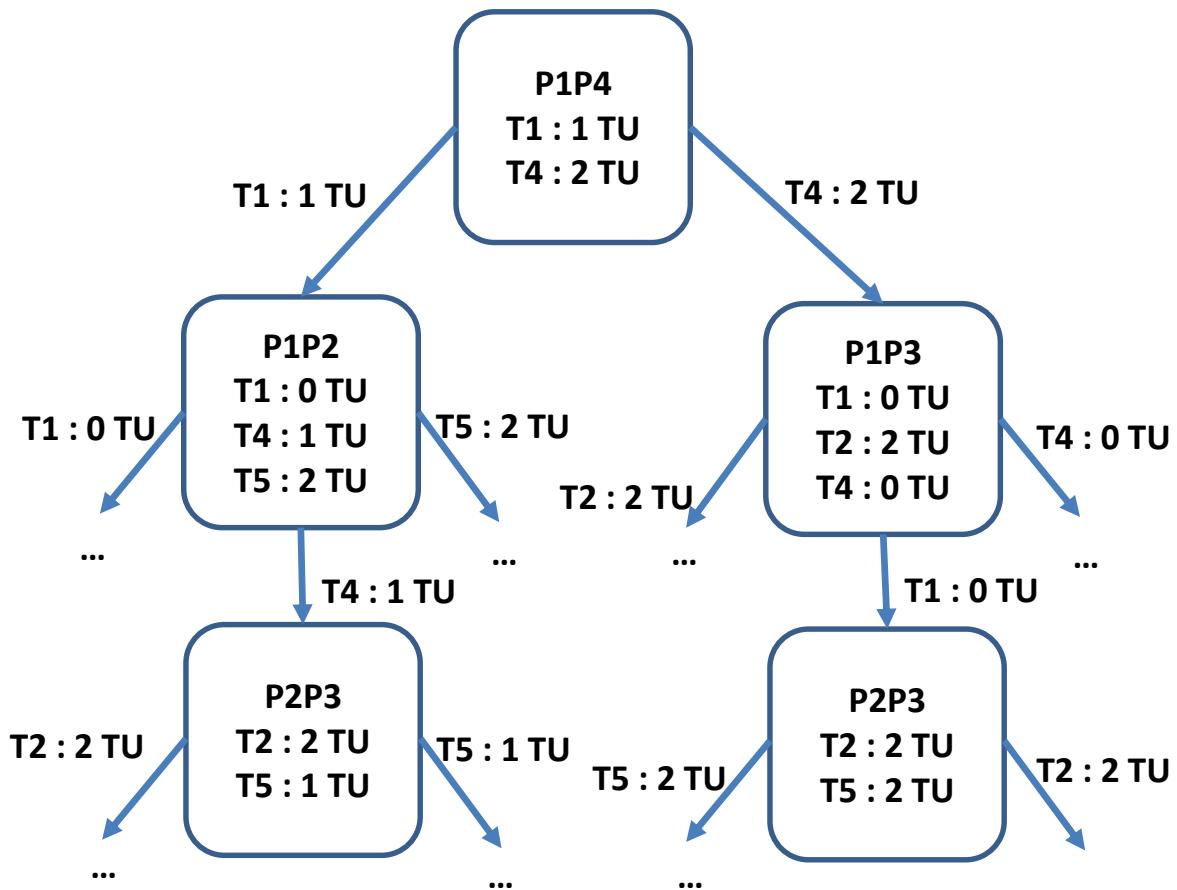
Exercice



Exercice 2



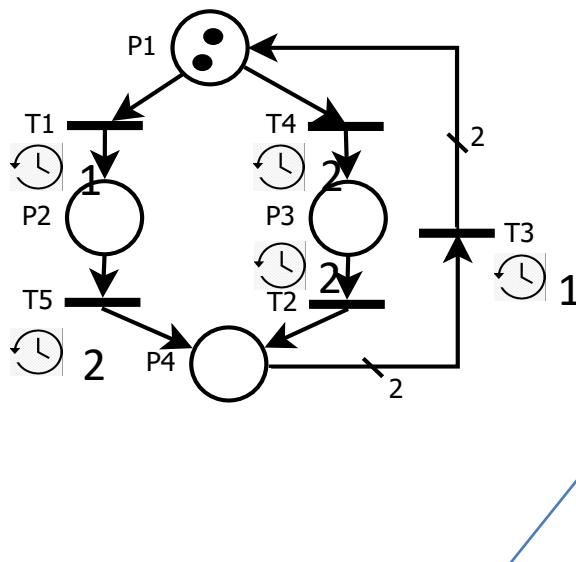
10 states



14 states

MATLAB

>> [TERG,TERG_time,S]=ApTERG_MACS(Wpr,Wpo,MI,delta,0)



Label of the transitions

TERG =

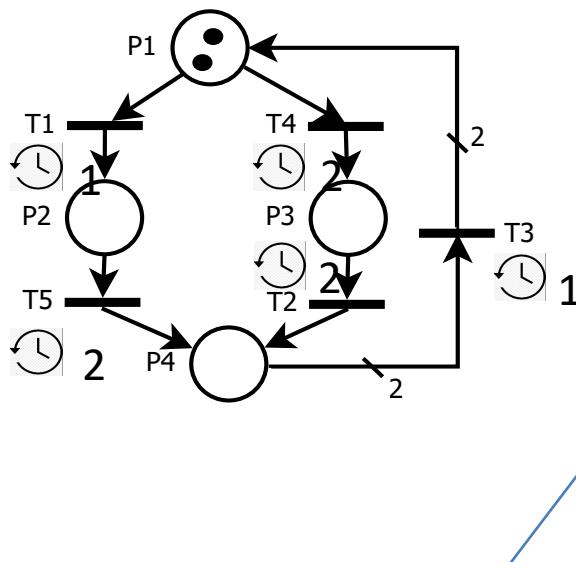
0	1	4	Inf										
Inf	0	Inf	1	4	5	Inf							
Inf	Inf	0	Inf	Inf	2	1	4	Inf	Inf	Inf	Inf	Inf	Inf
Inf	Inf	Inf	0	Inf	Inf	Inf	Inf	5	Inf	Inf	Inf	Inf	Inf
Inf	Inf	Inf	Inf	0	Inf	Inf	Inf	2	5	Inf	Inf	Inf	Inf
Inf	Inf	Inf	Inf	Inf	0	Inf	Inf	Inf	Inf	1	4	Inf	Inf
Inf	Inf	Inf	Inf	Inf	Inf	0	Inf	Inf	Inf	Inf	5	Inf	Inf
Inf	0	Inf	2	Inf	Inf	Inf	Inf						
Inf	0	Inf	Inf	Inf	Inf	2							
Inf	0	Inf	Inf	Inf	5								
Inf	0	Inf	Inf	2									
Inf	0	Inf	5										
Inf	0	2											
3	Inf	0											

S =

2	1	1	0	0	1	0	0	0	0	0	0	0	0
0	1	0	2	1	0	1	0	1	0	1	0	0	0
0	0	1	0	1	0	1	2	0	1	0	1	1	0
0	0	0	0	0	1	0	0	1	1	1	1	1	2

MATLAB

>> [TERG,TERG_time,S]=ApTERG_MACS(Wpr,Wpo,MI,delta,0)



Transition firing duration

TERG_time =

```

0   1   2   Inf  Inf
Inf  0   Inf  0   1   2   Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf
Inf  Inf  0   Inf  Inf  2   0   0   Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf
Inf  Inf  Inf  0   Inf  Inf  Inf  Inf  Inf  2   Inf  Inf  Inf  Inf  Inf  Inf
Inf  Inf  Inf  Inf  0   Inf  Inf  Inf  Inf  Inf  2   1   Inf  Inf  Inf  Inf
Inf  Inf  Inf  Inf  Inf  0   Inf  Inf  Inf  Inf  Inf  2   1   Inf  Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  0   Inf  Inf  Inf  Inf  Inf  Inf  0   0   Inf  Inf
Inf  Inf  Inf  Inf  Inf  Inf  Inf  0   Inf  Inf  Inf  Inf  Inf  Inf  Inf  0
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0   Inf  Inf  Inf  Inf  Inf  Inf  Inf  0
Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  Inf  0   Inf  Inf  Inf  Inf  Inf  Inf  1
Inf  0   Inf  Inf  Inf  Inf  Inf  2
Inf  0   Inf  Inf  Inf  Inf  2
Inf  0   Inf  Inf  Inf  0
1   Inf  0

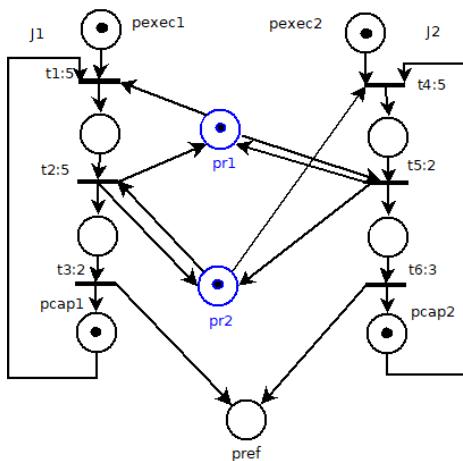
```

S =

```

2   1   1   0   0   1   0   0   0   0   0   0   0   0   0   0
0   1   0   2   1   0   1   0   1   0   1   0   1   0   0   0   0
0   0   1   0   1   0   1   2   0   1   0   1   0   1   1   1   0
0   0   0   0   0   0   1   0   0   1   1   1   1   1   1   2

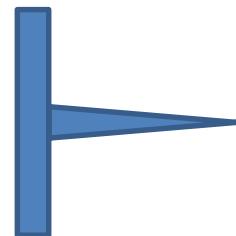
```



MATLAB

```
>> [TERG,TERG_time,S]=ApTERG_MACS(Wpr,Wpo,MI,delta,0)
```

TERG timed =



16 x 16



19 x 19

Timed Extended Reachability Graph

Proposition: If $S(M_i)$ is of finite cardinality N , then $S_E(M_i)$ is of finite cardinality N_E that satisfies:

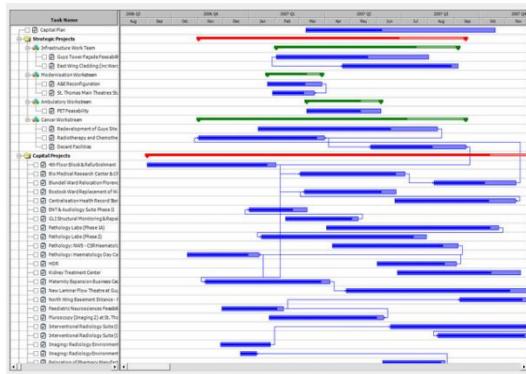
$$N \leq N_E \leq N \cdot ((D_m + 1) \cdot (D_M + 1))^{k \cdot q}$$

with $D_m = \max\{d_{min_j} : t_j \in T\}$ and $D_M = \max\{d_{MAX_j} : t_j \in T \text{ and } d_{MAX_j} < \infty\}$

The number of states of the Timed Extended Reachability Graph INCREASES EXPONENTIALLY with respect to the time specifications

D. Lefebvre, Approximated Timed Reachability Graphs for the robust control of discrete event systems, *Discrete Event Dynamic Systems: theory and applications*, 29(1), 31-56, 2019

Scheduling problems



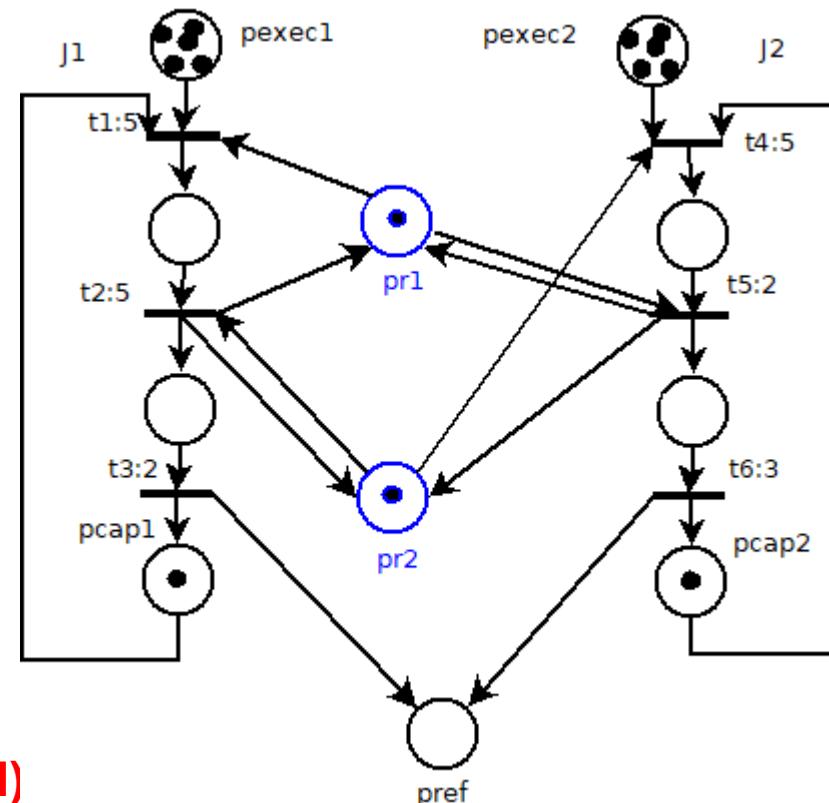
MATLAB

```
>> [Wpr,Wpo,delta,MI,Pjob]
= Exemple2_MACS(5,5)
```

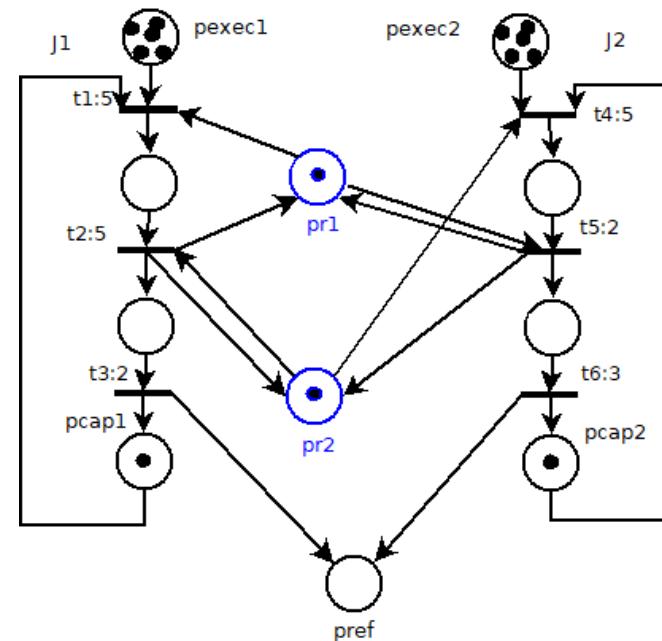
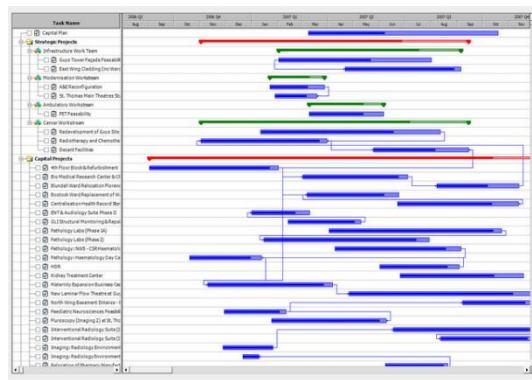
```
>>[RG,S]=RG_MACS(Wpr,Wpo,MI)
```

RG: 256 states

```
>>[TERG,TERG_time,S]=ApTERG_MACS(Wpr,Wpo,MI,delta,0)
TERG : 443 states
```



Scheduling problems

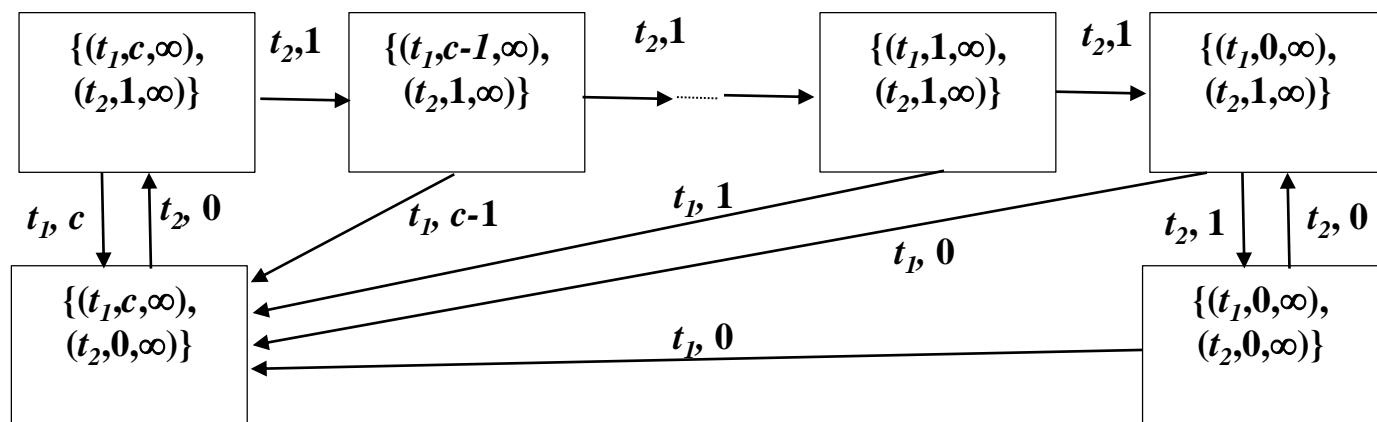
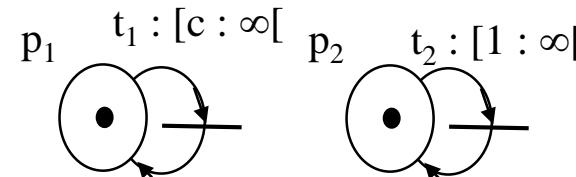


MATLAB

```
>>[precedesseur] = Dijkstra_predecessor_MACS([1:443],TERG_time,1)
>>MS=443
>>[seq,cost] = Dijkstra_seq_MACS(TERG,TERG_time,precedesseur,1,MS)
```

```
seq =[ 4 5 1 6 2 4 3 5 1 6 2 4 3 5 1 6 2
      4 3 5 1 6 2 4 3 5 6 1 2 3]
```

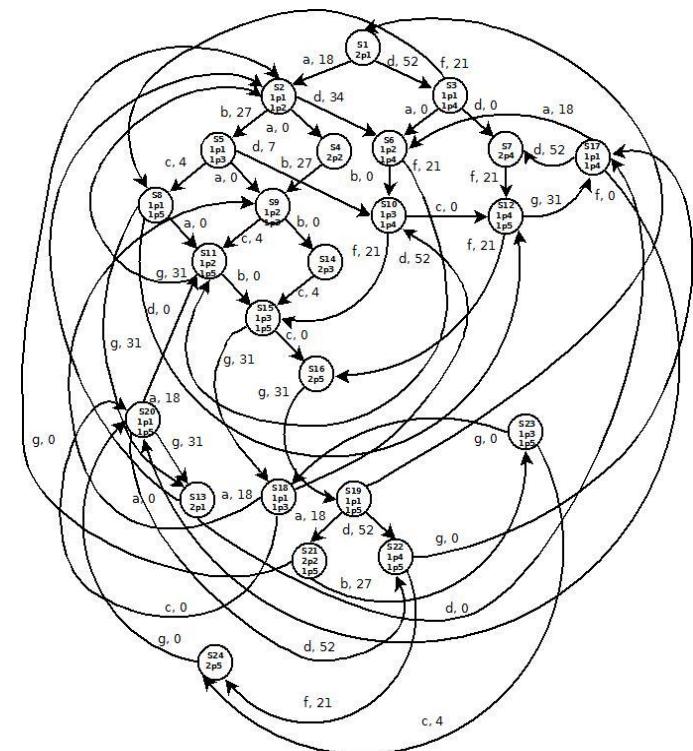
```
duration = 87
```



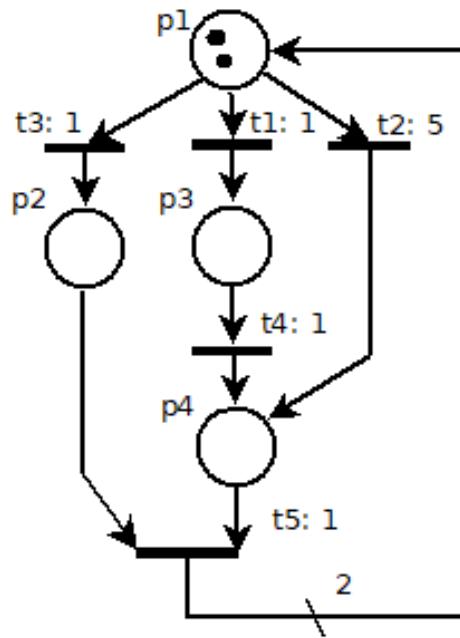
TERG tends to infinite size with respect to c

Outline

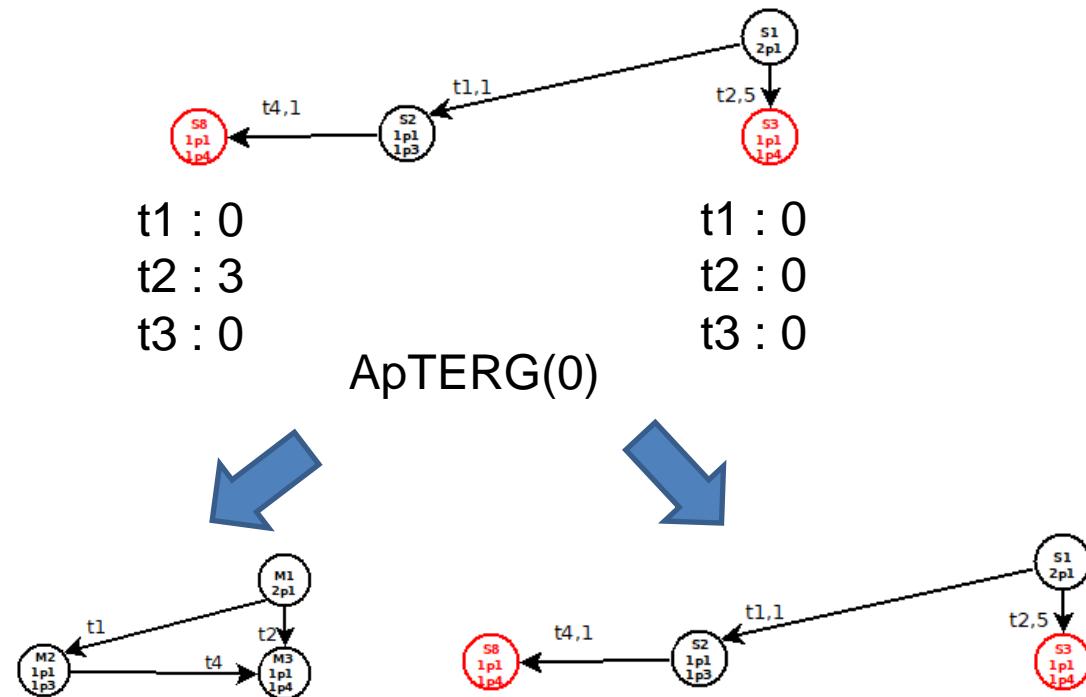
1. Problems and objectives
2. Modeling scheduling problems with T-TPN
3. Timed Extended Reachability Graph
- 4. Approximated Timed Extended Reachability Graph**
5. Beam Search
6. Conclusion and future works



ApTERG



$$\begin{aligned}
 M_1 &= (2 \ 0 \ 0 \ 0)^T = 2p_1 \\
 M_{ref} &= (0 \ 1 \ 0 \ 1)^T = 1p_2 \ 1p_4 \\
 \delta &= (1 \ 5 \ 1 \ 1 \ 1)^T \\
 \Delta &= (\infty \ \infty \ \infty \ \infty \ \infty)^T
 \end{aligned}$$



ApTERG(3.01)
to
ApTERG(∞)

ApTERG(0)
to
ApTERG(2.99)

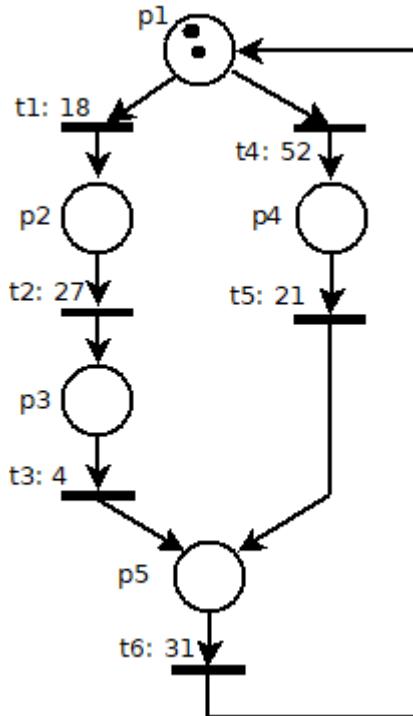
Approximated TERG (ApTERG)

Proposition: Let $\langle G, STI, M_I, A \rangle$ be a TPN system that behaves under earliest firing policy and its corresponding $\text{ApTERG}(\lambda) = \langle S_{ApE}(M_I), \Omega_{ApE}, B_{ApE}, S_0 \rangle$.

If the timed trajectory $(\sigma, M_j) = M(0) [(t_{j1}, \tau_1) > M(1) \dots > M(h-1) [(t_{jh}, \tau_h) > M(h)]$ with $M(0) = M_I$ and $t_{jk} \in T$, $k = 1, \dots, h$, is feasible in TPN system, then a path $S_0 S_1 \dots S_h$ exists in $\langle S_{ApE}(M_I), \Omega_{ApE}, B_{ApE}, S_0 \rangle$ st (1) $M(S_k) = M(k)$ for $k = 0, \dots, h$; (2) $\Omega_{ApE}(S_{k-1}, S_k) = t_{jk}$ and $|B_{ApE}(S_{k-1}, S_k).dt - (\tau_k - \tau_{k-1})| \leq \lambda.dt$ for $k = 1, \dots, h$.

If $S_0 S_1 \dots S_h$ is a path in $\text{ApTERG}(\lambda) = \langle S_{ApE}(M_I), \Omega_{ApE}, B_{ApE}, S_0 \rangle$ with S_0 the root node of the TERG then a timed trajectory $(\sigma, M_j) = M(0) [(t_{j1}, \tau_1) > M(1) \dots > M(h-1) [(t_{jh}, \tau_h) > M(h)]$ with (1) $M(k) = M(S_k)$, $k = 0, \dots, h$; (2) $t_{jk} = \Omega_{ApE}(S_{k-1}, S_k)$, $k = 1, \dots, h$; (3) $|\tau_1 - B_{ApE}(S_0, S_1).dt| \leq \lambda.dt$; (4) $|\tau_k - \tau_{k-1} - B_{ApE}(S_{k-1}, S_k).dt| \leq \lambda.dt$, $k = 1, \dots, h$; is feasible in TPN.

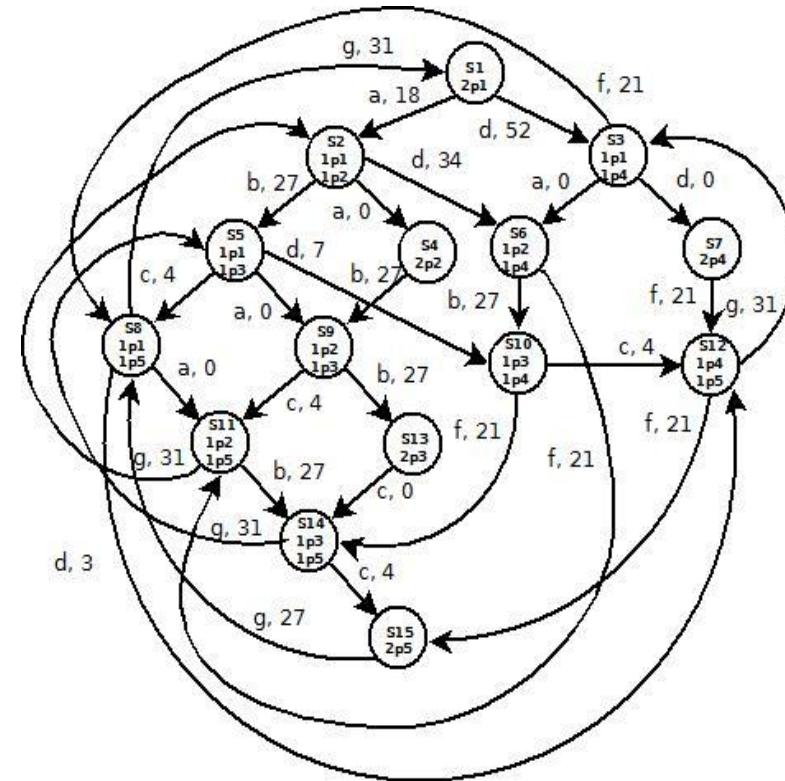
All feasible timed trajectories in time PN starting from M_I and executed under earliest firing policy are APPROXIMATED in $\text{ApTERG}(\lambda)$ and the error is bounded depending on the granularity parameter λ



$$M_I = (2 \ 0 \ 0 \ 0 \ 0)^T = 2p_1$$

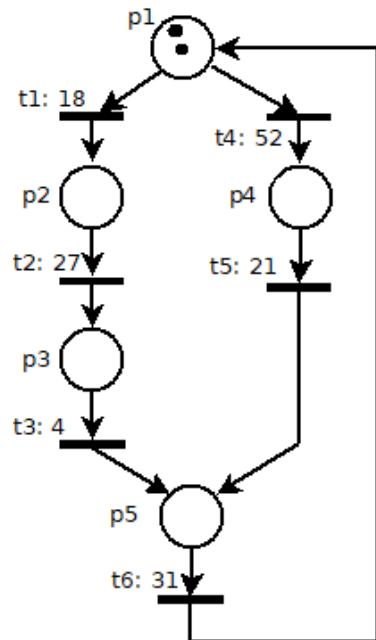
$$M_{Ref} = (0 \ 0 \ 0 \ 0 \ 2)^T = 2p_5$$

t_i	$t_1(a)$	$t_2(b)$	$t_3(c)$	$t_4(d)$	$t_5(f)$	$t_6(g)$
δ_i	18	27	4	52	21	31
Δ_i	∞	∞	∞	∞	∞	∞



ApTERG(∞) = Untimed RG: 15 states

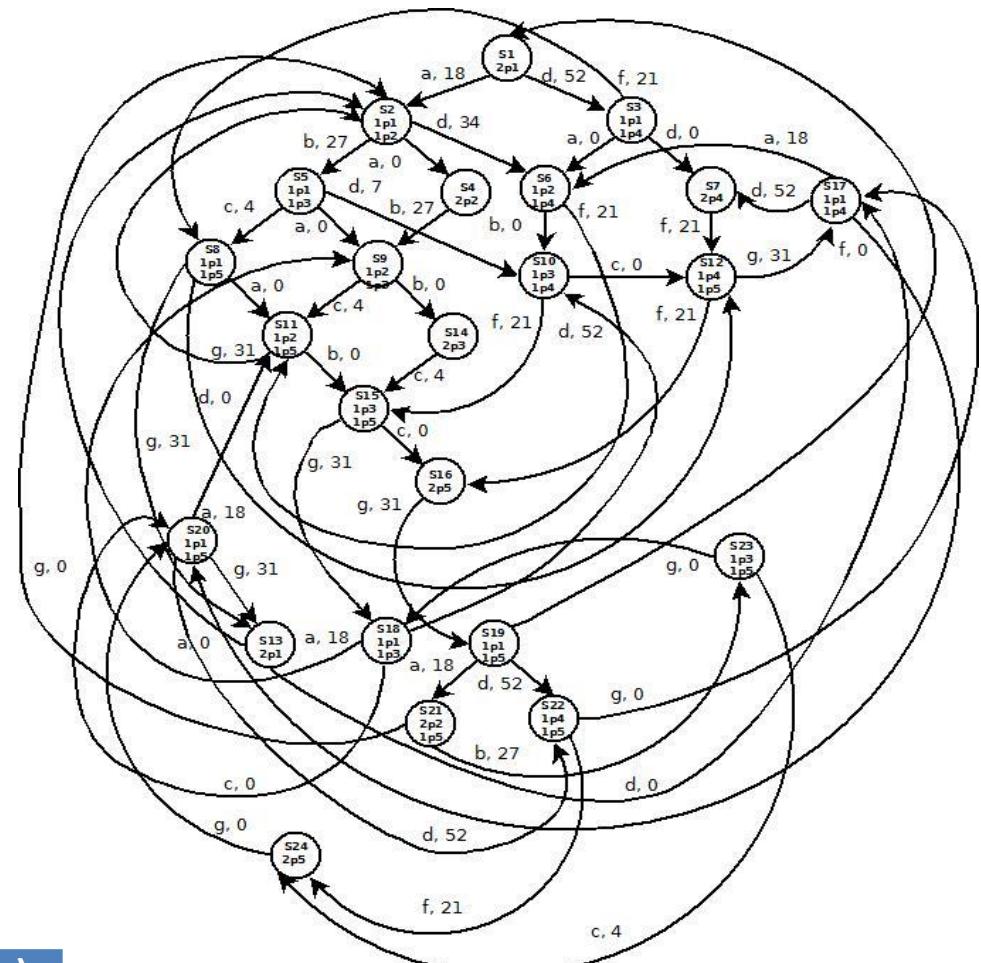
$$\sigma = T_1(18) T_2(45) T_3(49) T_4(52) T_4(73)$$



$$M_I = (2 \ 0 \ 0 \ 0 \ 0)^T = 2p_1$$

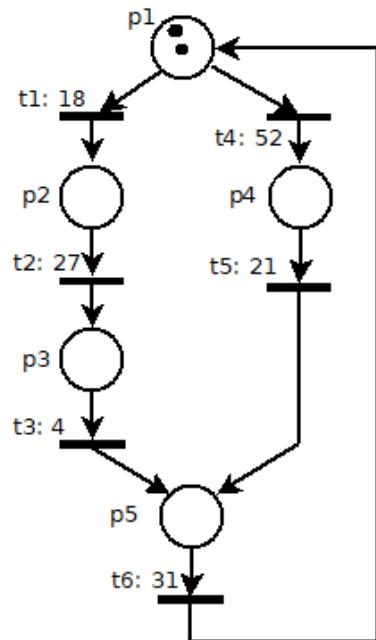
$$M_{Ref} = (0 \ 0 \ 0 \ 0 \ 2)^T = 2p_5$$

t_i	$t_1(a)$	$t_2(b)$	$t_3(c)$	$t_4(d)$	$t_5(f)$	$t_6(g)$
δ_j	18	27	4	52	21	31
Δ_i	∞	∞	∞	∞	∞	∞



ApTERG(30) : 24 states

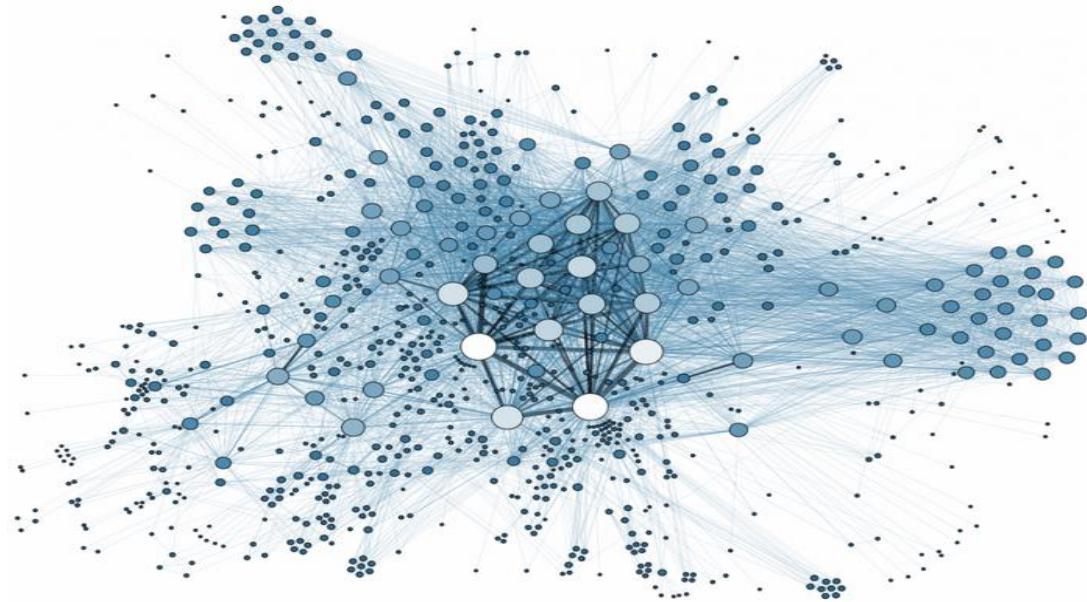
$\sigma = T_1(18) T_2(45) T_3(49) T_4(52) T_4(73)$



$$M_I = (2 \ 0 \ 0 \ 0 \ 0)^T = 2p_1$$

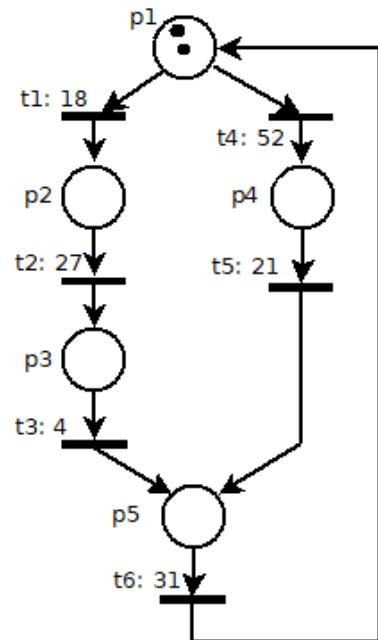
$$M_{Ref} = (0 \ 0 \ 0 \ 0 \ 2)^T = 2p_5$$

t_j	$t_1(a)$	$t_2(b)$	$t_3(c)$	$t_4(d)$	$t_5(f)$	$t_6(g)$
δ_j	18	27	4	52	21	31
Δ_i	∞	∞	∞	∞	∞	∞



ApTERG(0) =) : 327 states

$$\sigma = T_1(18)T_1(18)T_2(45) \\ T_2(45) \ T_3(49)T_3(49)$$



Number of states in ApTERG(λ)

λ	∞	50	40	30	20	10	1	0
N_{ApE}	15	18	19	24	37	42	214	327

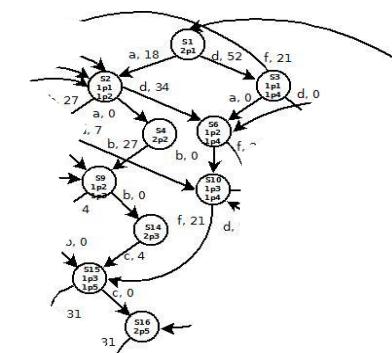
The number of states of ApTERG(λ) IS BOUNDED depending on λ

D. Lefebvre, Approximated Timed Reachability Graphs for the robust control of discrete event systems, *Discrete Event Dynamic Systems: theory and applications*, 29(1), 31-56, 2019

t_i	t_1	t_2	t_3	t_4	t_5	t_6
δ_j	18	27	4	52	21	31
Δ_i	∞	∞	∞	∞	∞	∞

Outline

1. Problems and objectives
 2. Modeling scheduling problems with T-TPN
 3. Timed Extended Reachability Graph
 4. Approximated Timed Extended Reachability Graph
 - 5. Beam Search**
 6. Conclusion and future works



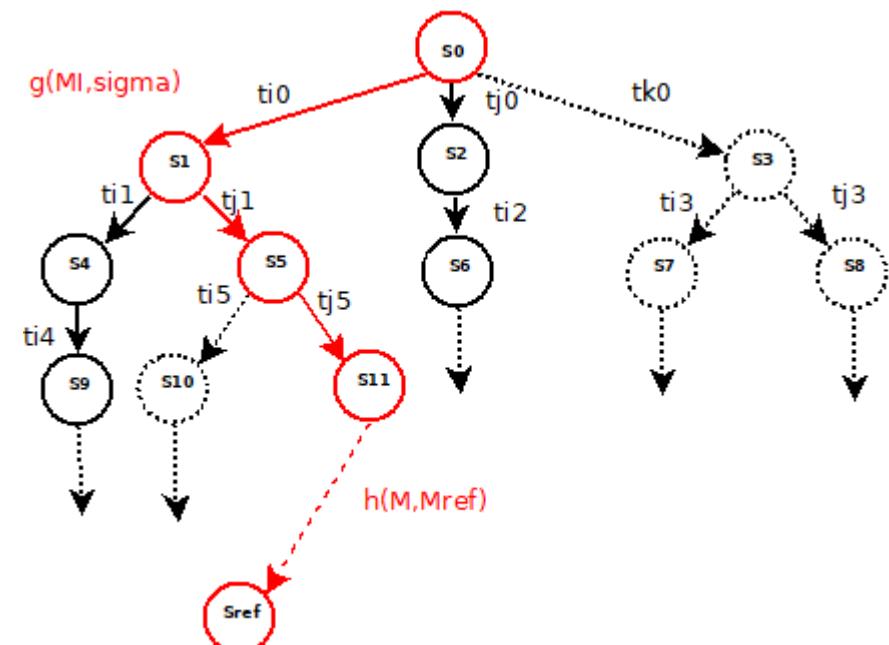
Local exploration in the untimed reachability graph driven by a cost function

$g(\sigma_1, M_1)$ = the cumulative duration of the already computed trajectory

$h(M, M_{ref})$ = estimation of the residual duration of (σ_2, M)

If $h(M, M_{ref})$ is a lower bound of the actual residual duration ($h(M, M_{ref})$ is said to be admissible), then the search find a least-cost trajectory

$$f(M, M, M_{ref}) = g(\sigma_1, M_1) + h(M, M_{ref})$$



Local exploration in the untimed reachability graph driven by a cost function

$g(\sigma_1, M_1)$ = the cumulative duration
of the already computed trajectory

$h(M, M_{ref})$ = estimation of the residual
duration of (σ_2, M)

$$f(M_1, M, M_{ref}) = g(\sigma_1, M_1) + h(M, M_{ref})$$

$$g(\sigma_1, M_1) = \tau_1 - \tau_0$$


$$(M_1, \tau_0) \xrightarrow{\sigma_1} (M, \tau) \quad DFS \text{ and } WFS : h(M, M_{ref}) = 0$$

$$\text{Another strategy : } h(M, M_{ref}) = \max\{ \chi^*(p_i, p_{ref}) : p_i \in P(M) \}$$

....

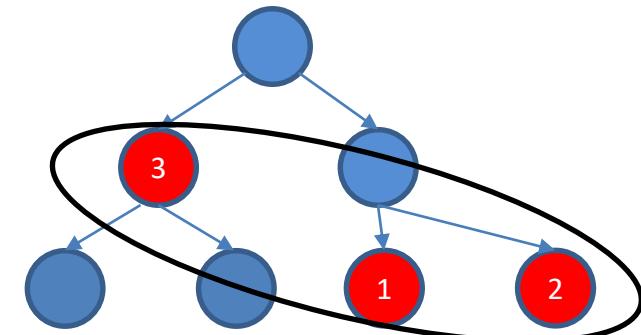
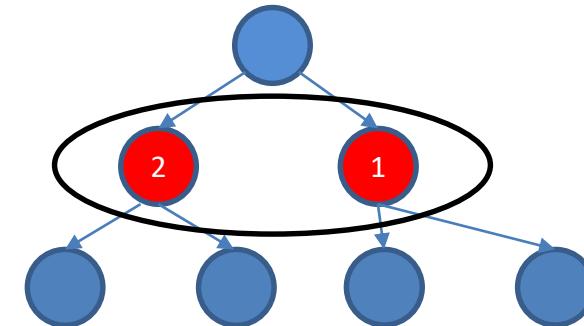
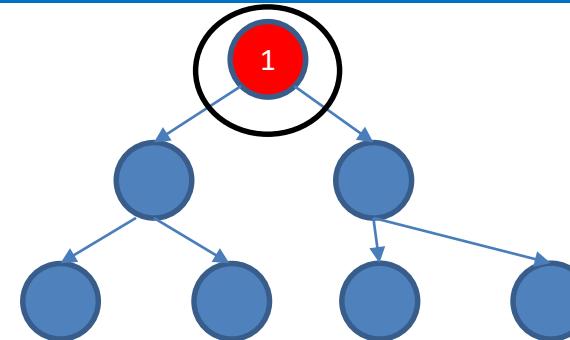
A* algorithm

At each iteration:

- (1) Rank the list of candidates
- (2) Expand the best candidate of the list of candidates
- (3) Store the new candidate(s)

Problem: the size of the list of candidates increases

P. E. Hart, N. J. Nilsson et B. Raphael, « A Formal Basis for the Heuristic Determination of Minimum Cost Paths », *IEEE Transactions on Systems Science and Cybernetics*, vol. 4, no 2, 1968, p. 100–107



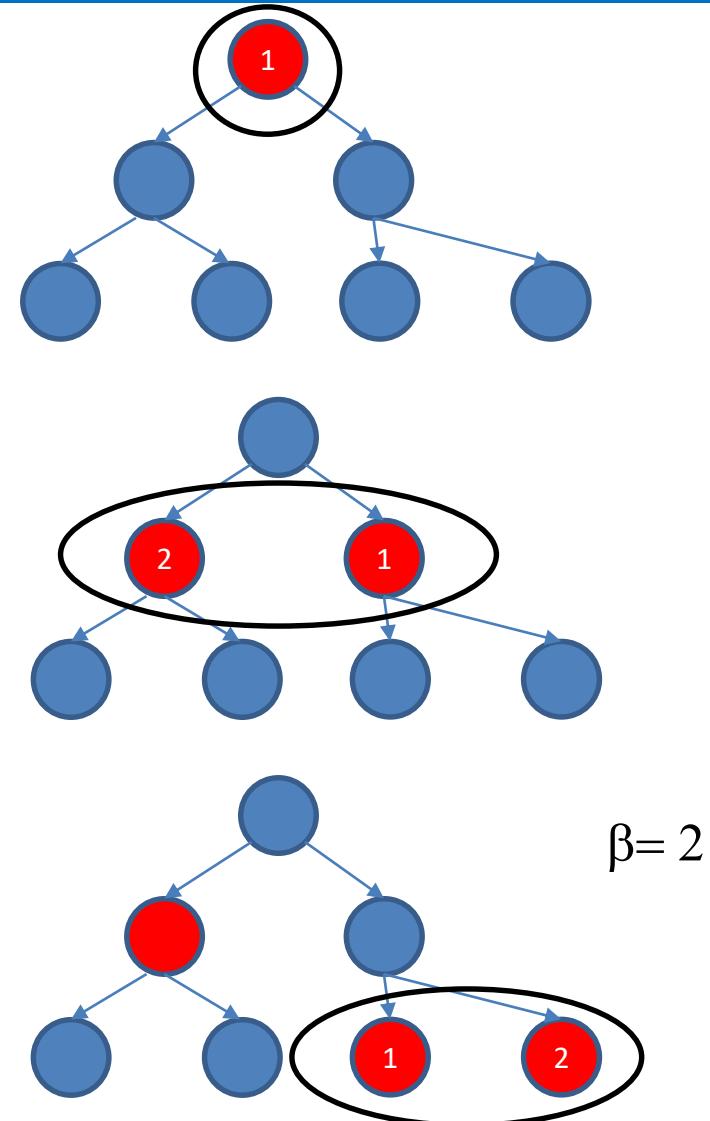
Beam Search Algorithm

At each iteration:

- (1) Rank the list of candidates
- (2) Expand the best candidate of the list of candidates
- (3) Store only the best β candidates selected with a global filter

Problem: Lack of diversity :
Concentrate the candidates issued from the same parent

Russell, S., & Norvig, P. (1995). *Artificial Intelligence: A Modern Approach*. Upper Saddle River, NJ (USA): Prentice Hall.



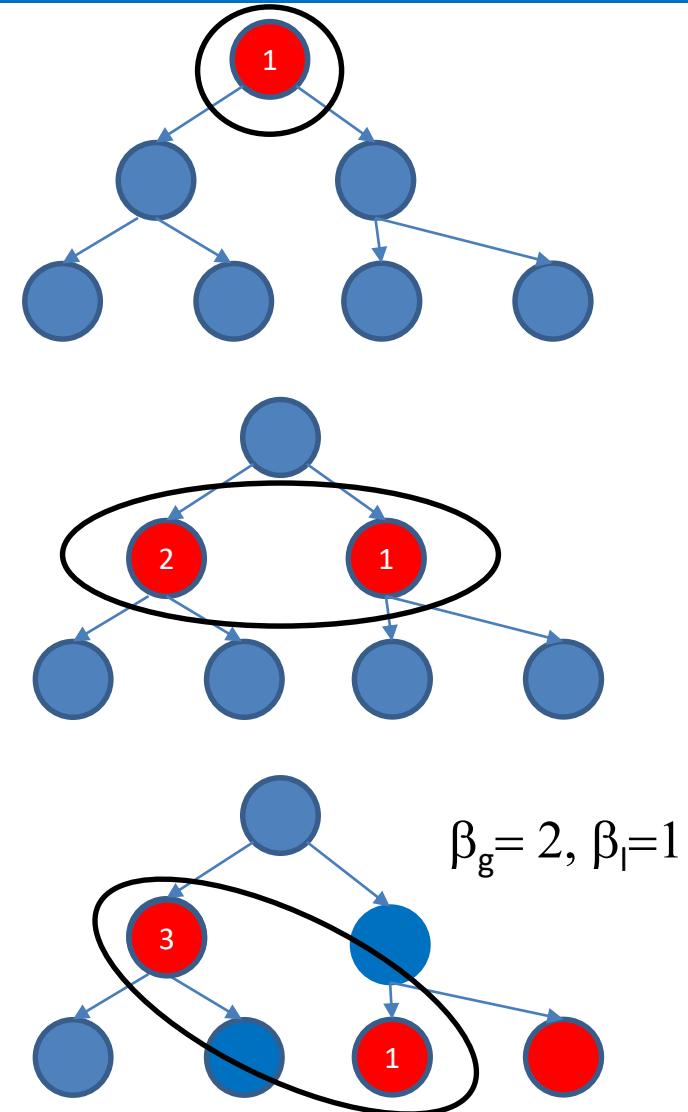
Filtered Beam Search

At each iteration:

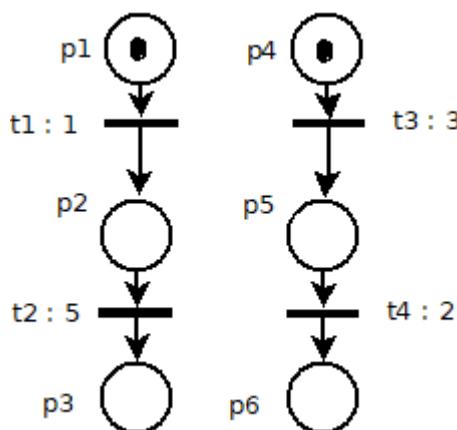
- (1) Rank the list of candidates
- (2) Expand the best candidate of the list of candidates
- (3) A local filter limits the number β_l of childrens of a given parent
- (4) A global filter selects the best β_g nodes at each iteration

Explores different regions of the RG

Mejía, G., & Niño, K. (2017). A new Hybrid Filtered Beam Search algorithm for deadlock-free scheduling of flexible manufacturing systems using Petri Nets. *Computers & Industrial Engineering*, 108, 165–176.



Exercice 3



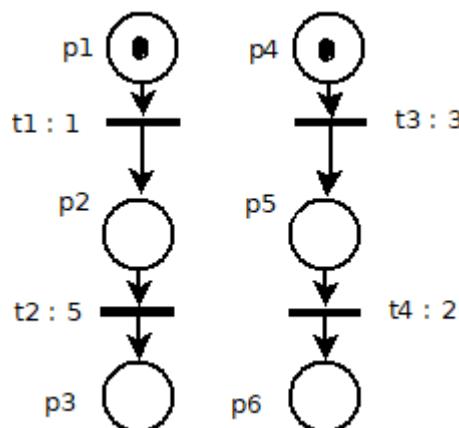
$$M_I = p_1 p_4$$

$$M_D = p_3 p_6$$

σ	t1	t3					
g/h/f	1/0/1	3/0/3					
σ
g/h/f
σ							
g/h/f							
σ							
g/h/f							
σ							
g/h/f							
σ							
g/h/f							

A* with $h(M, M_{ref}) = 0$

Exercice 3



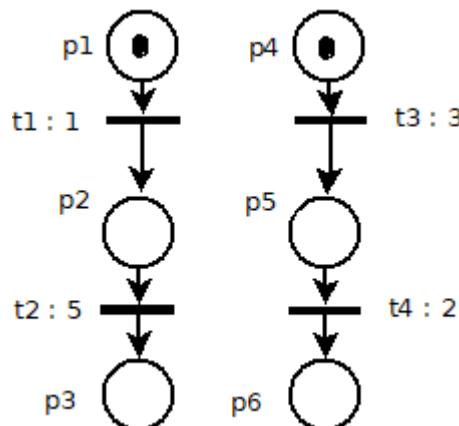
$$M_I = p_1 p_4$$

$$M_D = p_3 p_6$$

σ	t1	t3					
g/h/f	1/0/1	3/0/3					
σ	t3	t1t3	t1t2				
g/h/f	3/0/3	3/0/3	6/0/6				
σ
g/h/f
σ							
g/h/f							
σ							
g/h/f							

A* with $h(M, M_{ref}) = 0$

Exercice 3



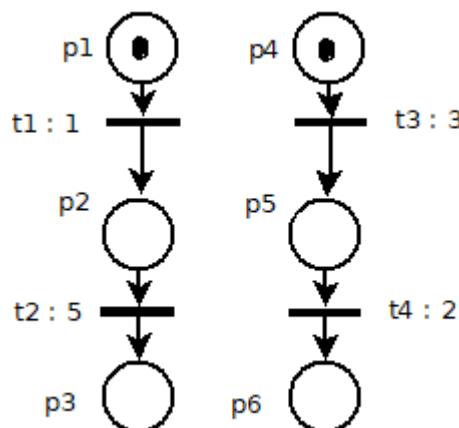
$$M_I = p_1 p_4$$

$$M_{MD} = p_3 p_6$$

σ	t1	t3					
$g/h/f$	1/0/1	3/0/3					
σ	t3	t1t3	t1t2				
$g/h/f$	3/0/3	3/0/3	6/0/6				
σ	t3t1	t1t3	t3t4	t1t2			
$g/h/f$	3/0/3	3/0/3	5/0/5	6/0/6			
σ	t1t3	t3t4	t3t1t4	t1t2	t3t1t2		
$g/h/f$	3/0/3	5/0/5	5/0/5	6/0/6	8/0/8		
σ
$g/h/f$

A* with $h(M, M_{ref}) = 0$

Exercice 3



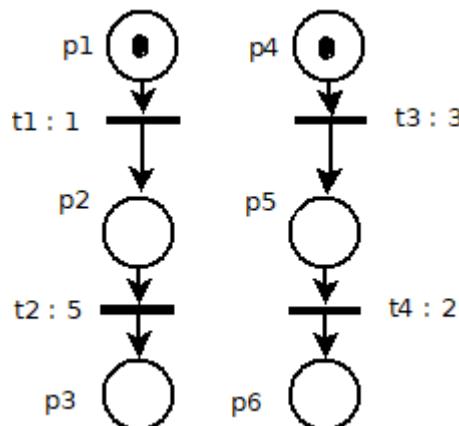
$$M_I = p_1 p_4$$

$$M_D = p_3 p_6$$

σ	t1	t3					
g/h/f	1/5/6	3/6/9					
σ	t1t3	t3	t1t2				
g/h/f	3/5/8	3/6/9	6/5/11				
σ	t1t3t2	t3	t1t3t4	t1t2			
g/h/f	6/2/8	3/6/9	5/5/10	6/5/11			
σ	t1t3t2t4	t3	t1t3t4	t1t2			
g/h/f	6/0/6	3/6/9	5/5/10	6/5/11			
σ							
g/h/f							

$$\text{A* with } h(M, M_{ref}) = \max\{ \chi^*(p_i, p_{ref}) : p_i \in P(M) \}$$

Exercice 3



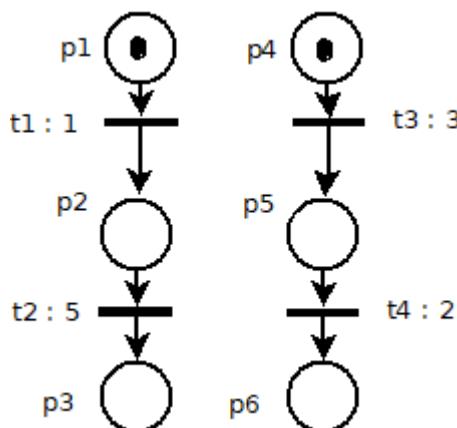
$$M_I = p_1 p_4$$

$$M_D = p_3 p_6$$

σ	t1	t3		X	X	X	X
g/h/f	1/5/6	3/6/9		X	X	X	X
σ	t3	t1t3	t1t2	X	X	X	X
g/h/f	3/6/9	3/5/8	6/5/11	X	X	X	X
σ	t1t3t2	t3	t1t3t4	X	X	X	X
g/h/f	6/2/8	3/6/9	5/5/10	X	X	X	X
σ	t1t3t2t4	t3	t1t3t4	X	X	X	X
g/h/f	6/0/6	3/6/9	5/5/10	X	X	X	X
σ				X	X	X	X
g/h/f				X	X	X	X

Beam with $\beta = 3$ and with $h(M, M_{ref}) = \max\{ \chi^*(p_i, p_{ref}) : p_i \in P(M) \}$

Exercice 3

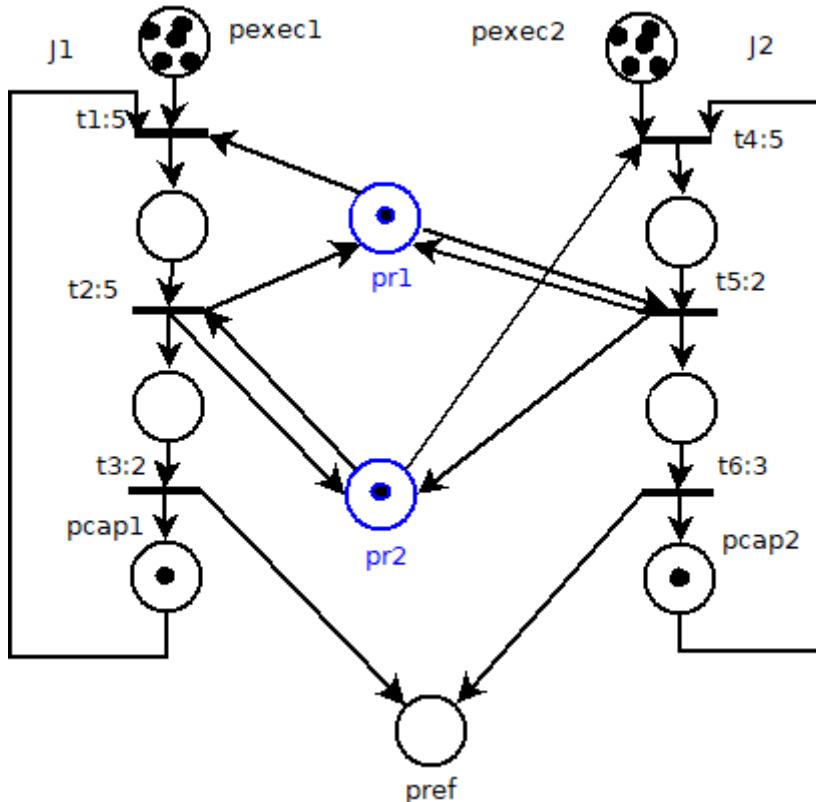


$$M_I = p_1 p_4$$

$$M_D = p_3 p_6$$

σ	t1	t3		X	X	X	X
g/h/f	1/5/6	3/6/9		X	X	X	X
σ	t1t3	t3	t3	X	X	X	X
g/h/f	3/5/8	3/6/9	3/6/9	X	X	X	X
σ	t1t3t2	t3		X	X	X	X
g/h/f	6/2/8	3/6/9		X	X	X	X
σ	t1t3t2t4	t3		X	X	X	X
g/h/f	6/0/6	3/6/9		X	X	X	X
σ				X	X	X	X
g/h/f				X	X	X	X

Beam with $\beta_g = 3$ and $\beta_l = 1$
and with $h(M, M_{ref}) = \max\{ \chi^*(p_i, p_{ref}) : p_i \in P(M) \}$



MATLAB

$A^* ???$

Beam with $\beta = 5 ???$

Beam with $\beta_g = 5$ and $\beta_l = 2 ???$

Beam with $\beta_g = 5$ and $\beta_l = 1 ???$

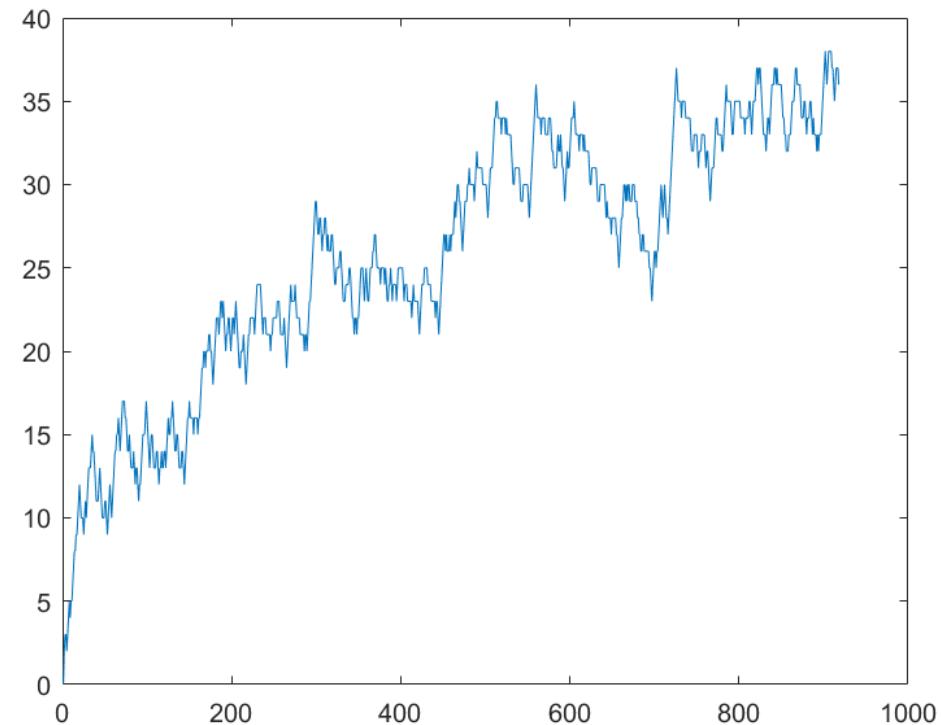
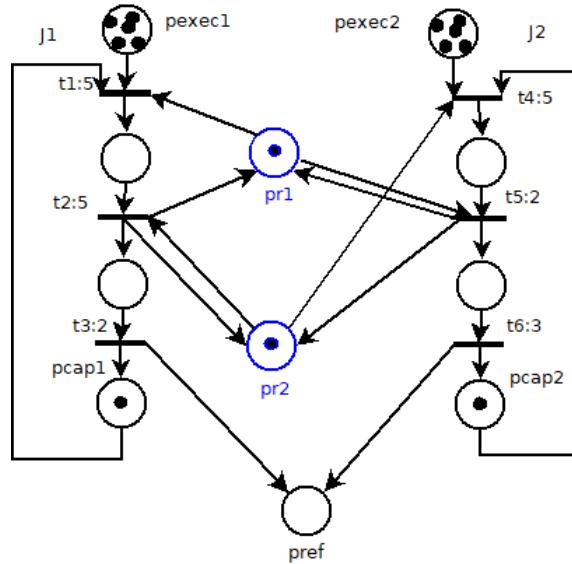
MATLAB : A*

`>>MD = [0 1 0 0 0 1 0 0 1 1 10]'`

`>> [seq,criteria]=Astar_MACS(Wpr,Wpo,Pjob,MI,MD,delta)`

Duration of the obtained sequence = 87

917 explored candidates



Size of list OPEN

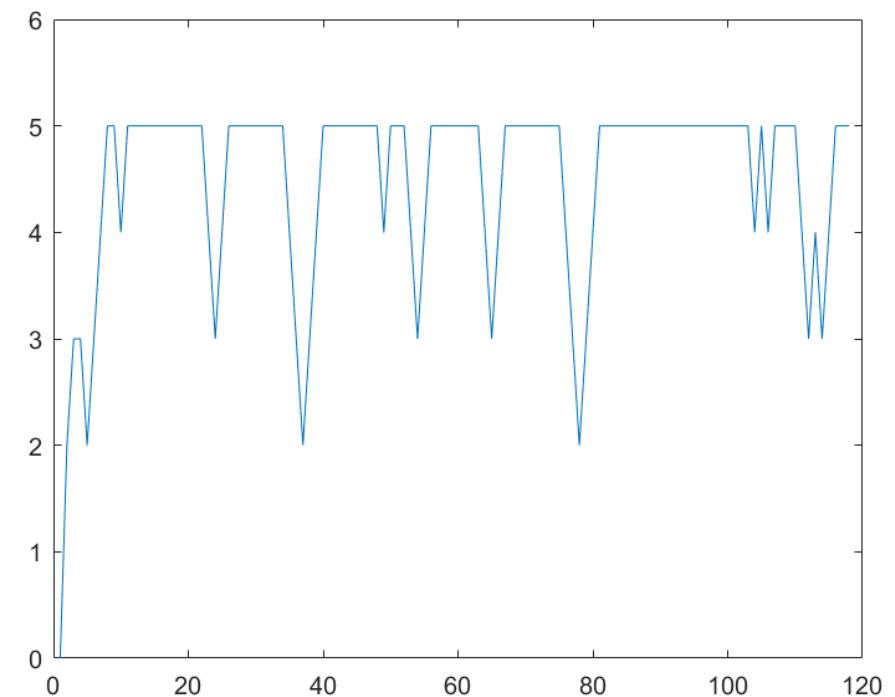
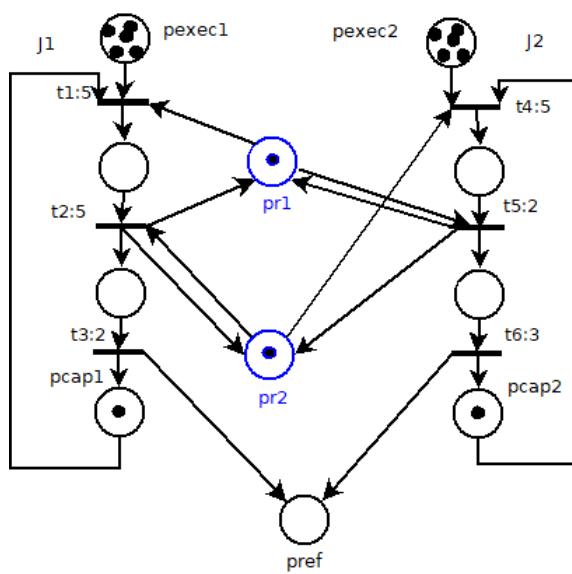
MATLAB : Beam with $\beta_g = 5$ and $\beta_l = 2$

>>MD = [0 1 0 0 0 1 0 0 1 1 1 0]'

>> [seq,criteria]=beam_MACS(Wpr,Wpo,Pjob,MI,MD,delta,5,2)

Duration of the obtained sequence = 88

118 explored candidates



Size of list OPEN

MATLAB : Beam with $\beta_g = 5$ and $\beta_l = 1$

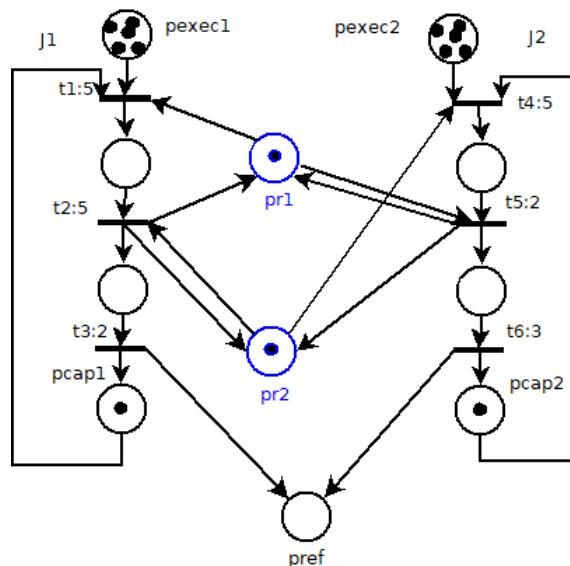
>>MD = [0 1 0 0 0 1 0 0 1 1 10]'

>> [seq,criteria]=beam_MACS(Wpr,Wpo,Pjob,MI,MD,delta,5,1)

Duration = X

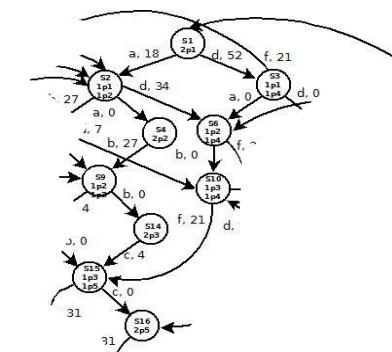
4 explored candidates

Algorithm fails

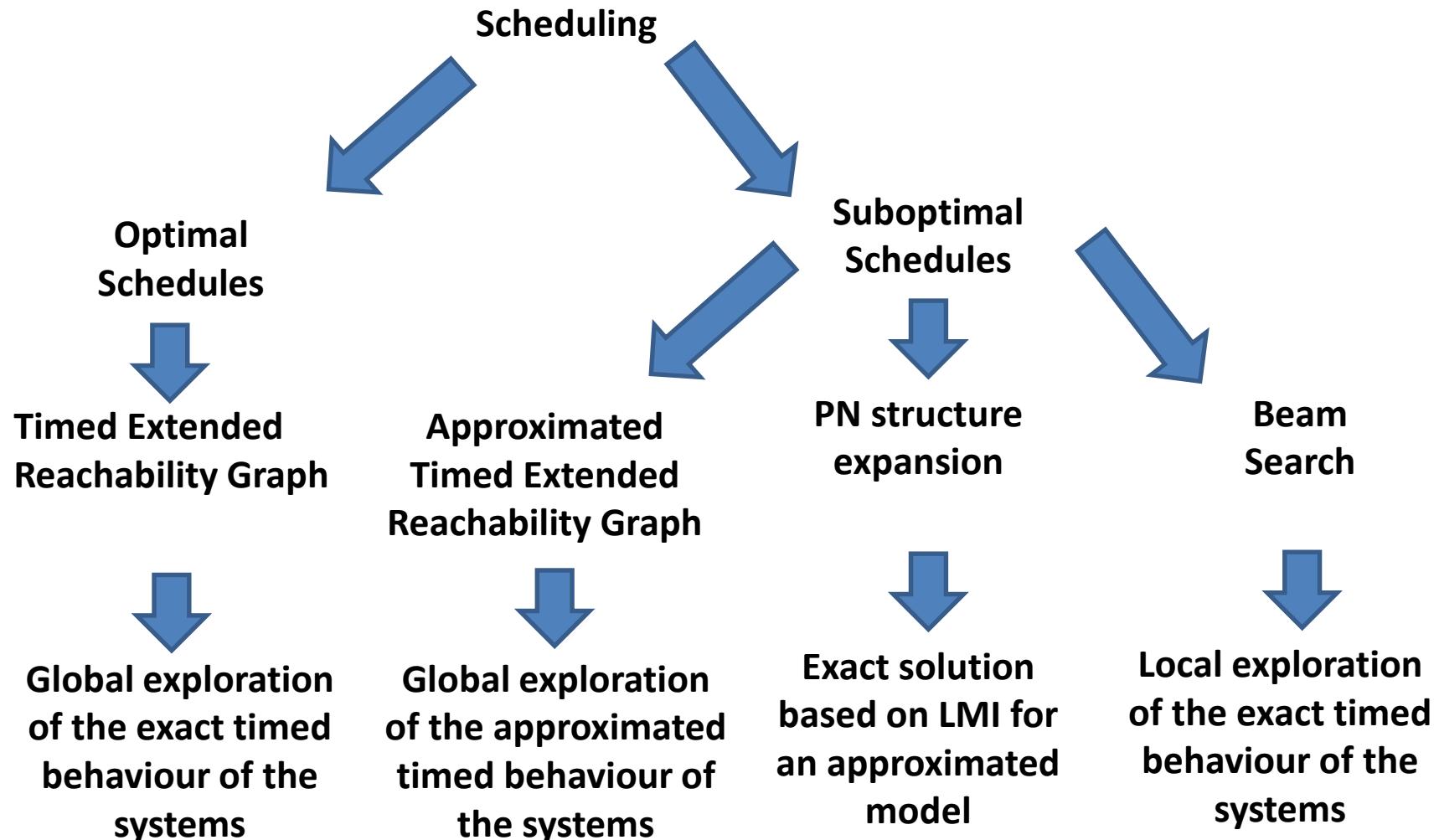


Outline

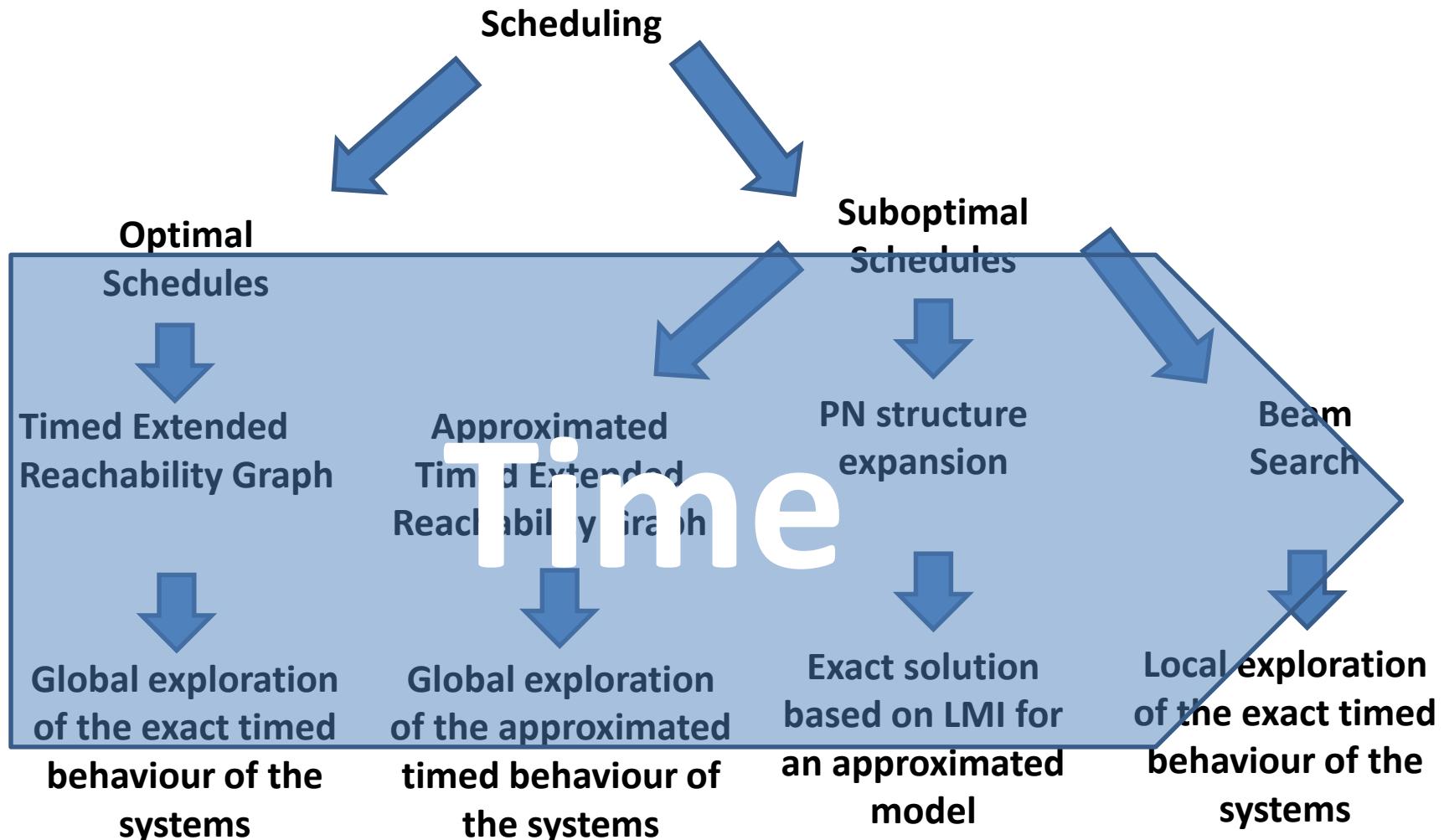
- 1. Problems and objectives**
- 2. Modeling scheduling problems with T-TPN**
- 3. Timed Extended Reachability Graph**
- 4. Approximated Timed Extended Reachability Graph**
- 5. Beam Search**
- 6. Conclusion and future works**



Conclusions and future works

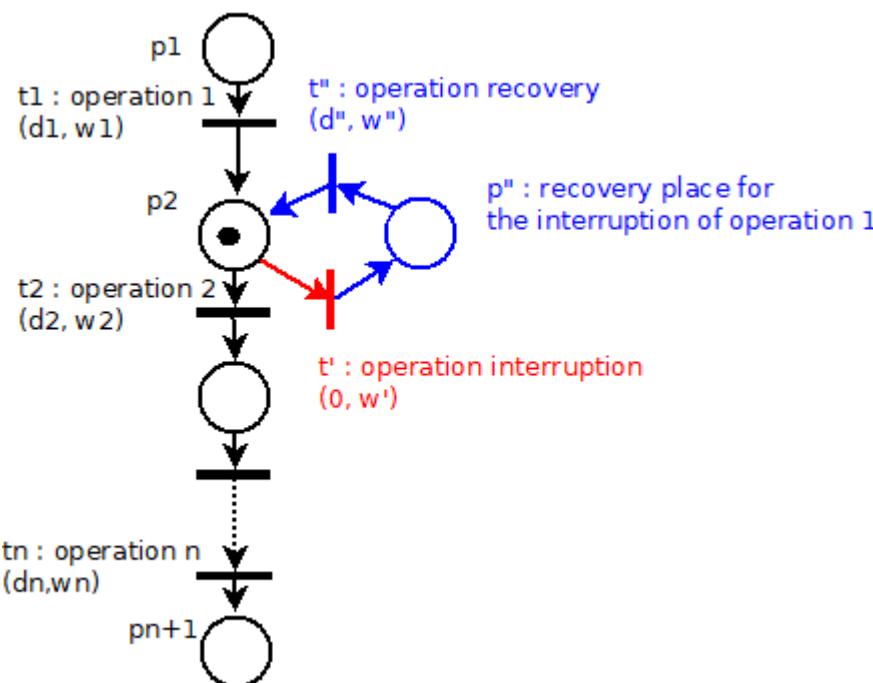


Conclusions and future works

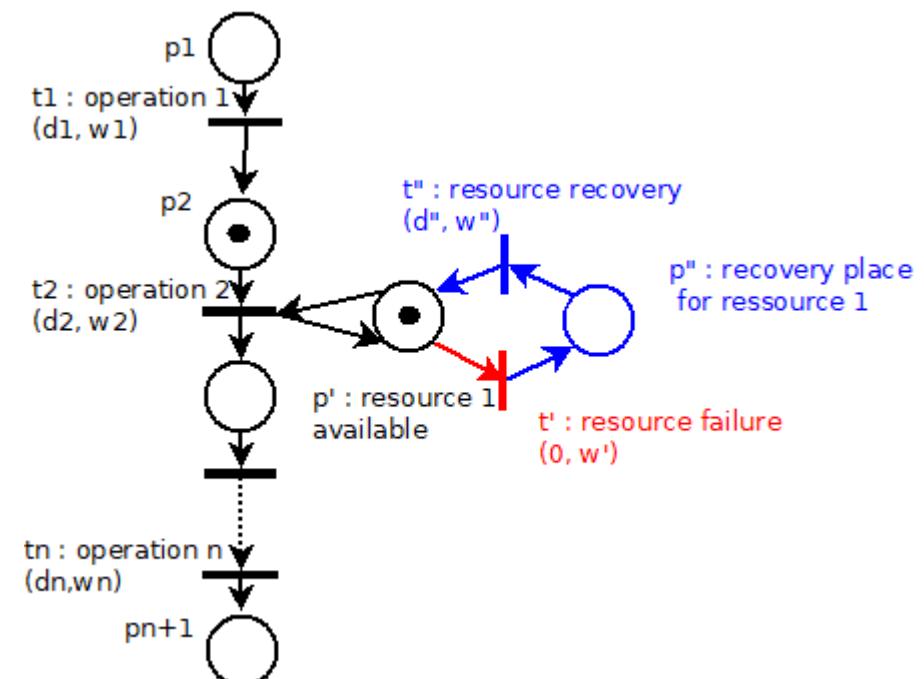


Future works

Robust scheduling in uncertain environments



Operation interruption



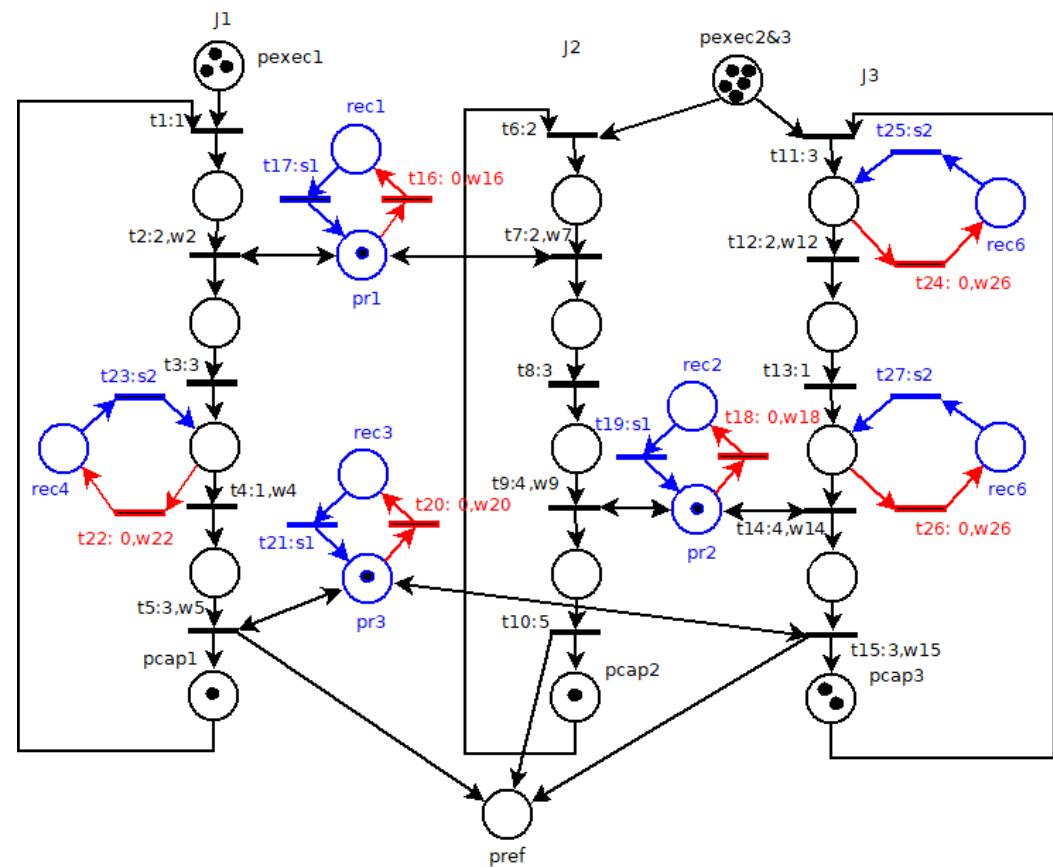
Unreliable resource

Future works

Robust scheduling in uncertain environments

2) Local exploration

$$\begin{aligned}
 f(M_I, M, M_{ref}) = & \\
 + (\sigma_1, M_I) & \text{ duration} \\
 + g_r(\sigma_1, M_I) & \text{ risk} \\
 + h_d(M, M_{ref}) & \text{ duration} \\
 + h_r(M, M_{ref}) & \text{ risk}
 \end{aligned}$$





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Marwa Taleb**

...



Some of our references in control theory

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Thank you !!!

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<http://www.univ-lehavre.fr/recherche/greah/>

