

# Chapitre 3

## Discrete event systems

### 3.1 Petri nets

#### 3.1.1 Definition

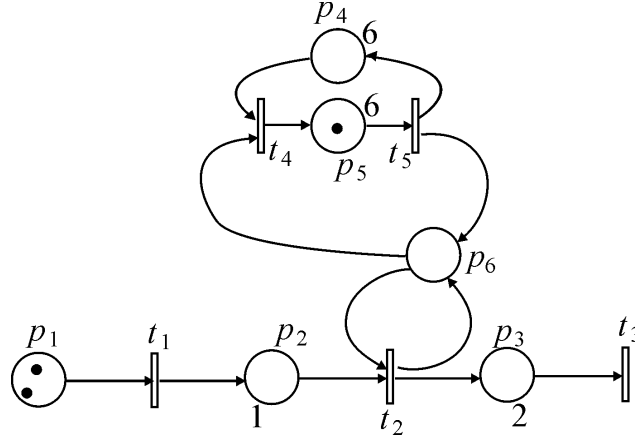
A *Petri net* is a graphical modeling language that can be used for the description of distributed systems or robots. A Petri net is a directed bipartite graph with two types of nodes : the *transitions* and *places*. The transitions are signified by bars and places are represented by circles. Arcs run either from a place to a transition or from a transition to a place. Places may contain *tokens*. A distribution of tokens over the places is called a *marking*. A transition may be *fired* if there is a token at the start of all its upstream place. When a transition fires one token is taken at each upstream place and one token is added to each downstream place.

**Example.** Consider a mission involving two sailboat robots. At initial time, both boats are in the harbor. In the way to reach the ocean, there is a channel with a strong current that changes with tides every 6 hours. It takes one hour for each boat to reach the channel from the harbor and two hours to reach the ocean from the entry of the channel, when the current is favorable. When the current is not favorable, the boats have to wait at the entrance of the channel. The situation can be represented by the Petri net of Figure 3.1. The meaning of places and transitions are given below

Transitions	Events
$t_1$	one boat leaves the port
$t_2$	one boat enters the channel
$t_3$	one boat reaches the ocean
$t_4$	low tide
$t_5$	high tide

Places	Tokens
$p_1$	boats in the harbor
$p_2$	boats moving toward the channel
$p_3$	boats inside the channel
$p_4$	favorable current
$p_5$	unfavorable current
$p_6$	allowed to enter the channel

**Formal definition.** A Petri net is a 3-tuple  $(\mathcal{P}, \mathcal{T}, w)$  where  $\mathcal{P}$  is a finite set of places,  $\mathcal{T}$  is a finite set of transitions,  $w : (\mathcal{P} \times \mathcal{T}) \cup (\mathcal{T} \times \mathcal{P}) \rightarrow \{0, 1\}$  is a set of arcs.



**Figure 3.1** – Petri net representing a mission involving two robots

We define the *preset* of a transition  $t$  as the set of all its upstream places

$$\text{preset}(t) = \{p \in \mathcal{P}, w(p, t) = 1\}.$$

We define the *postset* of a transition  $t$  as the set of all the downstream places

$$\text{postset}(t) = \{p \in \mathcal{P}, w(t, p) = 1\}.$$

For instance,  $\text{preset}(t_4) = \{p_4, p_6\}$  and  $\text{postset}(t_4) = \{p_5\}$ .

**Incidence matrix**

We define the *forward incidence matrix*  $\mathbf{W}^-$  and the *backward incidence matrix*  $\mathbf{W}^+$ , as the matrices with entries

$$w_{ij}^- = w(p_i, t_j) \text{ and } w_{ij}^+ = w(t_j, p_i).$$

For our example, we have

$$\mathbf{W}^- = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 1 & 0 \end{pmatrix} \text{ and } \mathbf{W}^+ = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 \end{pmatrix}.$$

The *incidence matrix* is defined by

$$\mathbf{W} = \mathbf{W}^+ - \mathbf{W}^-.$$

For our example, we have

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

A *marking* is vector of integers which represents the number of tokens that are assigned to each place. It corresponds to the state vector of the system. For our example, the initial marking is

$$\mathbf{m}_0 = (2 \ 0 \ 0 \ 0 \ 1 \ 0)^T.$$

Denote by  $\mathbf{s}$  the vector the  $i$ th entry of which represents the numbers of firing of transition  $t_i$  from the beginning. Then, the marking is

$$\mathbf{m} = \mathbf{m}_0 + \mathbf{W}\mathbf{s}.$$

For instance, if the following sequence of transition  $t_1, t_5, t_2$  has been fired, then we have

$$\mathbf{m} = \begin{pmatrix} 2 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{pmatrix} + \begin{pmatrix} -1 & 0 & 0 & 0 & 0 \\ 1 & -1 & 0 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \\ 0 \\ 1 \end{pmatrix}$$

Which means that one boat is still in the harbor and one is in the channel.

### 3.1.2 Invariance

The transition vector (or *sequence*)  $\mathbf{s}$  is said to be *t-invariant* if  $\mathbf{W}\mathbf{s} = \mathbf{0}$ . If a nonzero *t-invariant*  $\mathbf{s}$  exists then some periodic behaviors can appear in the Petri net. For our saiboat problem, we get

$$\mathbf{W}\mathbf{s} = \mathbf{0} \Rightarrow \mathbf{s} = \mathbf{0}$$

and thus the system does not have a periodic solution (i.e., it will be either stable or unstable).

The mark  $\mathbf{n}$  is said to be *p-invariant* if  $\mathbf{n}^T \mathbf{W} = \mathbf{0}^T$ . In such a case, for all  $\mathbf{s}$ , we have  $\mathbf{n}^T \mathbf{W}\mathbf{s} = \mathbf{n}^T (\mathbf{m} - \mathbf{m}_0) = \mathbf{0}$ . If there exists a p-invariant  $\mathbf{n}$  such that  $\forall i, n_i > 0$  then the number of tokens has an upper bound. The Pétro net is said to be *bounded*.

**Example.** The Pétro net of Figure 3.2 represents a conveyor which is a mechanism to transport materials from one location to another. The conveyor has a single trolley. The place  $p_2$  means that the trolley is empty. Place  $p_1$  and  $p_2$  correspond to an object of type 1 and 2 is the trolley. Here, we have

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

To get the t-invariant sequences, we compute the kernel of  $\mathbf{W}$  :

$$\begin{pmatrix} 1 & 0 & -1 & 0 \\ -1 & -1 & 1 & 1 \\ 0 & 1 & 0 & -1 \end{pmatrix} \begin{pmatrix} t_1 \\ t_2 \\ t_3 \\ t_4 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}.$$

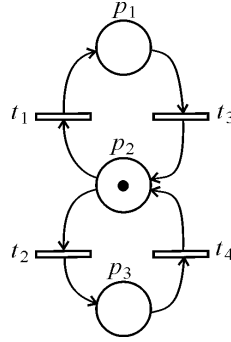


Figure 3.2 – Pétri net of a conveyor

We get  $t_1 = t_3$ , and  $t_2 = t_4$ . For instance, the sequence  $\mathbf{s} = (2, 5, 2, 5)^T$  is  $t$ -invariant. To compute the  $p$  invariants, we solve

$$\begin{pmatrix} n_1 & n_2 & n_3 \end{pmatrix} \begin{pmatrix} 1 & 0 & -1 & 0 \\ -1 & -1 & 1 & 1 \\ 0 & 1 & 0 & -1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \end{pmatrix}$$

and we get  $n_1 = n_2 = n_3$ . It means that

$$\begin{pmatrix} 1 & 1 & 1 \end{pmatrix} (\mathbf{m} - \mathbf{m}_0) = 0$$

i.e., the number of tokens is constant. Note also that, since  $\mathbf{n}^T = (1 \ 1 \ 1) > 0$ , we can conclude that the Pétri net is bounded.

### 3.2 Max-plus algebra

#### 3.2.1 Example 1

Consider a simple public transportation system. There are two stations where passengers can change lines, and four buses connecting the two stations. The corresponding Pétri net is represented by Figure ???. Each token corresponds to a bus and structure of the Pétri net shows that a synchronization exists between the buses at each station. The firing times of transitions  $t_1$  and  $t_2$  represent the departure times of the buses in station 1 and 2, respectively. These times can therefore be interpreted as the “time table” for the simple public transportation system.

Denote by  $x_i(k)$  the time at which the transition  $i$  is fired for the  $k$ th time.

– If the initial condition is  $x_1(1) = x_2(1) = 0$ , the time table is as follows

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 5 \\ 3 \end{pmatrix}, \begin{pmatrix} 8 \\ 8 \end{pmatrix}, \begin{pmatrix} 13 \\ 11 \end{pmatrix}, \begin{pmatrix} 16 \\ 16 \end{pmatrix}, \dots$$

– If the initial condition is  $x_1(1) = 1, x_2(1) = 0$ , the time table is

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 5 \\ 4 \end{pmatrix}, \begin{pmatrix} 9 \\ 8 \end{pmatrix}, \begin{pmatrix} 13 \\ 12 \end{pmatrix}, \begin{pmatrix} 17 \\ 16 \end{pmatrix}, \dots$$

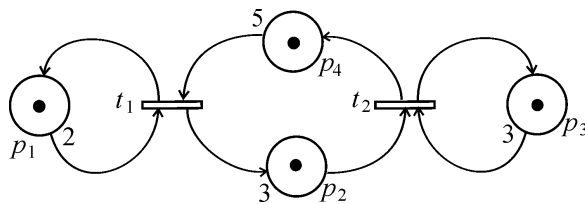


Figure 3.3 – Timed event graph of the bus network

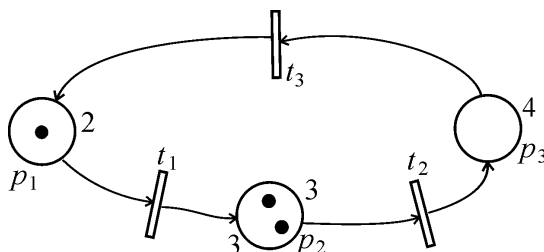


Figure 3.4 –

In both cases the average departure interval is 4 units of time, however, in the second case the departure interval is constant, i.e., the system has a so-called 1-periodic behavior, while in the first case the system shows a 2-periodic behavior. The recursive equations for the firing times of transitions  $t_1$  and  $t_2$  are

$$\begin{cases} x_1(k+1) = \max(x_1(k) + 2, x_2(k) + 5) \\ x_2(k+1) = \max(x_1(k) + 3, x_2(k) + 3) \end{cases}$$

### 3.2.2 Example 2

Consider now the example of Figure 3.4. We have the following equations

$$\begin{cases} x_1(k+1) = x_3(k) \\ x_2(k+1) = x_1(k-2) \\ x_3(k+1) = x_2(k+1) \end{cases}$$

which is not a state equation. From this example, we conclude that we have naturally a state equation of the form  $\mathbf{x}(k+1) = \mathbf{f}(\mathbf{x}(k))$  only if for the initial time, each place contains a single token (which is not the case here). Not also that due to the specific structure of the system (i.e. each transition has exactly one upstream and one downstream places) the max operator does not appear in the equations.

### 3.2.3 Example 3

Let us come back to Example 1. The bus company may consider to realize shorter (average) departure intervals. This could be achieved by using an additional bus. For example the company provides a second bus in the inner loop, e.g., initially on the way  $t_1$  to  $t_2$ . With respect to the timed event graph shown in Figure 3.3, this means to add a second token in place  $p_2$  as shown in Figure 3.5. We get

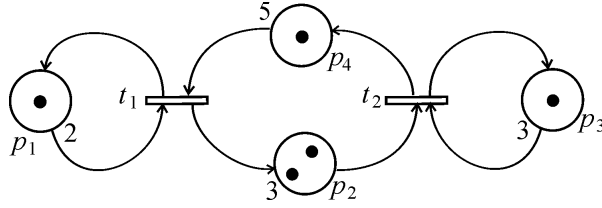


Figure 3.5 –

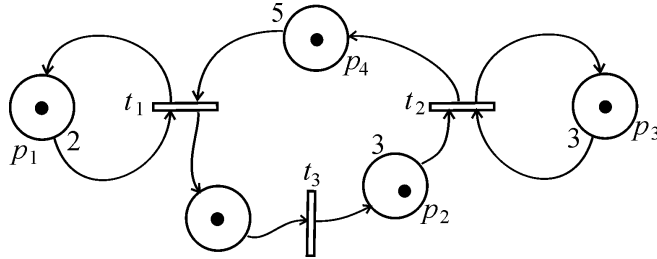


Figure 3.6 –

$$\begin{cases} x_1(k+1) = \max(x_1(k) + 2, x_2(k) + 5) \\ x_2(k+1) = \max(x_1(k-1) + 3, x_2(k) + 3), \end{cases}$$

which does not correspond to a state equation. The problem is due to the fact that  $x_2(k+1)$  depends on  $x_1(k-1)$  which make the recurrence relation of order 2. This problem appear as soon as we have more than one token at one place when  $k = 0$ . To get a state equation, we introduce a new variable  $x_3$  which correspond to a new transition  $t_3$  just between  $t_1$  and  $p_2$ , as shown in Figure 3.6 We get

$$\begin{cases} x_1(k+1) = \max(x_1(k) + 2, x_2(k) + 5) \\ x_2(k+1) = \max(x_3(k) + 3, x_2(k) + 3) \\ x_3(k+1) = x_1(k). \end{cases}$$

If the initial condition is  $x_1(1) = x_2(1) = x_3(1) = 0$ , the time table is

$$\begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 5 \\ 3 \\ 0 \end{pmatrix}, \begin{pmatrix} 8 \\ 6 \\ 5 \end{pmatrix}, \begin{pmatrix} 11 \\ 9 \\ 8 \end{pmatrix}, \begin{pmatrix} 14 \\ 12 \\ 11 \end{pmatrix}.$$

After a short transient phase, bus start from both stations in intervals of three units of time. Obviously, shorter intervals cannot be reached by additional buses in the inner loop of the system, as the outer loop at station 2 now represents the “bottleneck” of the system. In this simple example, several phenomena have been encountered. These phenomena (and more) can be conveniently analyzed and explained within the formal framework of idempotent semirings, or dioids.

### 3.2.4 Dioid

**Definition 5 (Monoid)**  $(\mathcal{M}, \oplus, \varepsilon)$  is a monoid if  $\oplus$  is a closed law, associative, and having a neutral element denoted  $\varepsilon$  ( $\forall a \in \mathcal{M}, a \oplus \varepsilon = \varepsilon \oplus a = a$ ). If law  $\oplus$  is commutative, the monoid is said to be

commutative. The monoid is idempotent if  $\forall a \in \mathcal{M}, a \oplus a = a$ .

For instance,  $(\mathbb{R}, \max)$  is an idempotent commutative monoid.

**Definition 6 (Semiring, dioid)**  $(\mathcal{D}, \oplus, \otimes)$  is an idempotent semiring, also called dioid, if

- $(\mathcal{D}, \oplus, \varepsilon)$  is an idempotent commutative monoid,  $\forall a \in \mathcal{D}, a \oplus a = a$ ,
- $(\mathcal{D}, \otimes, e)$  is a monoid,
- law  $\otimes$  distributes over law  $\oplus$ ,  $(a \otimes (b \oplus c) = (a \otimes b \oplus a \otimes c))$
- $\varepsilon$  is absorbing for law  $\otimes$ ,  $\forall a \in \mathcal{D}, a \otimes \varepsilon = \varepsilon \otimes a = \varepsilon$ .

If  $(\mathcal{D}, \otimes, e)$  is a commutative monoid, the idempotent semiring  $(\mathcal{D}, \oplus, \otimes)$  is said to be commutative.

For instance,  $(\overline{\mathbb{R}}, \max, +)$ , where  $\overline{\mathbb{R}} = \mathbb{R} \cup \{-\infty, +\infty\}$  is a dioid with  $\varepsilon = -\infty$ ,  $e = 0$ .

**Theorem.** In a dioid  $(\mathcal{D}, \oplus, \otimes)$ , a solution of the implicit equation

$$x = a \otimes x \oplus b$$

is  $x = a^* \otimes b$  where

$$a^* = \bigoplus_{k \geq 0} a^k = e \oplus a \oplus a^2 \oplus a^3 \oplus \dots$$

**Proof.** If  $x = a^* \otimes b$ , we have

$$\begin{aligned} a \otimes x \oplus b &= \underbrace{a \otimes a^*}_{= a \otimes (e \oplus a \oplus a^2 \oplus \dots)} \otimes b \oplus \underbrace{b}_{= e \otimes b} \\ &= a \otimes (e \oplus a \oplus a^2 \oplus \dots) \\ &= (a \oplus a^2 \oplus a^3 \oplus \dots) \\ &= (e \oplus a \oplus a^2 \oplus a^3 \oplus \dots) \otimes b \\ &= a^* \otimes b = x. \end{aligned}$$

**Example.** Consider the equation

$$\begin{aligned} x &= -2 \otimes x \oplus 3 \\ &= \max(x - 2, 3) \end{aligned}$$

in the dioid  $(\overline{\mathbb{R}}, \max, +) = (\overline{\mathbb{R}}, \oplus, \otimes)$ . One solution is  $a^* \otimes b$  with  $a = -2$  and  $b = 3$ . Thus

$$a^* \otimes b = \underbrace{(e \oplus a \oplus a^2 \oplus a^3 \oplus \dots)}_{= \max(0, -2, -4, \dots)} \otimes b = 0 \otimes 3 = 3.$$

Note that a simple fixed point method could also be used to solve the equation. It suffices to iterate the sequence

$$x_{k+1} = \max(x_k - 2, 3).$$

The convergence speed depends on the initial condition.

### 3.2.5 Matrices in dioids

We shall now consider matrices in  $(\overline{\mathbb{R}}, \max, +)$ . The matrix sum is defined componentwise. For instance

$$\begin{pmatrix} 2 & 5 \\ 3 & 7 \end{pmatrix} \oplus \begin{pmatrix} e & 8 \\ 1 & 3 \end{pmatrix} = \begin{pmatrix} 2 & 8 \\ 3 & 7 \end{pmatrix}.$$

If  $\mathbf{A} \in \mathbb{R}^{m \times p}$ ,  $\mathbf{B} \in \mathbb{R}^{p \times n}$ , the product  $\mathbf{C} = \mathbf{A} \otimes \mathbf{B} \in \mathbb{R}^{m \times n}$  is the matrix with entries

$$c_{ij} = \bigoplus_{k=1}^p a_{ik} \otimes a_{kj}.$$

The null matrix, denoted by  $\varepsilon$ , is the matrix whose entries are equal to  $\varepsilon$ . In the same manner the identity matrix, denoted by  $\mathbf{I}$ , is the matrix whose entries are all equal to  $\varepsilon$  excepted the diagonal entries which are equal to  $e$ . For example, if

$$\begin{pmatrix} 2 & 5 \\ \varepsilon & 3 \\ 1 & 8 \end{pmatrix} \otimes \begin{pmatrix} e \\ 1 \end{pmatrix} = \begin{pmatrix} 6 \\ 4 \\ 9 \end{pmatrix}.$$

By extension for  $n \in \mathbb{N}$ ,

$$\mathbf{A}^n = \underbrace{\mathbf{A} \otimes \mathbf{A} \otimes \dots \otimes \mathbf{A}}_{n \text{ times}}$$

with  $\mathbf{A}^0 = \mathbf{I}$  the identity matrix. It can easily shown that the set of matrices equipped with operations  $\oplus, \otimes$ , is a dioid. Moreover, the vector  $\mathbf{x} = \mathbf{A}^* \mathbf{b}$  is a solution of the implicit equation

$$\mathbf{x} = \mathbf{A} \otimes \mathbf{x} \oplus \mathbf{b}. \quad (3.1)$$

## 3.3 Timed event graphs

Timed event graphs (TEG) constitute a subclass of timed Petri nets. Each place admits one and only one upstream transition and one and only one downstream transition. These dynamical systems can be represented by the following linear state equations

$$\begin{aligned} \mathbf{x}(k) &= \mathbf{A}\mathbf{x}(k-1) \oplus \mathbf{B}\mathbf{u}(k) \\ \mathbf{y}(k) &= \mathbf{C}\mathbf{x}(k), \end{aligned}$$

where the algebraic structure is changed into a  $(\max, +)$  or a  $(\min, +)$  algebra. The vector of input transitions is  $\mathbf{u}$ , the vector of internal transitions is  $\mathbf{x}$  and  $\mathbf{y}$  is the vector of output transitions. To each place is associated a delay which characterizes the minimal time that a token has to stay in a place before to contribute to the firing of the downstream transition. A transition is fired when each upstream place has a valid token, i.e. a token having spent the minimal time specified by the temporization. Two dual approaches exist to model a TEG by state equations : the dater approach and the counter approach.



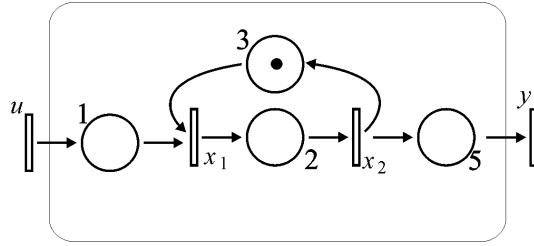


Figure 3.7 – A simple TEG

### 3.3.1 Dater state equations

To the  $i$ th transition we associate the date  $x_i(k) \in \mathbb{R}$  of the occurrence of  $k$ th firing of the transition. If we denote by  $\oplus$  the max operator and by  $\otimes$  the operator  $+$ . It is trivial to describe the dynamic of a given TEG by recurrence equations of the form

$$\begin{aligned} \mathbf{x}(k) &= \mathbf{A}_0 \mathbf{x}(k) \oplus \mathbf{A}_1 \mathbf{x}(k-1) \oplus \mathbf{B} \mathbf{u}(k), \\ \mathbf{y}(k) &= \mathbf{C} \mathbf{x}(k). \end{aligned} \quad (3.2)$$

To get this form, it is sometimes necessary to enlarge the graph in order to guarantee that each place is initially with at the most one token. The evolution equation has an implicit form :

$$\mathbf{x}(k) = \mathbf{A}_0 \mathbf{x}(k) \oplus \underbrace{\mathbf{A}_1 \mathbf{x}(k-1) \oplus \mathbf{B} \mathbf{u}(k)}_{\mathbf{b}}.$$

Since the system is deterministic, it has a unique solution. From Equation (3.1), the solution of

$$\mathbf{x} = \mathbf{A}_0 \mathbf{x} \oplus \mathbf{b}$$

is

$$\mathbf{x} = \mathbf{A}_0^* \mathbf{b}.$$

Thus

$$\begin{aligned} \mathbf{x}(k) &= \mathbf{A}_0^* (\mathbf{A}_1 \mathbf{x}(k-1) \oplus \mathbf{B} \mathbf{u}(k)) \\ &= \mathbf{A}_0^* \mathbf{A}_1 \mathbf{x}(k-1) \oplus \mathbf{A}_0^* \mathbf{B} \mathbf{u}(k), \end{aligned}$$

or equivalently

$$\begin{aligned} \mathbf{x}(k) &= \mathbf{A} \mathbf{x}(k-1) \oplus \mathbf{B} \mathbf{u}(k) \\ \mathbf{y}(k) &= \mathbf{C} \mathbf{x}(k) \end{aligned}$$

with  $\mathbf{A} = \mathbf{A}_0^* \mathbf{A}_1$  and  $\mathbf{B} = \mathbf{A}_0^* \mathbf{B}_0$ . The (max,+) toolbox of SCILAB, is very efficient to handle this kind of model.

**Example 1.** Consider the TEG of Figure 3.7. To the  $i$ th transition we associate the date  $x_i(k) \in \mathbb{R}$  of the occurrence of  $k$ th firing of the transition. We have

$$\begin{aligned} x_1(k) &= \max(1 + u(k), 3 + x_2(k-1)) \\ x_2(k) &= 2 + x_1(k) \\ y(k) &= x_2(k) + 5 \end{aligned}$$

or equivalently

$$\begin{aligned} x_1(k) &= 1 \otimes u(k) \oplus 3 \otimes x_2(k-1) \\ x_2(k) &= 2 \otimes x_1(k) \\ y(k) &= 5 \otimes x_2(k) \end{aligned}$$

which is linear in the idempotent semiring  $(\mathbb{R}, \max, +)$ . In a matrix form, we get

$$\begin{cases} \mathbf{x}(k) = \underbrace{\begin{pmatrix} \varepsilon & \varepsilon \\ 2 & \varepsilon \end{pmatrix}}_{\mathbf{A}_0} \mathbf{x}(k) \oplus \underbrace{\begin{pmatrix} \varepsilon & 3 \\ \varepsilon & \varepsilon \end{pmatrix}}_{\mathbf{A}_1} \mathbf{x}(k-1) \oplus \underbrace{\begin{pmatrix} 1 \\ \varepsilon \end{pmatrix}}_{\mathbf{B}} \mathbf{u}(k) \\ y(k) = \begin{pmatrix} \varepsilon & 5 \end{pmatrix} \mathbf{x}(k). \end{cases}$$

The evolution equations are

$$\mathbf{x}(k) = \mathbf{A}_0^* \mathbf{A}_1 \mathbf{x}(k-1) \oplus \mathbf{A}_0^* \mathbf{B} u(k).$$

Now,

$$\mathbf{A}_0^* = \begin{pmatrix} \varepsilon & \varepsilon \\ 2 & \varepsilon \end{pmatrix}^* = \begin{pmatrix} e & \varepsilon \\ \varepsilon & e \end{pmatrix} \oplus \begin{pmatrix} \varepsilon & \varepsilon \\ 2 & \varepsilon \end{pmatrix} \oplus \underbrace{\begin{pmatrix} \varepsilon & \varepsilon \\ 2 & \varepsilon \end{pmatrix}^2}_{\begin{pmatrix} \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \end{pmatrix}} \oplus \underbrace{\begin{pmatrix} \varepsilon & \varepsilon \\ 2 & \varepsilon \end{pmatrix}^3}_{\begin{pmatrix} \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \end{pmatrix}} + \dots = \begin{pmatrix} e & \varepsilon \\ 2 & e \end{pmatrix}.$$

and

$$\mathbf{A}_0^* \mathbf{A}_1 = \begin{pmatrix} e & \varepsilon \\ 2 & e \end{pmatrix} \begin{pmatrix} \varepsilon & 3 \\ \varepsilon & \varepsilon \end{pmatrix} = \begin{pmatrix} \varepsilon & 3 \\ \varepsilon & 5 \end{pmatrix}.$$

Moreover

$$\mathbf{A}_0^* \mathbf{B} = \begin{pmatrix} 0 & \varepsilon \\ 2 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 3 \end{pmatrix}.$$

The state equations of our TEG are

$$\begin{aligned} \mathbf{x}(k) &= \begin{pmatrix} \varepsilon & 3 \\ \varepsilon & 5 \end{pmatrix} \mathbf{x}(k-1) \oplus \begin{pmatrix} 1 \\ 3 \end{pmatrix} \mathbf{u}(k) \\ \mathbf{y}(k) &= \begin{pmatrix} \varepsilon & 5 \end{pmatrix} \mathbf{x}(k). \end{aligned}$$

Note that for this simple example, the same result could have been obtained directly from the initial equations. Since

$$\begin{aligned} x_1(k) &= \max(1 + u(k), 3 + x_2(k-1)) \\ x_2(k) &= 2 + x_1(k) \\ y(k) &= x_2(k) + 5 \end{aligned}$$

we have

$$\begin{aligned} x_1(k) &= \max(1 + u(k), 3 + x_2(k-1)) \\ x_2(k) &= 2 + \max(1 + u(k), 3 + x_2(k-1)) = \max(3 + u(k), 5 + x_2(k-1)) \\ y(k) &= x_2(k) + 5. \end{aligned}$$

It corresponds to what has been obtained with max-plus tools.

**Example 2.** Consider the TEG of Figure 3.8.

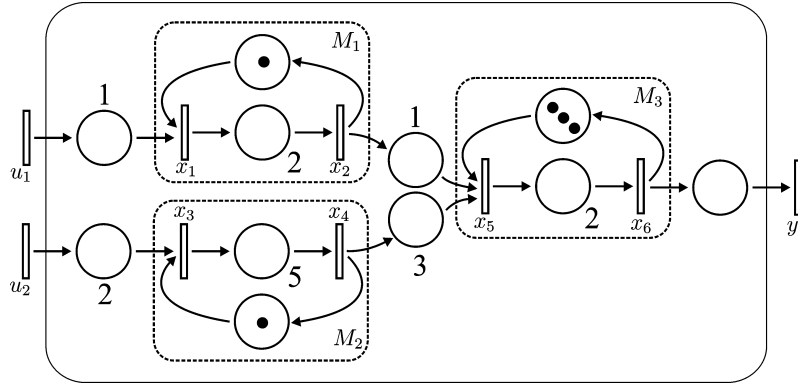


Figure 3.8 – Another timed event graph

The input transitions are  $u_1, u_2$ , the internal transitions are  $x_i, i \in \{1, \dots, 6\}$  and  $y$  is the output transition. This TEG can represent the behavior of an assembly line, constituted of 3 machines  $M_1, M_2$  and  $M_3$ . Transition  $u_1$  characterizes the input of raw materials in the system, transition  $x_1$  represents the input of material in machine  $M_1$ , it is possible if a token is available in the place located between transitions  $x_2$  and  $x_1$  (*i.e.* the machine has to be available), and transition  $u_1$  has to be fired for one time unit. After 2 time units transition  $x_2$  will be fired (output of machine  $M_1$ ). Transition  $x_5$  represents the input of machine  $M_3$  which ensures the assembly of products coming from machines  $M_1$  and  $M_2$ . For our TEG, we can write

$$\begin{aligned}
 x_1(k) &= \max(1 + u_1(k), x_2(k - 1)) \\
 x_2(k) &= 2 + x_1(k) \\
 x_3(k) &= \max(2 + u_2(k), x_4(k - 1)) \\
 x_4(k) &= 5 + x_3(k) \\
 x_5(k) &= \max(3 + x_4(k), 1 + x_2(k), x_6(k - 3)) \\
 x_6(k) &= 2 + x_5(k) \\
 y(k) &= x_6(k)
 \end{aligned}$$

or equivalently

$$\begin{aligned}
 x_1(k) &= 1 \otimes u_1(k) \oplus x_2(k - 1) \\
 x_2(k) &= 2 \otimes x_1(k) \\
 x_3(k) &= 2 \otimes u_2(k) \oplus x_4(k - 1) \\
 x_4(k) &= 5 \otimes x_3(k) \\
 x_5(k) &= 3 \otimes x_4(k) \oplus 1 \otimes x_2(k) \oplus x_6(k - 3) \\
 x_6(k) &= 2 \otimes x_5(k) \\
 y(k) &= x_6(k)
 \end{aligned}$$



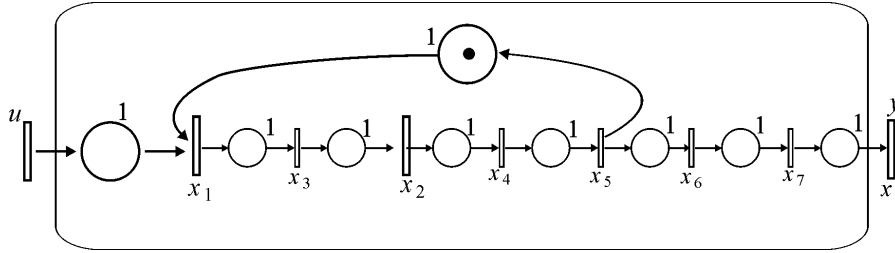


Figure 3.10 – Enlarged TEG associated with Example 1

For our example this calculus leads to

$$\mathbf{A} = \mathbf{A}_0^* \mathbf{A}_1 = \begin{pmatrix} \varepsilon & e & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & 2 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & e & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & 5 & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & 3 & \varepsilon & 8 & \varepsilon & \varepsilon & \varepsilon & e \\ \varepsilon & 5 & \varepsilon & 10 & \varepsilon & \varepsilon & \varepsilon & 2 \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & e & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & e & \varepsilon \end{pmatrix} \quad \text{and} \quad \mathbf{B} = \mathbf{A}_0^* \mathbf{B} = \begin{pmatrix} 1 & \varepsilon \\ 3 & \varepsilon \\ \varepsilon & 2 \\ \varepsilon & 7 \\ 4 & 10 \\ 6 & 12 \\ \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \end{pmatrix}.$$

### 3.3.2 Counter state equations

From a dual point of view, the behavior of a TEG can be described by considering a dynamic system in the time domain. A counter function  $x_i$  is associated to each transition which counts the number of firing of the transition at a time  $k \in \mathbb{Z}$ .

**Example 1.** Consider again the TEG of Example 1 page 35. The system can be represented by the following equations :

$$\begin{aligned} x_1(k) &= \min(u(k-1), 1 + x_2(k-3)) \\ x_2(k) &= x_1(k-2) \\ y(k) &= x_2(k-5) \end{aligned}$$

An equivalent enlarged TEG is given by Figure 3.10 where all delays are equal to 1. The corresponding state equations are

$$\begin{aligned} x_1(k) &= \min(u(k-1), 1 + x_5(k-1)) \\ x_2(k) &= x_3(k-1) \\ x_3(k) &= x_1(k-1) \\ x_4(k) &= x_2(k-1) \\ x_5(k) &= x_4(k-1) \\ x_6(k) &= x_5(k-1) \\ x_7(k) &= x_6(k-1) \\ x_8(k) &= x_7(k-1) \\ y(k) &= x_8(k) \end{aligned}$$

or equivalently

$$\begin{aligned}
 \mathbf{x}(k) &= \begin{pmatrix} \varepsilon & \varepsilon & \varepsilon & \varepsilon & 1 & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & e & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ e & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & e & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & e & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & e & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & e & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & e & \varepsilon \end{pmatrix} \mathbf{x}(k-1) \oplus \begin{pmatrix} e \\ \varepsilon \\ \varepsilon \\ \varepsilon \\ \varepsilon \\ \varepsilon \\ \varepsilon \\ \varepsilon \end{pmatrix} \mathbf{u}(k-1) \\
 y(k) &= \begin{pmatrix} \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & e \end{pmatrix} \mathbf{x}(k)
 \end{aligned}$$

where the operations have to be understood with respect to the  $(\min, +)$ -algebra.

**Example 2.** Consider now the TEG of Example 2 page 36 and depicted on Figure 3.8. The system can be represented by the following equations :

$$\begin{aligned}
 x_1(k) &= \min(u_1(k-1), 1 + x_2(k)) \\
 x_2(k) &= x_1(k-2) \\
 x_3(k) &= \min(u_2(k-2), 1 + x_4(k)) \\
 x_4(k) &= x_3(k-5) \\
 x_5(k) &= \min(x_4(k-3), x_2(k-1), 3 + x_6(k)) \\
 x_6(k) &= x_5(k-2) \\
 y(k) &= x_6(k)
 \end{aligned}$$

In the dioid  $(\mathbb{Z}, \min, +)$ , these equations are expressed by :

$$\begin{aligned}
 x_1(k) &= u_1(k-1) \oplus 1 \otimes x_2(k) \\
 x_2(k) &= x_1(k-2) \\
 x_3(k) &= u_2(k-2) \oplus 1 \otimes x_4(k) \\
 x_4(k) &= x_3(k-5) \\
 x_5(k) &= x_4(k-3) \oplus x_2(k-1) \oplus 3 \otimes x_6(k) \\
 x_6(k) &= x_5(k-2) \\
 y(k) &= x_6(k).
 \end{aligned}$$

These dynamic equations are linear in  $(\mathbb{Z}, \min, +)$ . In a vector form, we have

$$\left\{ \begin{array}{l} \mathbf{x}(k) = \begin{pmatrix} \varepsilon & 1 & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & 1 & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \mathbf{3} \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \end{pmatrix} \mathbf{x}(k) \oplus \begin{pmatrix} \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & e & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \end{pmatrix} \mathbf{x}(k-1) \oplus \begin{pmatrix} \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ e & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & e \end{pmatrix} \mathbf{x}(k-2) \\ \\ \oplus \begin{pmatrix} \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & e & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \end{pmatrix} \mathbf{x}(k-3) \oplus \begin{pmatrix} \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & e & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \end{pmatrix} \mathbf{x}(k-5) \\ \\ \oplus \begin{pmatrix} e & \varepsilon \\ \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \end{pmatrix} \mathbf{u}(k-1) \oplus \begin{pmatrix} \varepsilon & \varepsilon \\ \varepsilon & e \\ \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \\ \varepsilon & \varepsilon \end{pmatrix} \mathbf{u}(k-2) \\ \\ y(k) = \begin{pmatrix} \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & e \end{pmatrix} \mathbf{x}(k) \end{array} \right.$$

In a general manner a TEG can be represented in  $(\mathbb{Z}, \min, +)$  by the following equations :

$$\left\{ \begin{array}{l} \mathbf{x}(k) = \bigoplus_{i=0}^{i_{\max}} \mathbf{A}_i \mathbf{x}(k-i) \oplus \bigoplus_{j=0}^{j_{\max}} \mathbf{B}_j \mathbf{u}(k-j) \\ \mathbf{y}(k) = \bigoplus_{\ell=0}^{\ell_{\max}} \mathbf{C}_\ell \mathbf{x}(k-\ell). \end{array} \right.$$

After some extension of the state, it is possible to get a recursive formulation with a delay of one time unit, it consists of increasing the state in order to have only temporization of one time unit on each place. Figure 3.11 yields the corresponding extension of the TEG of Figure 3.8. As a consequence, our TEG can be written under the form

$$\left\{ \begin{array}{l} \mathbf{x}(k) = \mathbf{A}_0 \mathbf{x}(k) \oplus \mathbf{A}_1 \mathbf{x}(k-1) \oplus \mathbf{B}_0 \mathbf{u}(k) \\ \mathbf{y}(k) = \mathbf{C}_0 \mathbf{x}(k). \end{array} \right.$$

From Equation (3.1), it is then possible to obtain the following explicit formulation :

$$\left\{ \begin{array}{l} \mathbf{x}(k) = \mathbf{A} \mathbf{x}(k-1) \oplus \mathbf{B} \mathbf{u}(k) \\ y(k) = \mathbf{C} \mathbf{x}(k), \end{array} \right.$$

with  $\mathbf{A} = \mathbf{A}_0^* \mathbf{A}_1$  and  $\mathbf{B} = \mathbf{A}_0^* \mathbf{B}_0$ .

### 3.3.3 Gamma-delta representation

Figure 3.12 represents discret event signal in the  $\gamma$ - $\delta$  plane.

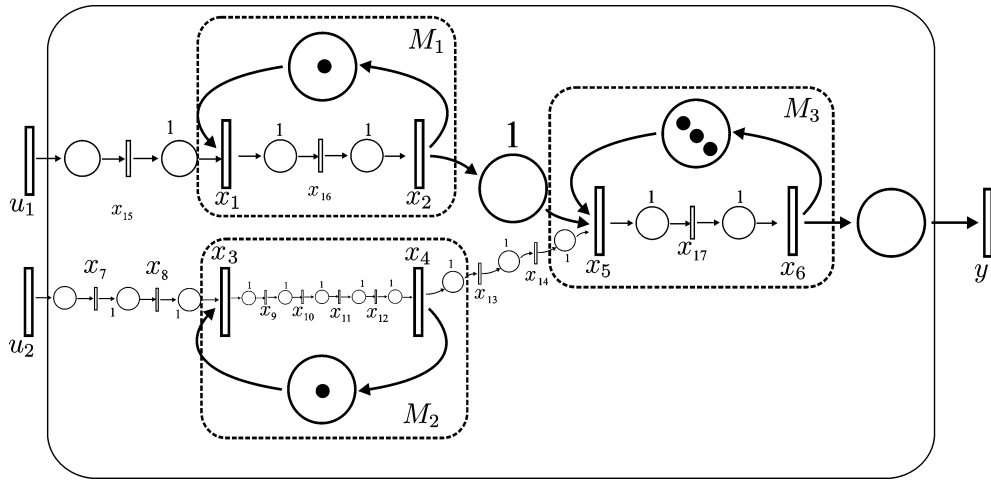


Figure 3.11 – Extended TEG to get one unit temporization at each place

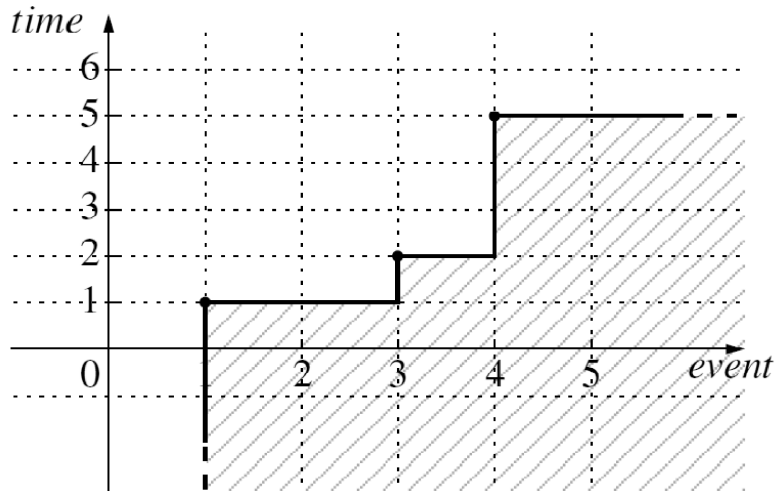


Figure 3.12 –



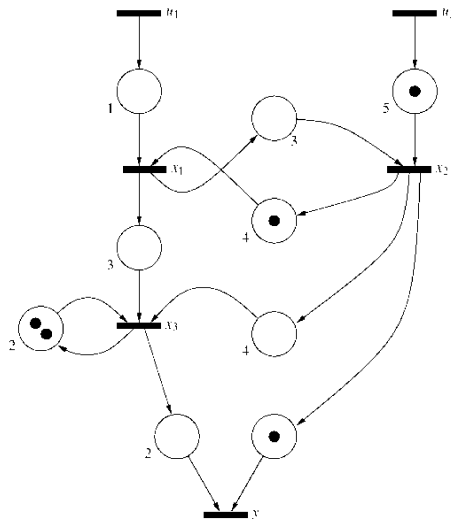


Figure 3.13 – Example of TEG

This signal can be represented by the series

$$\gamma^1 \delta^1 \oplus \gamma^3 \delta^2 \oplus \gamma^4 \delta^5.$$

The following interpretation of the monomial  $\gamma^k \delta^t$  is : the  $k + 1$  occurrence of the event happens at time  $t$ . It corresponds to the *south-east cone* of the  $\gamma$ - $\delta$  plane with vertex  $(k, t)$ . The operation  $\oplus$  corresponds to the union between south-east cones. The multiplication is defined by

$$\gamma^{k_1} \delta^{t_1} \otimes \gamma^{k_2} \delta^{t_2} = \gamma^{k_1+k_2} \delta^{t_1+t_2}.$$

We can show that the set of series is a dioid. The zero, unit and top elements are

$$\begin{aligned} \varepsilon &= \gamma^{+\infty} \cdot \delta^{+\infty} \\ e &= \gamma^0 \cdot \delta^0 \\ \top &= \gamma^{-\infty} \cdot \delta^{+\infty}. \end{aligned}$$

**Simplification rules**

$$\begin{aligned} \gamma^k \delta^t \oplus \gamma^k \delta^\tau &= \gamma^k \cdot \delta^{\max(t,\tau)} \\ \gamma^k \delta^t \oplus \gamma^\ell \delta^t &= \gamma^{\min(k,\ell)} \cdot \delta^t \end{aligned}$$

**Application to TEG**

Consider the TEG represented in Figure 3.13.

We have

$$\begin{aligned} x_1 &= \gamma^1 \delta^4 x_2 \oplus \gamma^0 \delta^1 u_1 \\ x_2 &= \gamma^0 \delta^3 x_1 \oplus \gamma^1 \delta^5 u_2 \\ x_3 &= \gamma^0 \delta^3 x_1 \oplus \gamma^0 \delta^4 x_2 \oplus \gamma^2 \delta^2 x_3 \\ y &= \gamma^1 \delta^0 x_2 \oplus \gamma^0 \delta^2 x_3 \end{aligned}$$

or in a matrix form

$$\begin{aligned} \mathbf{x} &= \begin{pmatrix} \varepsilon & \gamma^1 \delta^4 & \varepsilon \\ \gamma^0 \delta^3 & \varepsilon & \varepsilon \\ \gamma^0 \delta^3 & \gamma^0 \delta^4 & \gamma^2 \delta^2 \end{pmatrix} \mathbf{x} \oplus \begin{pmatrix} \gamma^0 \delta^1 & \varepsilon \\ \varepsilon & \gamma^1 \delta^5 \\ \varepsilon & \varepsilon \end{pmatrix} \mathbf{u} \\ y &= \begin{pmatrix} \varepsilon & \gamma^1 \delta^0 & \gamma^0 \delta^2 \end{pmatrix}. \end{aligned}$$