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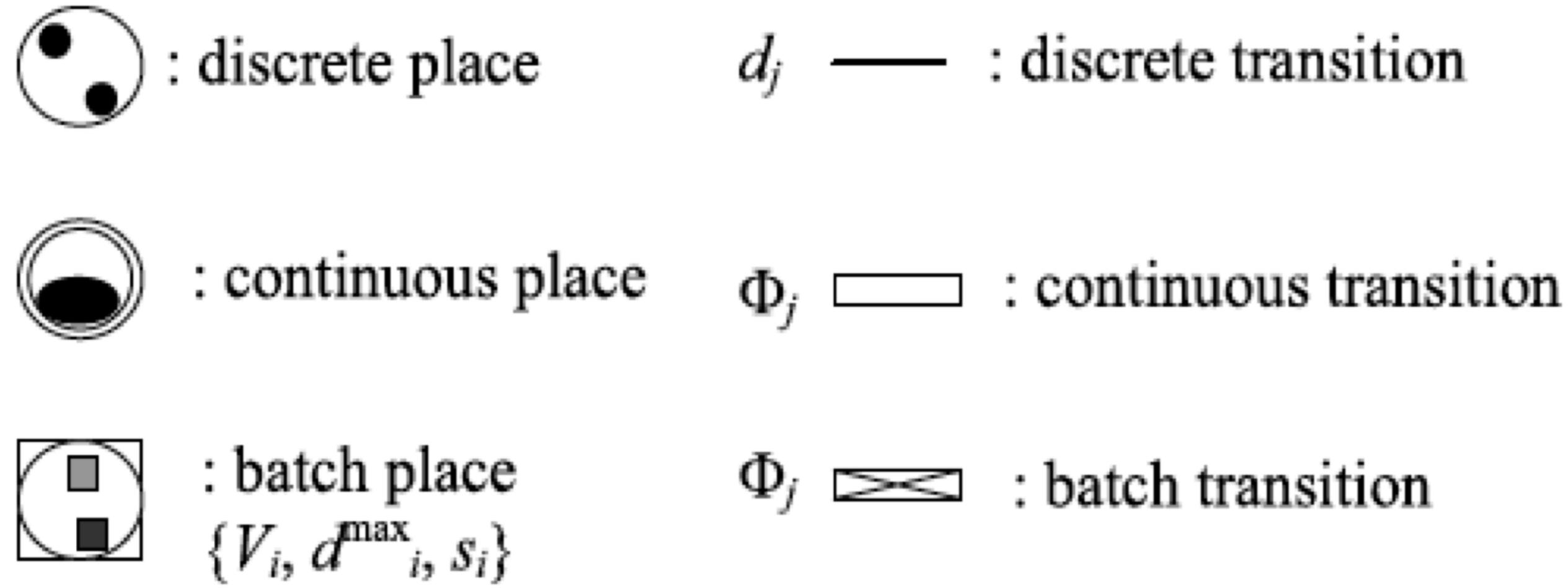
Algorithm for Controlling the Transient Behavior of Controlled Generalized Batches Petri Nets

· Ruotian Liu, Rabah Ammour, Leonardo Brenner, Isabel Demongodin

Objective

Compute a **control trajectory** to reach a target **steady state** from a given **initial state** in hybrid / discrete event systems.

Controlled GBPN



- A **batch** β_k at time τ is: $\beta_k(\tau) = (l_k(\tau), d_k(\tau), x_k(\tau))$, where $l_k(\tau) \in \mathbb{R}_{\geq 0}$ is the length, $d_k(\tau) \in \mathbb{R}_{\geq 0}$ is the density and $x_k(\tau) \in \mathbb{R}_{\geq 0}$ is the head position.
- The **marking** of a batch place p_i is a series of batches: $m_i(\tau) = \{\beta_1(\tau), \dots, \beta_r(\tau)\}$.
- The **marking quantity** vector is: $\mathbf{q} = \mu(\mathbf{m})$. For place p_i :

$$q_i(\tau) = \begin{cases} m_i(\tau) & \text{if } p_i \in P^C \\ \sum_{k=1}^r l_k(\tau) \cdot d_k(\tau) & \text{if } p_i \in P^B \end{cases}$$

- The **instantaneous firing flow**: $\varphi_j \leq \Phi_j$ represents the firing quantity of transition t_j by time unit.
- The **controlled firing flow** vector at time τ : $\mathbf{u}(\tau)$ is the control input that drives the evolution. $0 \leq u_j(\tau) \leq \Phi_j$

Steady state

The cGBPN system $\langle N, \mathbf{m}_0 \rangle$ is in a **steady state** at time τ_s if for $\tau \geq \tau_s$ the marking \mathbf{m}^s and the instantaneous firing flow vector φ^s remain constant. Thus a steady state is defined as a pair $(\mathbf{m}^s, \varphi^s)$.

Control of the transient behavior

Given a cGBPN system $\langle N, \mathbf{m}_0 \rangle$, a **control trajectory** is given as $(\mathbf{u}^0, \tau_0), (\mathbf{u}^1, \tau_1), \dots, (\mathbf{u}^i, \tau_i), \dots, (\mathbf{u}^n, \tau_n)$ such that the controlled firing flow vector \mathbf{u}^i is applied at date τ_i until τ_{i+1} .

Control strategy for reaching the steady state

- OFF: $u_j(\tau) = 0$ if t_j is not enabled or the marking quantity of one of its input places p_i is lower than its steady marking quantity $q_i(\tau) < q_i^s$.
- ON: maximize $\mathbf{u}(\tau)$ for approaching the steady marking quantity vector φ^s .

$$\mathbf{q}^s = \mathbf{q}^0 + C \cdot \left(\int_{\tau_0}^{\tau_1} \mathbf{u}^0(\rho) \cdot d\rho + \dots + \int_{\tau_{n-1}}^{\tau_n} \mathbf{u}^{n-1}(\rho) \cdot d\rho + \int_{\tau_n}^{+\infty} \mathbf{u}^n(\rho) \cdot d\rho \right)$$

with $C \cdot \int_{\tau_n}^{+\infty} \mathbf{u}^n(\rho) \cdot d\rho = 0$; $\mathbf{u}^n = \varphi^s$

Assumptions

- No discrete nodes ($P^D = T^D = \emptyset$) and no continuous transitions ($T^C = \emptyset$).
- The steady firing flow vector is positive ($\varphi^s > 0$).

Algorithm

Algorithm: Computation of control trajectory

Input: A cGBPN system $\langle N, \mathbf{m}_0 \rangle$ and the steady state $(\mathbf{m}^s, \varphi^s)$.

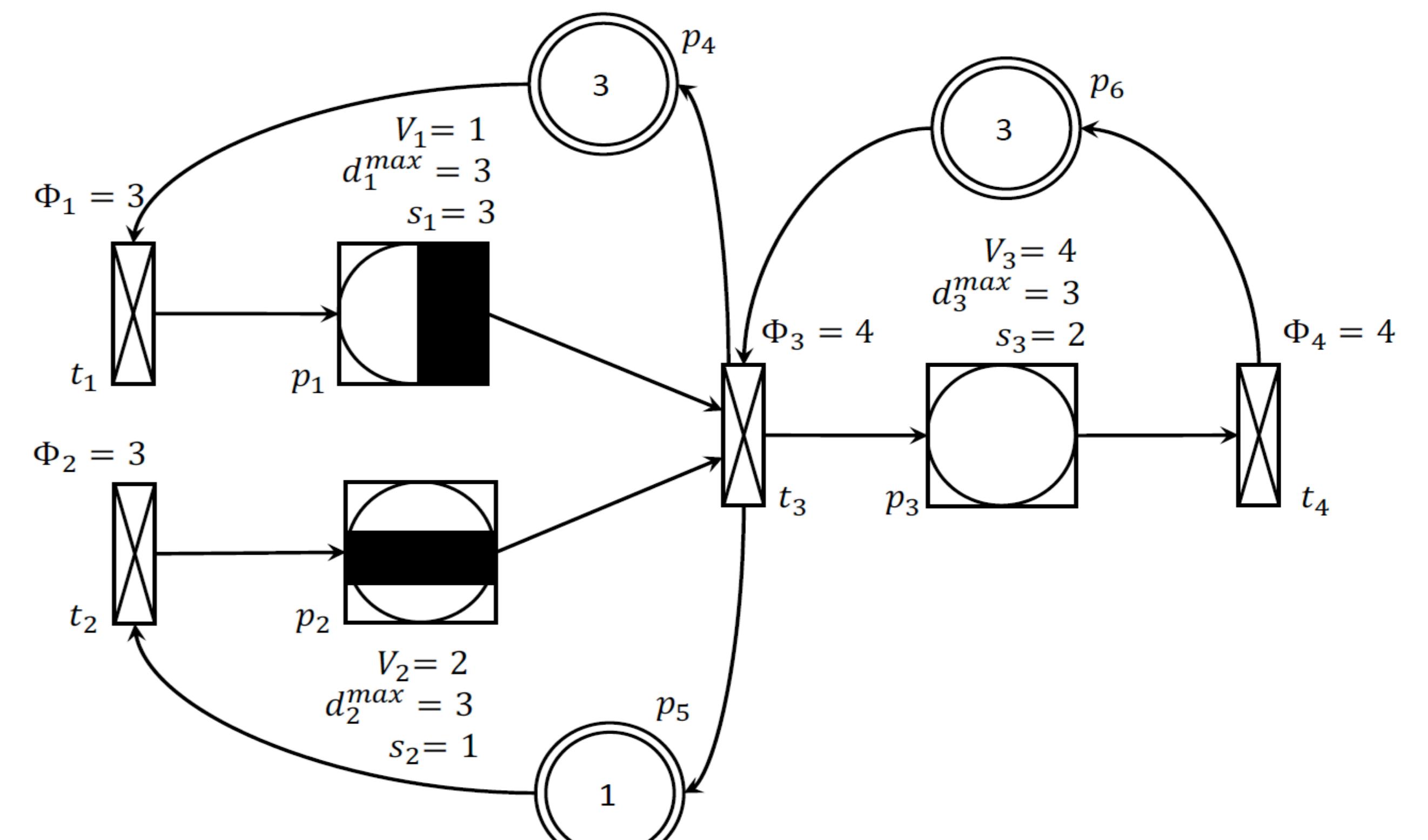
Output: Control trajectory $(\mathbf{u}^0, \tau_0), (\mathbf{u}^1, \tau_1), \dots$

- 1 Initialize : $q^0 = \mu(\mathbf{m}_0)$, $u^0 = 0$, $\tau_0 = 0$, $i = 0$;
- 2 **while** $\mathbf{m}^i \neq \mathbf{m}^s$ and $\mathbf{u}^i \neq \varphi^s$ **do**
- 3 Determine the sets $T_N(\mathbf{m}^i)$, $T_L(\mathbf{m}^i)$, $S_G(\mathbf{m}^i)$, $S_E(\mathbf{m}^i)$;
- 4 Solve the following LPP: $\max \mathbf{1}^T \cdot \mathbf{u}^i$ s.t.

$$\begin{cases} (a) 0 \leq u_j^i \leq \varphi_j^s & \forall t_j \in T \\ (b) u_j^i = 0 & \forall t_j \in T_N(\mathbf{m}) \cup T_L(\mathbf{m}) \\ (c) C(p_k, \cdot) \cdot \mathbf{u}^i \leq 0 & \forall p_k \in S_G(\mathbf{m}) \\ (d) C(p_k, \cdot) \cdot \mathbf{u}^i = 0 & \forall p_k \in S_E(\mathbf{m}) \\ (e) Pre(p_k, \cdot) \cdot \mathbf{u}^i \leq V_k \cdot d_k^{out} & \forall p_k \in P^B \end{cases}$$
- 5 Determine all the next timed events from considered ones, select the nearest at time τ_{i+1} ;
- 6 Determine the new marking \mathbf{m}^{i+1} and the marking quantity vector q^{i+1} ;
- 7 *i* = *i* + 1;
- 8 Return $(\mathbf{u}^0, \tau_0), (\mathbf{u}^1, \tau_1), \dots$

Example

$$\begin{aligned} \mathbf{m}_0 &= [\{(2, 3, 3)\} \{(1, 2, 1)\} \emptyset 3 1 3]^T \\ \mathbf{m}^s &= [\{(1, 2, 1), (2, 3, 3)\} \{(1, 1, 1)\} \{(2, 0.5, 2)\} 1 2 2]^T \\ \varphi^s &= [2 2 2 2]^T \end{aligned}$$



Control trajectory: $(\mathbf{u}^0, 0), (\mathbf{u}^1, 1), (\mathbf{u}^2, 1.5), (\mathbf{u}^3, 2)$ with:

$$\begin{array}{cccc} \begin{pmatrix} 2 \\ 0 \\ 0 \\ 0 \end{pmatrix} & \xrightarrow{\tau_1 = 1} & \begin{pmatrix} 2 \\ 0 \\ 2 \\ 0 \end{pmatrix} & \xrightarrow{\tau_2 = 1.5} \begin{pmatrix} 2 \\ 2 \\ 2 \\ 0 \end{pmatrix} \\ \mathbf{u}^0 & & \mathbf{u}^1 & \xrightarrow{\tau_3 = 2} \begin{pmatrix} 2 \\ 2 \\ 2 \\ 2 \end{pmatrix} \\ \tau_0 = 0 & & & \mathbf{u}^2 & \xrightarrow{\tau_4} \begin{pmatrix} 2 \\ 2 \\ 2 \\ 2 \end{pmatrix} = \varphi^s \end{array}$$

[1] I. Demongodin and A. Giua. Dynamics and steady state analysis of controlled generalized batches Petri nets. NAHS, 2014.

[2] L. Wang, C. Mahulea, J. Julvez and M. Silva. On/off strategy based minimum-time control of continuous Petri nets. NAHS, 2014.