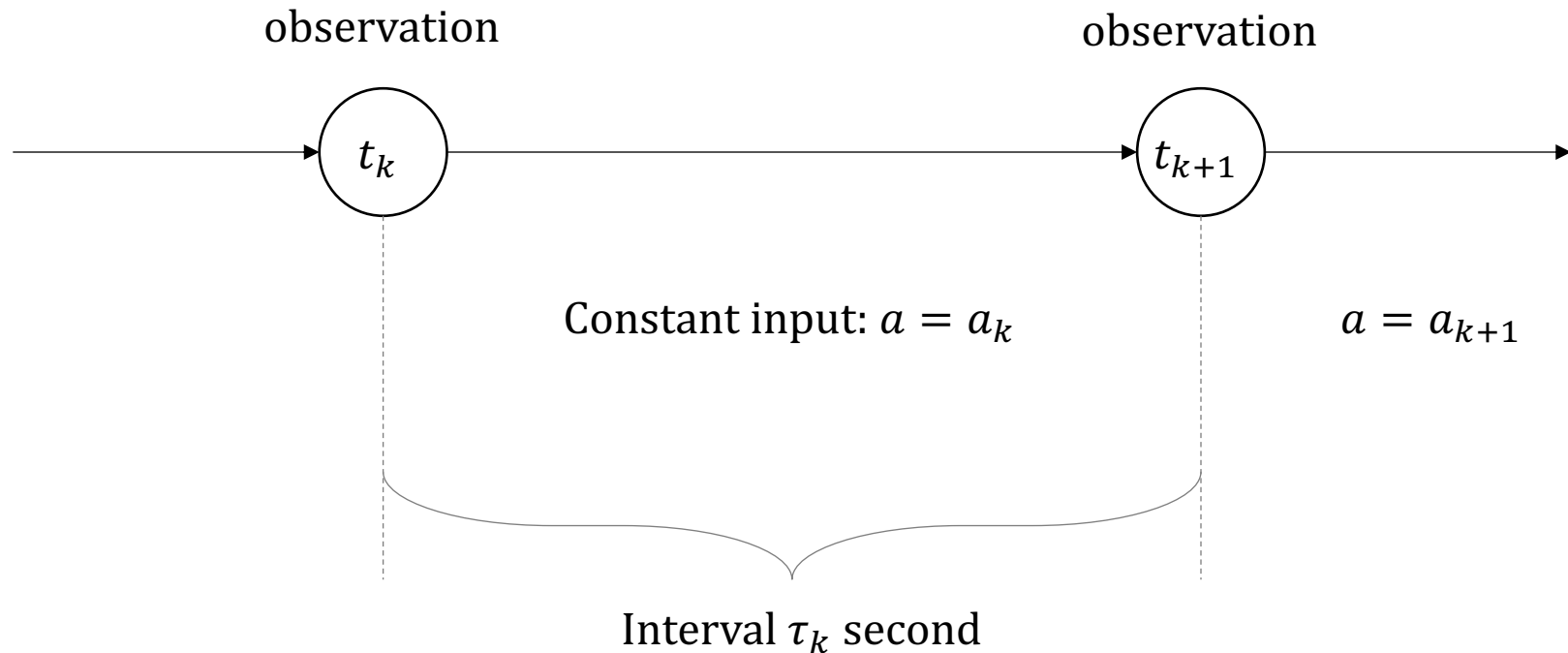


- Self-triggered control (cf. : Event-triggered control)
  - Controller decides action signal and next observation time
  - (State feedback) control law function:  $\pi(s) = [a(s) \quad \tau(s)]$



- Reinforcement learning for optimal self-triggered control  $\pi^*$

$$\pi^*(s) = \operatorname{argmax}_{\pi} J(\pi)$$

$$J(\pi) = \mathbb{E}_{s_0}[V^{\pi}(s_0)]$$

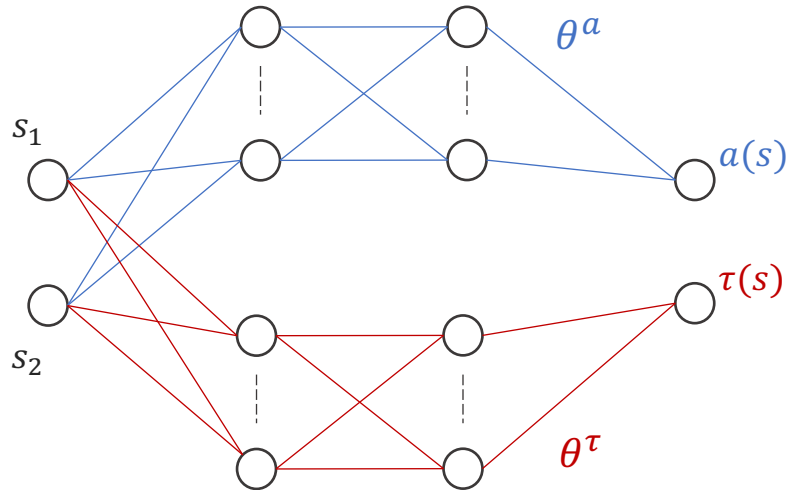
$$V^{\pi}(s) = \sum_{i=0}^{\infty} \gamma^i r_i^{\pi}$$

$$r_i^{\pi} = - \int_{t_k}^{t_{k+1}} s(t)^T Q s(t) dt - \tau_i a_i^T R a_i + \lambda \tau_i$$

- Assume every  $i$ -th step's access to data tuple  $(s_t, r_i^{\pi})$
- Assume the system to be control affine

$$\dot{s} = f(s) + g(s)a$$

- How to get  $\pi^*$  ?
  - Parametrize the function as  $\pi_{\theta^\pi}$  (express with NN)
  - Use gradient method to update NN, with policy gradient  $\nabla_{\theta^\pi} J(\theta^\pi)$   
 $\nabla_{\theta^\pi} J(\pi_{\theta^\pi})$
- Policy function has 2 elements :  $\pi(s) = [a(s) \quad \tau(s)]$ 
  - Divide policy network into 2 parametrized function



- $\theta^\pi$  expresses the combined network

$$\theta^\pi = [\theta^a \quad \theta^\tau]$$

- Theme for master thesis
  - Research the relation between  $\nabla_{\theta^a} J(\theta^\pi), \nabla_{\theta^\tau} J(\theta^\pi)$  limited to this problem (RL for self-triggered control)
- Analytical calculation
  - By the definition of  $J(\theta^\pi)$ , focus on  $\nabla_{\theta} V^{\theta^\pi}(s)$
  - $$\begin{aligned}\nabla_{\theta} V^{\theta^\pi}(s) &= \nabla_{\theta} [r(s, \pi(s|\theta^\pi)) + \gamma V^{\theta^\pi}(s'(\theta^\pi))] \\ &= \nabla_{\theta} r(s, \pi(s|\theta^\pi)) + \gamma \nabla_{\theta} \{V^{\theta^\pi}(s'(\theta^\pi))\}\end{aligned}$$
  - 1<sup>st</sup> element is gradient for step reward

$$\begin{aligned}\nabla_{\theta^a} r(s, \pi(s|[\theta^a, \theta^\tau])) &= \nabla_{\theta^a} a(s|\theta^a) \nabla_a r(s, [a \ \tau])|_{a=a(s|\theta^a), \tau=\tau(s|\theta^\tau)} \\ \nabla_{\theta^\tau} r(s, \pi(s|[\theta^a, \theta^\tau])) &= \nabla_{\theta^\tau} \tau(s|\theta^\tau) \nabla_\tau r(s, [a \ \tau])|_{a=a(s|\theta^a), \tau=\tau(s|\theta^\tau)}\end{aligned}$$
  - Calculate  $\nabla_a r(s, u)|_{u=\pi(s|\theta^\pi)}, \nabla_\tau r(s, u)|_{u=\pi(s|\theta^\pi)}$

- Analytical calculation
  - $\nabla_{\theta} V^{\pi_{\theta}}(s) = \nabla_{\theta} r(s, \pi(s|\theta^{\pi})) + \gamma \nabla_{\theta} \{V^{\theta^{\pi}}(s'(\theta^{\pi}))\}$
  - 2<sup>nd</sup> element is gradient for  $V^{\theta^{\pi}}$  at next state
  - When parameter  $\theta$  changes
    - $V^{\theta^{\pi}}(\cdot)$  changes
    - $s'(\theta^{\pi})$  changes
  - How to calculate this gradient ...?

$$\begin{aligned} \gamma \nabla_{\theta} \{f^{\theta}(g(\theta))\} &= \lim_{h \rightarrow 0} \frac{f^{\theta+h}(g(\theta+h)) - f^{\theta}(g(\theta+h))}{h} \\ &\quad + \lim_{h \rightarrow 0} \frac{f^{\theta}(g(\theta+h)) - f^{\theta}(g(\theta))}{h} \end{aligned}$$