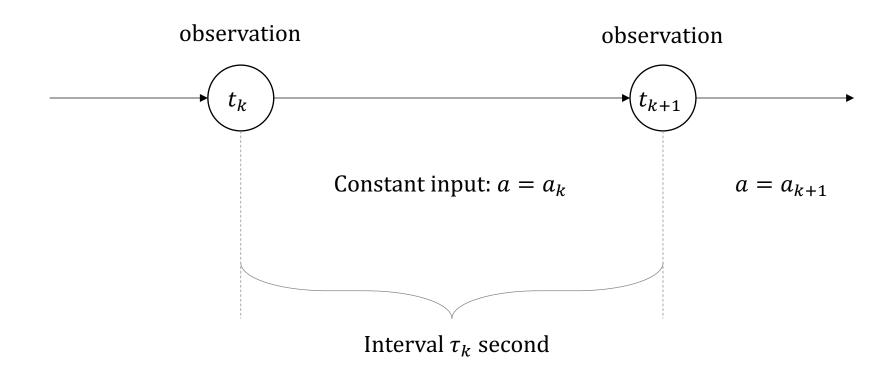
- Self-triggered control (cf.: Event-triggered control)
 - Controller decides action signal and next observation time
 - (State feedback) control low function: $\pi(s) = [a(s) \quad \tau(s)]$



• Reinforcement learning for optimal self-triggered control π^*

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

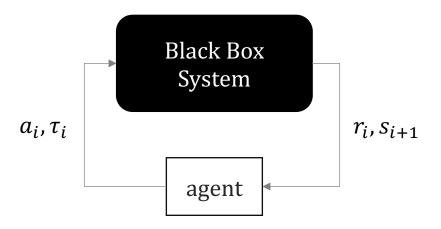
$$J(\pi) = \mathbb{E}_{s_0} [V^{\pi}(s_0)]$$

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

$$r_i^{\pi} = -\int_{T_i}^{T_{i+1}} s(t)^T Qs(t) dt - \tau_i a_i^T R a_i + \lambda \tau_i$$

• Assume every *i-th* step's access to data tuple (s_t, r_i^{π})

We can regard this task as a simple RL framework



- It will converge to optimal policy by long long iteration
- I want to discuss unique feature of this problem for master thesis
 - Ex.) step size ratio of gradient method for $a_{\mu}(s)$, $\tau_{\varphi}(s)$ $\nabla_{\mu} J(\pi), \nabla_{\varphi} J(\pi)$