

Meeting

Po Hsun Wu

October 18, 2022

Progress report

$$\bar{y} = 1 - \frac{y}{\max(y)} \quad (1)$$

	β	ϕ
Rise time(sec)	$\mu=0.5887$ $\sigma=0.3575$	$\mu=0.4132$ $\sigma=0.4636$
2% Settling time(sec)	$\mu=4.6819$ $\sigma=3.3599$	$\mu=4.3516$ $\sigma=3.3567$
5% Settling time(sec)	$\mu=2.4018$ $\sigma=2.8485$	$\mu=2.2058$ $\sigma=2.7957$
Overshoot(%)	$\mu= 9.0915$ $\sigma=21.5718$	$\mu= 8.9186$ $\sigma=21.3232$
Undershoot(%)	$\mu= 6.3101$ $\sigma=23.0001$	$\mu= 4.6244$ $\sigma=17.7231$

Stochastic policy

Using Stochastic policy to replace the controller. ($\pi : S \rightarrow \mathbb{P}(X \in A)$)

$$a_t \sim \pi_{\theta}(\cdot | s_t) \quad (2)$$



Figure 1: Stochastic policy

Multinomial logistic regression

The one way to determine the probability of classify problem is using multinomial logistic regression.

$$\ln (\mathbb{P}(Y = i)) \triangleq \beta_i \mathbf{x} - \ln z \quad (3)$$

$$\mathbb{P}(Y = i) = \frac{1}{z} e^{\beta_i \mathbf{x}} \quad (4)$$

where $i \in [1, n]$, $\mathbf{x} \in \mathbb{R}^n$ and $\beta \in \mathbb{R}^{n \times n}$. The β is adjustable value. From the probability theory,

$$\begin{aligned} \sum_{j=1}^n \mathbb{P}(Y = j) &= 1 = \frac{1}{z} e^{\beta_j \mathbf{x}} \\ \Rightarrow z &= \sum_{j=1}^n e^{\beta_j \mathbf{x}} \end{aligned} \quad (5)$$

Stochastic policy

We use Softmax function(or normalization exponential function) to transfer the output to the probability.

$$\mathbb{P}(Y) = \text{softmax}(i, \mathbf{x}) = \frac{e^{x_i}}{\sum_{j=1}^n e^{x_j}} \quad (6)$$