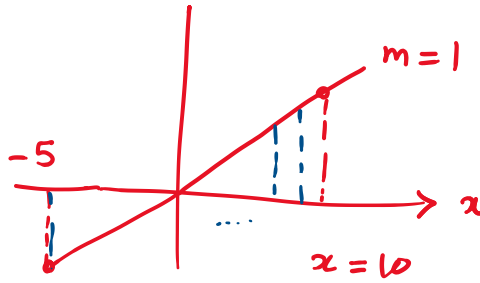


loss

$$\min_x f(x) \\ x \geq -5$$



$$\nabla f = \begin{cases} 1 & 0.5 \\ -1 & 0.5 \end{cases}$$

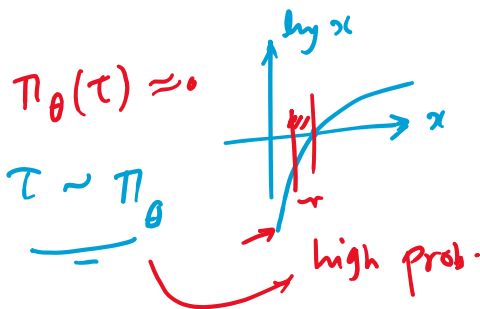
$$\eta = 1$$

$$\nabla_x f = 1 + X$$

$$X = \begin{cases} 2 & 0.5 \\ -2 & 0.5 \end{cases}$$

true grad.

stoch. grad.

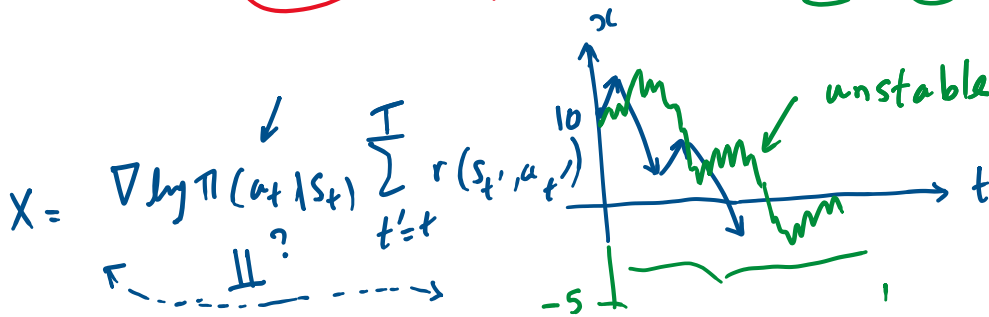
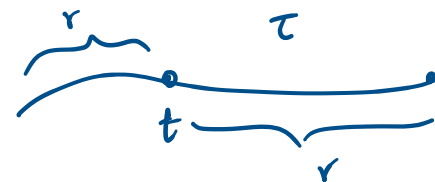


$$x \leftarrow x - 1$$

$$x \leftarrow x - 1 \pm 2$$

$$\begin{matrix} -3 & +1 \end{matrix}$$

$$\text{Var}[\nabla_x f] \downarrow$$



$$\tau = s_1, a_1, s_2, a_2, \dots, s_t, a_t$$

$$\mathbb{E} \left\{ \nabla \log \pi(a_t | s_t) \sum_{t'=t}^T r(s_{t'}, a_{t'}) \right\} = 0$$

$$\mathbb{E}(g(x, \gamma)) = \mathbb{E}_x \mathbb{E}_{\gamma|x} g(x, \gamma)$$

$$= \mathbb{E}_{s_t, a_t} \mathbb{E}_{s_1, a_1, \dots, s_{t-1}, a_{t-1}} \left\{ \nabla \log \pi(a_t | s_t) \sum_{t'=1}^{t-1} r(s_{t'}, a_{t'}) \right\}$$

$$= \mathbb{E}_{s_t, a_t} \left\{ \sum_{t'=1}^{t-1} r(s_{t'}, a_{t'}) \mathbb{E}_{s_t, a_t | s_{t-1}, a_{t-1}} \nabla \log \pi(a_t | s_t) \right\}$$

$$= \int \nabla \pi / \pi \cdot \pi da_t \quad \leftarrow \int g(a_t) \cdot \pi(a_t | s_t) da_t \quad \nabla \pi_\theta / \pi_\theta$$

$$Y \stackrel{?}{\perp\!\!\!\perp} X$$

$$\mathbb{E}[Y] = 0 \leftarrow$$

$$\int \nabla \pi_{\theta} d\alpha_t$$

$$\nabla_{\theta} \int \pi_{\theta} d\alpha_t = 0$$

$$\text{Var}[X + \underline{Y}] \geq \text{Var}[X]$$

$$\text{Var}[X + Y] = \overbrace{\text{Var}[X]} + \overbrace{\text{Var}[Y]} + \underbrace{2 \text{Cov}(X, Y)}_0$$

On-policy
bad

$$\tau_1 \dots \tau_N \leftarrow \pi_{\underline{\theta}},$$

$$\theta' \leftarrow \theta + \underline{\text{noisy grad.}}$$

θ''