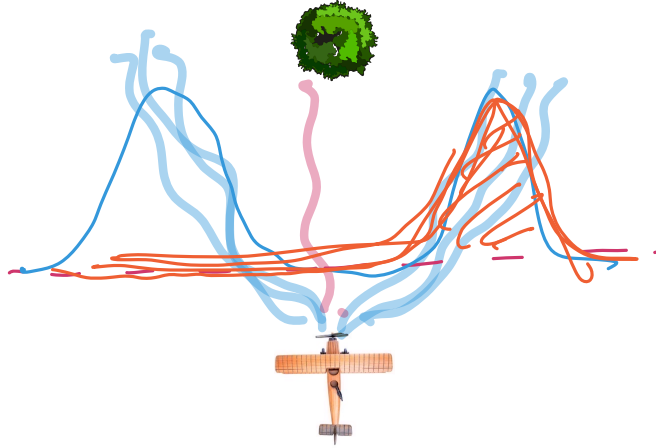


FORWARD KL

$$\min_{\theta} - \mathbb{E}_{z \sim p_{\text{exper}}(z)} \log P_{\theta}(z)$$

$$\downarrow$$
$$- \sum_{z \neq 0} p_{\text{exper}}(z) \log P_{\theta}(z) = 0$$



GOAL: LEARN $P_{\theta}(\Sigma)$

GIVEN: $\Sigma^h \sim P_{\text{exper}}(\Sigma^h)$

FIND

$$\max_{\theta} - \sum_{\Sigma} P_{\theta}(\Sigma) \log P_{\theta}(\Sigma) \quad 0.75$$

s.t.

$$\sum_{\Sigma} P_{\theta}(\Sigma) f^1(\Sigma) = E_{\Sigma \sim P_{\theta}} f^1(\Sigma)$$

$$\sum_{\Sigma} P_{\theta}(\Sigma) f^k(\Sigma) = E f^k(\Sigma)$$

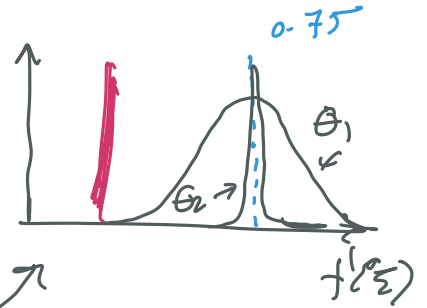
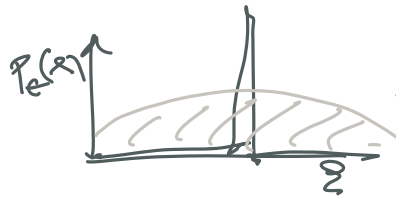
BASIS FEATURE VECTORS

$$f^1(\Sigma) \rightarrow \mathbb{R}$$

$$f^2(\Sigma) \rightarrow \mathbb{R}$$

\vdots

$$f^k(\Sigma) \rightarrow \mathbb{R}$$



= How much grass I drive on?

min
 θ

$$\sum_{\Sigma} P_{\theta}(\Sigma) \log P_{\theta}(\Sigma)$$

s.t

$$\sum_{\Sigma} P_{\theta}(\Sigma) f^1(\Sigma) = c_1 = \begin{matrix} 0.75 \\ 0.75 \end{matrix}$$

(1) Write out the

Lagrangian

$$\max_{\lambda} \min_{\theta} \sum_{\Sigma} P_{\theta}(\Sigma) \log P_{\theta}(\Sigma)$$

$$+ \lambda_1 \left[\sum_{\Sigma} P_{\theta} f^1(\Sigma) - c \right]$$

$$+ \lambda_2 [\dots]$$

$$+ \lambda_3 [\dots]$$

(2)

Take the gradient

$$\frac{\partial}{\partial \lambda} (-) = 0$$

(3) Substitute in

$$\sum_{\Sigma} P_{\theta}(\Sigma) = 1$$

$$P_{\theta}(\mathbf{z}) \propto \exp\left(-\lambda_1 f_1(\mathbf{z}) - \lambda_2 f_2(\mathbf{z}) - \lambda_3 f_3(\mathbf{z}) \dots\right)$$

$$\propto \exp\left(-\text{cost}_{\theta}(\mathbf{z})\right)$$

$$= \frac{1}{Z(\theta)} \exp\left(-\text{cost}_{\theta}(\mathbf{z})\right)$$

Play back in to optimization

$$\max_{\theta} \sum_{i=1}^N \log P_{\theta}(\mathbf{z}_i^h)$$

$$\sum_{i=1}^N \log \left(\frac{1}{Z(\theta)} \exp\left(-\text{cost}_{\theta}(\mathbf{z}_i^h)\right) \right)$$

$$= \sum_{i=1}^N -\text{cost}_{\theta}(\mathbf{z}_i^h) - \log Z(\theta)$$

$$\min_{\theta} \sum_{i=1}^N \text{cost}_{\theta}(\mathbf{z}_i^h) + \log Z(\theta)$$