Loss Entropy:

KL divergence:

Coin 1
- % heads (p_1)
- % tails (p_2)

Coin 2
- p heads
- q tails

$$\frac{P(\text{observations} \mid \text{real coin})}{P(\text{observations} \mid \text{coin 2})} = \left(\frac{p_1^{N_H} \; p_2^{N_T}}{q_1^{N_H} \; q_2^{N_T}}\right)^{\frac{1}{N}}$$

$$\xrightarrow[\log]{\frac{1}{N}} \log\left(\frac{p_1^{N_H} \; p_2^{N_T}}{q_1^{N_H} \; q_2^{N_T}}\right)^{\frac{1}{N}} = \frac{1}{N}\log p_1^{N_H} + \frac{1}{N}\log p_2^{N_T}$$

$$-\frac{1}{N}\log q_1^{N_H} - \frac{1}{N}\log q_2^{N_T} = p_1 \log p_1 + p_2 \log p_2$$

$$-p_1 \log q_1 - p_2 \log q_2 = p_1 \log \frac{p_1}{q_1} + p_2 \log \frac{p_2}{q_2},$$

$$D_{KL}(P \| Q) = \sum_i P(i) \log \frac{P(i)}{Q(i)}$$

Now cross Entropy Loss:

$$D_{KL}\left(\overbrace{p^*(y \mid x_i)}^{\text{True class distribution}} \;\|\; \overbrace{p(y \mid x_i; \theta)}^{\text{predicted class distribution}}\right) = \sum p^*(y \mid x_i) \log \frac{p^*(y \mid x_i)}{p(y \mid x_i; \theta)}$$

$$= \sum_y \underbrace{p^*(y|x_i) \log p^*(y|x_i)}_{\text{doesn't depend on } \theta} - \sum_y p^*(y|x_i) \log(y|x_i;\theta)$$

$$\Rightarrow \underset{\theta}{\operatorname{argmin}} \; D_{KL}(p^* \| p) = \underset{\theta}{\operatorname{argmin}} - \sum_y p^*(y|x_i) \log p(y|x_i;\theta)$$

$$\Downarrow$$

$$\underset{\theta}{\operatorname{argmin}} \; H(p^*, p)$$