

$$H(\vec{p}) = - \sum p_i \log p_i$$

$$\text{max if } p_1 = p_2 = \dots = p_n$$

prove this

#1: • Start with $n=2$ $\vec{p} = (p, 1-p)$

use calculus

• $n \geq 3$: use "2nd year" calculus

Optimization under constraint.

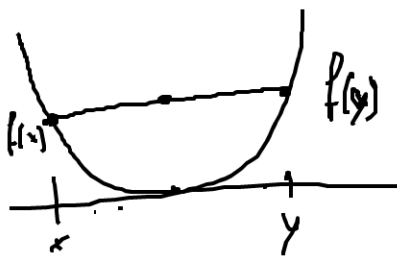
$$\text{max } f(\vec{x}) \text{ given } \underline{g(\vec{x}) = c}$$

$$\Lambda = f - \lambda(g - c) \quad \text{Lagrange multipliers} \quad \searrow$$

$$\text{max } H(\vec{p}) \text{ subject to } \sum p_i = 1$$

Jensen's inequality:

f convex



$$\lambda f(x) + (1-\lambda) f(y) \geq f(\lambda x + (1-\lambda)y)$$

$$\lambda \in [0, 1]$$

strictly convex : $= \Rightarrow \lambda = 0, 1$

- log convex

$$\mathbb{E}(f(X)) \geq f(\mathbb{E}X)$$

$$\mathbb{E}(f(X)) = f(\mathbb{E}X)$$

$$E(f(X)) = f(E(X))$$

X binary

$$P \cdot f(x_1) + (1-P) f(x_2) = f(Px_1 + (1-P)x_2)$$

$$E f(x) \quad f \text{ strictly convex}$$

$$\Rightarrow p = 0, 1$$

$$\Rightarrow X \text{ is constant}$$

$$\Rightarrow \underline{E X = X}$$

we induction to generalize to n -dim. X .
 Jensen's inequ. is used to show

$$\underline{D(P \parallel q) \geq 0}$$

$$\left. \begin{array}{l} \text{infer this} \\ D(P \parallel q) = 0 \\ \text{iff } P = q \end{array} \right\}$$

show that $D(p \parallel u) = \log |A| - H(x)$

\Downarrow \uparrow
 0 uniform
 \Downarrow \uparrow
 iff $p = u$ $x = A\text{-value}$

$$g: \underbrace{\mathcal{X}}_A \rightarrow \mathbb{R}$$

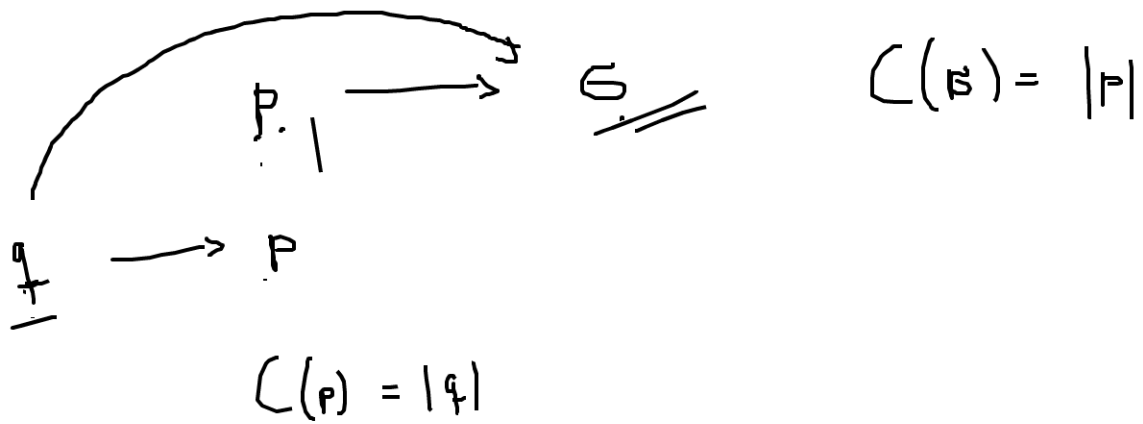
$$H(Y|X) \quad \text{definition}$$

$$\boxed{H(g(X)) \leq H(X)} \leftarrow$$

$$H(X, g(X)) = H(X) + \underbrace{H(g(X) | X)}_{?}$$

$$H(X, g(X)) = H(g(X)) + H(X | g(X))$$

$$\geq \dots$$



If $|q| \ll |P| \rightarrow C(\sigma) \ll |P|$

$$K(\sigma) \leq |\sigma| + 2 \log |\sigma|$$

$$K(\sigma) \leq 2|\sigma|$$

prefix-free copy machine

$$\begin{array}{c} |\sigma| \\ 0 \quad \sigma \\ \dots \end{array} \mapsto \sigma$$

$2|\sigma| + 1$

$|\sigma| = n \leftarrow$ binary repes. s_n

$$|s_n| = \log n$$

$$\begin{array}{c} s_n \\ 0 \quad |s_{|\sigma|}| \quad |s_{|\sigma|}| \quad \sigma \\ \dots \end{array} \mapsto \sigma$$

$$|s_{|\sigma|}| + 1 + |s_{|\sigma|}| + |\sigma| = |\sigma| + 2 \log |\sigma| + 1$$