

Differential Entropy

- * Extension of entropy for continuous r.v.s.
- * Let X be r.v. with pdf f whose support is \mathcal{X} .

$$\text{Differential entropy } h(X) = - \int_{\mathcal{X}} f(x) \log f(x) dx$$

Differential Entropy

$$\text{Differential entropy } h(x) = - \int_x f(z) \log f(z) dz$$

UNIFORM DISTRIBUTION

$$f(z) = \frac{1}{b-a} \quad ; \quad x = [a, b]$$

$$\begin{aligned} h(x) &= - \int_a^b \frac{1}{b-a} \log \left(\frac{1}{b-a} \right) dz = - \left[\left(\frac{1}{b-a} \right) \log \frac{1}{b-a} z \right]_a^b \\ &= - \left[\left(\frac{1}{b-a} \right)^{b-a} \log \frac{1}{b-a} \right] \\ &= \log(b-a) \end{aligned}$$

Differential Entropy

$$\text{Differential entropy } h(x) = - \int_{-\infty}^{\infty} f(x) \log f(x) dx$$

1D NORMAL DISTRIBUTION

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) : X = [-\infty, \infty]$$

$$h(x) = - \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) \log \left(\frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) \right) dx$$

Differential Entropy

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1D NORMAL DISTRIBUTION

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$$h(x) = - \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma^2} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) \left[\log\left(\frac{1}{\sqrt{2\pi}\sigma^2}\right) - \left(\frac{x-\mu}{2\sigma^2}\right)^2 \right] dx$$
$$= \log_e\left(\sigma\sqrt{2\pi e}\right) = \frac{1}{2} \log(2\pi e \sigma^2)$$

Differential Entropy

$$\text{Differential entropy } h(x) = - \int_{\mathcal{X}} f(x) \log f(x) dx$$

MULTIVARIATE GAUSSIAN

$$x \sim N(\mu, \Sigma)$$

$$h(x) = \frac{1}{2} \log [(2\pi e)^n |\Sigma|]$$

Differential Entropy

$$\text{Differential entropy } h(x) = - \int_{\mathcal{X}} f(x) \log f(x) dx$$

MULTIVARIATE GAUSSIAN

$$x \sim N_n(\mu, \Sigma)$$

$$h(x) = \frac{1}{2} \log [(2\pi e)^n |\Sigma|]$$

$$\text{e.g. } x \sim N \left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right)$$

$$h_1(x) = \frac{1}{2} \log [(2\pi e)^2 (1)]$$

Differential Entropy

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MULTIVARIATE GAUSSIAN

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$$\text{e.g. } x_1 \sim N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}\right)$$

$$h(x) = \frac{1}{2} \log [(2\pi e)^2 (1)]$$

$$\left. \begin{array}{l} x_2 \sim N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0.5 & 0.5 \\ 0.5 & 1 \end{bmatrix}\right) \\ h(x_2) = \frac{1}{2} \log [(2\pi e)^2 (0.75)] \end{array} \right\}$$

Differential Entropy

$$\text{Differential entropy } h(x) = - \int_{-\infty}^{\infty} f(x) \log f(x) dx$$

$$x_1 \sim N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}\right) \quad \left| \quad x_2 \sim N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0.5 \\ 0.5 & 1 \end{bmatrix}\right)\right.$$

$$h(x_1) = \frac{1}{2} \log \left[(2\pi e)^2 (1) \right] \quad h(x_2) = \frac{1}{2} \log \left[(2\pi e)^2 0.75 \right]$$

$$2h(x_2) - 2h(x_1) = \log \left[\frac{(2\pi e)^2}{(2\pi e)^2} \frac{0.75}{1} \right]$$

$$= \log (0.75) = -0.28$$

\Rightarrow More entropy in x_1

CONDITIONAL DIFFERENTIAL ENTROPY

let X and Y be continuous rvs with joint pdf
 $f(x,y)$

conditional Entropy

$$h(X|Y) = - \int_{-\infty}^{\infty} f(x,y) \log f(x|y) dx dy$$

CONDITIONAL DIFFERENTIAL ENTROPY

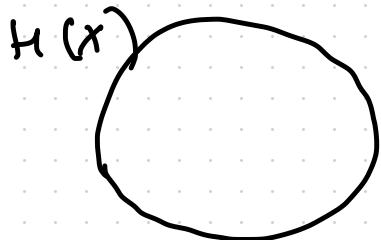
let X and Y be continuous rvs with joint pdf
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conditional Entropy

$$h(X|Y) = - \int_{x,y} f(x,y) \log f(x|y) dx dy$$

$$h(X|Y) = h(X,Y) - h(Y)$$

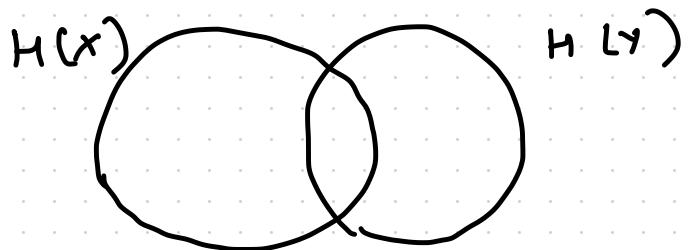
CONDITIONAL DIFFERENTIAL ENTROPY



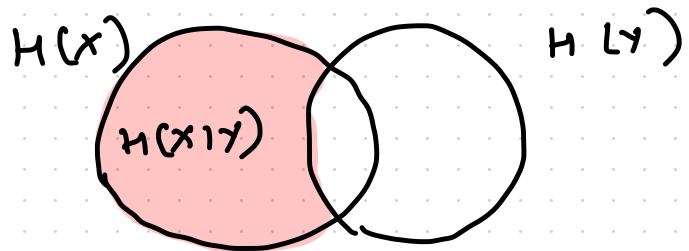
$H(x)$

$H \rightarrow$ Discrete
 $h \rightarrow$ Continuous

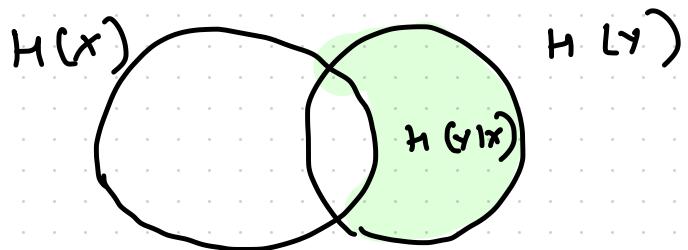
CONDITIONAL DIFFERENTIAL ENTROPY



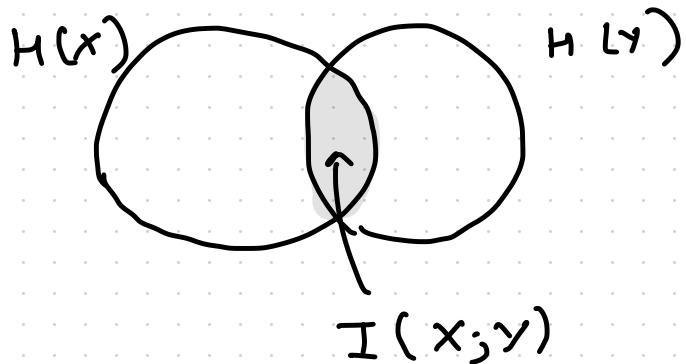
CONDITIONAL DIFFERENTIAL ENTROPY



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CONDITIONAL DIFFERENTIAL ENTROPY



MUTUAL INFORMATION

$$= H(X) - H(X|Y)$$

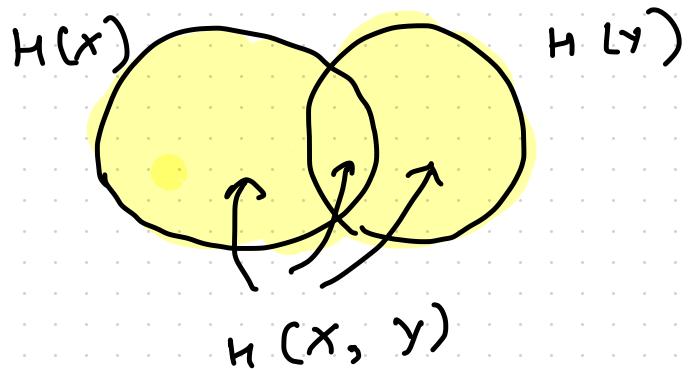
$$\boxed{I(X;Y) = H(Y) - H(Y|X)}$$

Discrete

$$\boxed{I(X;Y) = h(y) - h(y|x)}$$

Continuous

CONDITIONAL DIFFERENTIAL ENTROPY



CONDITIONAL

DIFFERENTIAL

ENTROPY

Let $X \sim N_2(0, \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix})$

CONDITIONAL

DIFFERENTIAL

ENTROPY

Let $X \sim N_2(0, \begin{bmatrix} 1 & p \\ p & 1 \end{bmatrix})$

$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \sim N \left[\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & p \\ p & 1 \end{pmatrix} \right]$

$$\begin{aligned} h(x) &= h(x_1, x_2) = \frac{1}{2} \log [(2\pi e)^n | \Sigma |] \\ &= \frac{1}{2} \log \left[(2\pi e)^2 (1-p^2) \right] \end{aligned}$$

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$$\begin{aligned} h(x) &= h(x_1, x_2) = \frac{1}{2} \log [(2\pi e)^n] \in \mathbb{R} \\ &= \frac{1}{2} \log \left[(2\pi e)^2 (1-p^2) \right] \end{aligned}$$

$$h(x_1) = \frac{1}{2} \log (2\pi e) = h(x_2)$$

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Let $X \sim N_2(0, \begin{bmatrix} 1 & p \\ p & 1 \end{bmatrix})$

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \sim N \left[\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & p \\ p & 1 \end{pmatrix} \right]$$

$$h(x) = \frac{1}{2} \log \left((2\pi e)^2 (1-p^2) \right)$$

$$h(x_1) = \frac{1}{2} \log (2\pi e) = h(x_2)$$

$$\begin{aligned} h(x_1|x_2) &= h(x_1, x_2) - h(x_2) = h(x) - h(x_2) \\ &= \frac{1}{2} \log \frac{(2\pi e)^2 (1-p^2)}{(2\pi e)} \end{aligned}$$

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$$\begin{aligned} I(x_1; x_2) &= h(x_1) - h(x_1|x_2) \\ &= \frac{1}{2} \log 2\pi e - \frac{1}{2} \log 2\pi e (1-p^2) = \frac{-1}{2} \log (1-p^2) \end{aligned}$$

CONDITIONAL

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Let $X \sim N_2(0, \begin{bmatrix} 1 & p \\ p & 1 \end{bmatrix})$

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$$p \rightarrow \pm 1 \Rightarrow I(X_1; X_2) \rightarrow \infty$$

Perfect Correlatⁿ: M.I. $\rightarrow \infty$

CONDITIONAL

DIFFERENTIAL

ENTROPY

Let $X \sim N_2(0, \begin{bmatrix} 1 & p \\ p & 1 \end{bmatrix})$

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$$p \rightarrow \pm 1 \Rightarrow I(X_1; X_2) \rightarrow \infty$$

Perfect Correlatⁿ: M.I. $\rightarrow \infty$

$p \rightarrow 0 \Rightarrow I(X_1; X_2) \rightarrow 0$ No correlatⁿ: M. I. $\rightarrow 0$

