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JOINT ENTROPY

$$H(X, Y) = \sum_i \sum_j P_{ij} \log_2 \left(\frac{1}{P_{ij}} \right)$$

$$P_{ij} = \Pr(X = x_i, Y = y_j)$$

↗ $H(X, Y) = 3.375 \text{ bits}$
From example.

Marginal Probabilities.

$$\begin{aligned} P_r(X = s_0) &= P_r(X = s_1) \\ &= P_r(X = s_2) = P_r(X = s_3) \\ &= \frac{1}{4}. \end{aligned}$$

Entropy of X is

$$H(X) = H\left(\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4}\right)$$

4 symbols with
probability $= \frac{1}{4}$
each

$$H(X) = 4 \times \frac{1}{4} \log_2 4$$

$$= 1 \times 2 = 2 \text{ bits.}$$

$$H(Y) = H\left(\frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \frac{1}{8}\right)$$

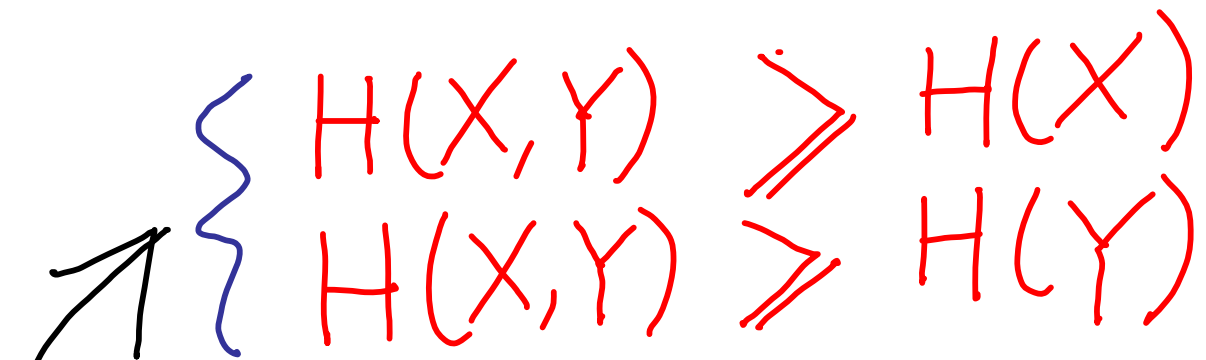
$$= \frac{1}{2} \log_2 2 + \frac{1}{4} \log_2 4$$

$$+ \frac{1}{8} \log_2 8 + \frac{1}{8} \log_2 8$$

$$= \frac{1}{2} + \frac{1}{2} + \frac{3}{8} + \frac{3}{8}$$

$$= 1 + \frac{3}{4} = \underline{1.75 \text{ bits.}}$$

$$H(X, Y) = 3.375 > H(X) = 2$$
$$H(X, Y) > H(Y) = 1.75$$


$$\begin{cases} H(X, Y) > H(X) \\ H(X, Y) > H(Y) \end{cases}$$

Joint Entropy or
information in X & Y
 $\geq$ information in
either X or Y .

$$\begin{aligned} H(X) + H(Y) &= 2 + 1.75 \\ &= 3.75 > H(X, Y) \\ &= 3.375 \end{aligned}$$

$$H(X, Y) \leq H(X) + H(Y)$$

Joint Entropy or
Total Information is
lower or equal to sum
of entropies of individual
sources.

$$H(X, Y) = \sum_i \sum_j P_r(x_i, y_j) \log_2 \frac{1}{P_r(x_i, y_j)}$$

$$H(X) = \sum_i \underbrace{P_r(x_i)} \log_2 \frac{1}{P_r(x_i)}$$

$$= \sum_i \underbrace{\sum_j P_r(x_i, y_j)} \log_2 \frac{1}{P_r(x_i)}$$

$$= P_r(x_i) \leftarrow$$

From total Prob rule.

$$H(Y) = \sum_j P_r(y_j) \log_2 \frac{1}{P_r(y_j)}$$

$$= \sum_j \sum_i P_r(x_i, y_j) \log_2 \left(\frac{1}{P_r(y_j)} \right)$$

$$H(X, Y) - H(X) - H(Y)$$

$$\begin{aligned}
 &= \underbrace{\sum_i \sum_j P_r(s_i, r_j) \log_2 \left(\frac{1}{P_r(s_i, r_j)} \right)}_{H(X, Y)} \\
 &\quad - \underbrace{\sum_i \sum_j P_r(s_i, r_j) \log_2 \left(\frac{1}{P_r(s_i)} \right)}_{H(X)} \\
 &\quad - \underbrace{\sum_i \sum_j P_r(s_i, r_j) \log_2 \left(\frac{1}{P_r(r_j)} \right)}_{H(Y)}.
 \end{aligned}$$

$$= \sum_i \sum_j P_{ij}(s_i, t_j) \left\{ \log_2 \frac{1}{P_{ij}(s_i, t_j)} - \log_2 \frac{1}{P_i(s_i)} - \log_2 \frac{1}{P_j(t_j)} \right\}$$

$$= \sum_i \sum_j P_{ij}(s_i, t_j) \log_2 \left(\frac{P_i(s_i) P_j(t_j)}{P_{ij}(s_i, t_j)} \right)$$

$$H(X, Y) - H(X) - H(Y)$$

$$= \sum_{i,j} P_{ij}(s_i, t_j) \log_2 \left(\frac{P_i(s_i) P_j(t_j)}{P_{ij}(s_i, t_j)} \right)$$

$$\sum_i p_i \log_2 x_i$$

$$\leq \log_2 \cdot \sum_i p_i x_i$$

Jensen's Inequality
Since $\log_2 x$ is
concave.

$$p_i \geq 0$$

$$\sum p_i = 1$$

$$\sum_{i,j} \Pr(x_i, y_j) \log_2 \left(\frac{\Pr(x_i) \Pr(y_j)}{\Pr(x_i, y_j)} \right)$$

$$\leq \log_2 \left(\sum_{i,j} \cancel{Pr(x_i, y_j)} \frac{Pr(x_i)Pr(y_j)}{\cancel{Pr(x_i, y_j)}} \right)$$

$$= \log_2 \left(\sum_i \sum_j Pr(x_i) Pr(y_j) \right)$$

$$= \log_2 \left(\underbrace{\left(\sum_i Pr(x_i) \right)}_1 \underbrace{\left(\sum_j Pr(y_j) \right)}_1 \right)$$

$$= \log_2 1 = 0.$$

$$H(X, Y) - H(X) - H(Y) \leq 0$$

Jensen's
Inequality

$$H(X, Y) - H(X) - H(Y) \leq 0$$

$$\Rightarrow H(X, Y) \leq H(X) + H(Y)$$