MTHE 474 Notes

Timothy Liu September 15, 2022

Information Measures for Discrete Systems

Definitions

• **Definition 2.2:** Entropy of discrete random variable X with pmf $P_X(*)$ is defined as

$$H(x) := -\sum_{x \in X} P_X(x) * \log_2 P_X(x)$$

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• Definition 2.8 (Joint entropy):

$$H(X,Y) := -\sum_{(x,y)\in\mathcal{X}\times\mathcal{Y}} P_{X,Y}(x,y) * \log_2 P(X,Y)(x,y)$$

• Definition 2.9 (Conditional entropy):

$$H(Y|X) := \sum_{x \in \mathcal{X}} P_X(x) \left(-\sum_{y \in \mathcal{Y}} P_{Y|X}(y|x) * \log_2 P_{Y|X}(y|x)\right)$$

Lemmas/Theorems

• Lemma 2.4 (Fundamental Inequality): $\forall x > 0$ and D > 1 we have

$$\log_D\left(X\right) \le \log_D e * (x-1)$$

- Lemma 2.5 (Non-negativity): $H(X) \ge 0$
- Lemma 2.6 (Entropy Upper-Bound): $H(X) \leq \log_2 |\mathcal{X}|$ where random variable X takes values from finite set \mathcal{X}
- Lemma 2.7 (Log-Sum inequality) Write this one out later
- Theorem 2.10 (Chain rule for entropy): H(X,Y) = H(X) + H(Y|X)