Information Theory

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INFORMATION THEORY

 Information Theory is a branch of applied mathematics that deals with quantifying information, communication, and data transmission.



How is Information Theory Relevant?









Data Science and Technology

- Machine learning
- AI
- Data Mining

Communication Technology

- Internet
- Mobile networks
- Broadcasting





Biological Sciences and Psychology/Human Perception

- Genetics analyzing
- Neuroscience
- Cognitive development
- Ecology

Fundamental Concepts of Information Theory

Entropy

Differential Entropy

For continuous variable with pdf *f*:

$$H(X) = \int f(x) \log \frac{1}{f(x)} dx$$

H(X) is the expectation of log(1/f(x)) which is called the *loss*.



Kullback-Leibler Divergence

Directed Divergence

Kullback-Leibler risk of *g* related to *f*:

$$\mathsf{CE}(\mathsf{P}||\mathsf{Q}) = \int \mathsf{P}(\mathsf{x}) \log \frac{\mathsf{P}(\mathsf{x})}{Q(\mathsf{x})} \, \mathsf{d}\mathsf{x}$$



Cross Entropy

• Cross Entropy $CE(g|f) = \int f(x) \log \frac{1}{g(x)}$ dx



Mutual Information



Applications of Information Theory

Parameter Estimation

- Information Theory offers methods that are relevant to parameter estimation including:
 - Maximum Likelihood Estimator (MLE)
 - Bayesian Estimation

Data Analysis

- Offers tools and concepts that are invaluable to data analysis
 - Entropy and Information Gain
 - Mutual Information
 - Kullback-Leibler Divergence
 - Coding theory and Compression

Variational Inference

- Alternative to MCMC
- This is a technique used in Bayesian statistics and machine learning and is meant to approximate complex probability distributions

Data Compression

- Primary goal: represent information in a more efficient manner, using as few bits as possible to store or transmit data
 - Entropy and Redundancy
 - Lossless vs Lossy Compression
 - Arithmetic Coding
 - Representing messages by a number within a specified range

Thank you!