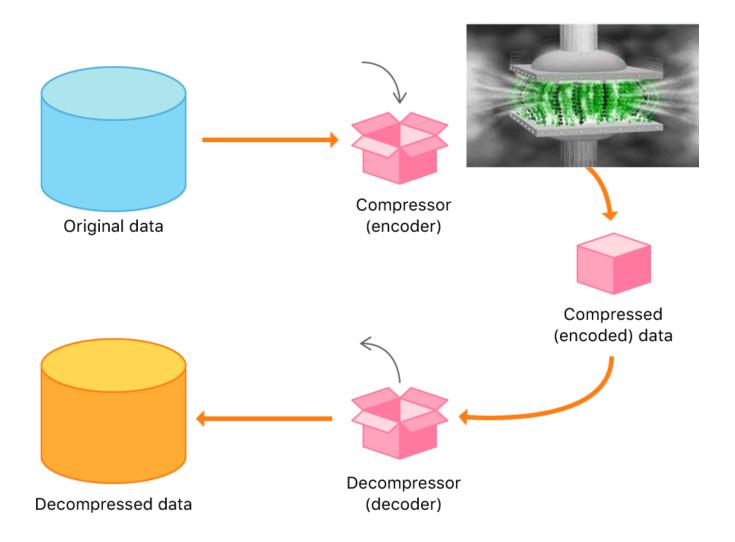


# Source Coding

## What is Data Compression?





## What is data compression?

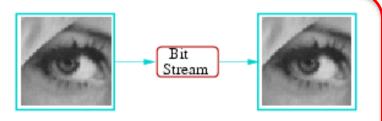


- Data compression is the representation of an information source (e.g., data file, speech signal, video signal) as accurately as possible using the fewest number of bits
- Compressed data can only be understood if the decoding method is known by the receiver

## Source Coding







 Lossless compression: the compressed image/video can be decompressed to be identical to the original

**Lossy** (images, videos)

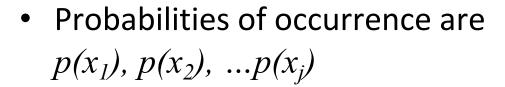


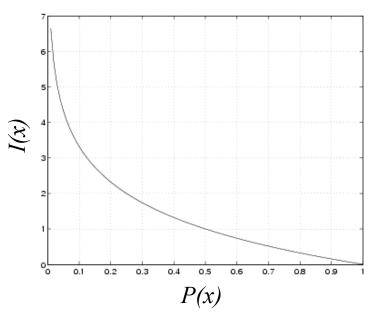
 With lossy compression, the image is degraded, mathematically not identical to original. Data compression introduces a distortion of the source.

#### Source Information



• Let X be a discrete random variable taking on values  $x_1, x_2, \dots x_j$  from a finite alphabet A





The information associated with symbol s<sub>i</sub> is defined to be

$$I(x_i) = \log_2 \frac{1}{p(x_i)}$$

### Entropy



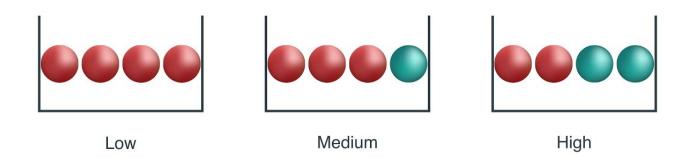
- (First-order) Entropy is a measure of the uncertainty of a random variable.
- Entropy is the average information per symbol

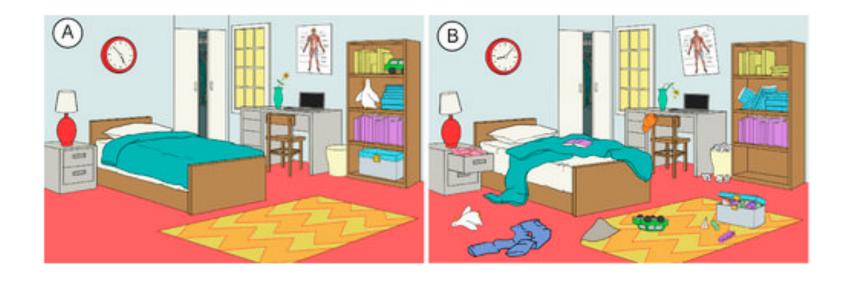
$$H(X) = \sum_{i=1}^{J} p(x_i) \log_2 \frac{1}{p(x_i)}$$

- Entropy is expressed in bits
- Entropy of a fair coin toss is 1 bit
- Average length of the shortest description of the random variable

## Entropy





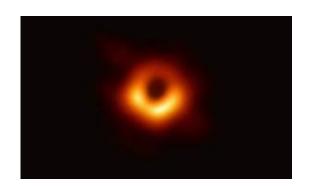


#### Source Information





Sunrise Time – not that much informative



Black Hole Pic – highly informative

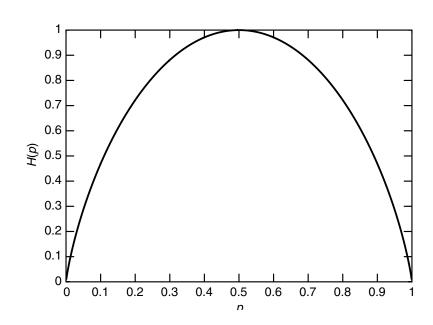
### Example 1



Let 
$$X = \begin{cases} 1 & \text{with probability } p, \\ 0 & \text{with probability } 1-p \end{cases}$$
  
Then  $H(X) = -p \log_2 p - (1-p) \log_2 (1-p) \stackrel{\text{def}}{=} H(p)$ 

Entropy of a fair coin toss is 1 bit.

Or in other words  $p=0.5 \rightarrow H(p)=1$ 



### Example 2



Let 
$$X = \begin{cases} a & \text{with probability } \frac{1}{2}, \\ b & \text{with probability } \frac{1}{4}, \\ c & \text{with probability } \frac{1}{8}, \\ d & \text{with probability } \frac{1}{8}. \end{cases}$$

Then, the entropy of X is

$$H(X) = -\frac{1}{2} \log_2 \frac{1}{2} - \frac{1}{4} \log_2 \frac{1}{4} - \frac{1}{8} \log_2 \frac{1}{8} - \frac{1}{8} \log_2 \frac{1}{8} = \frac{7}{4} \text{ bits}$$

What is the minimum number of questions you need to determine *X*?

The resulting expected number of binary questions required is 1.75.

The minimum expected number of binary questions required to determine X lies between H(X) and H(X) + 1.



## Thank You