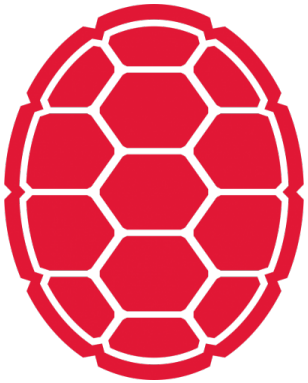
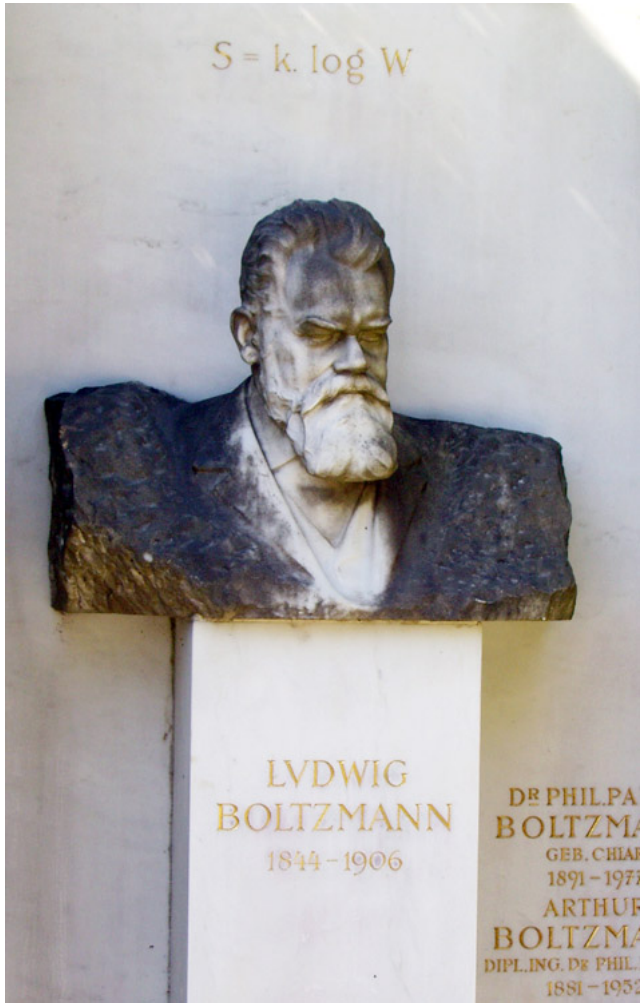


Mathematical Foundations

Data Science: Jordan Boyd-Graber
University of Maryland
SLIDES ADAPTED FROM DAVE BLEI AND LAUREN HANNAH

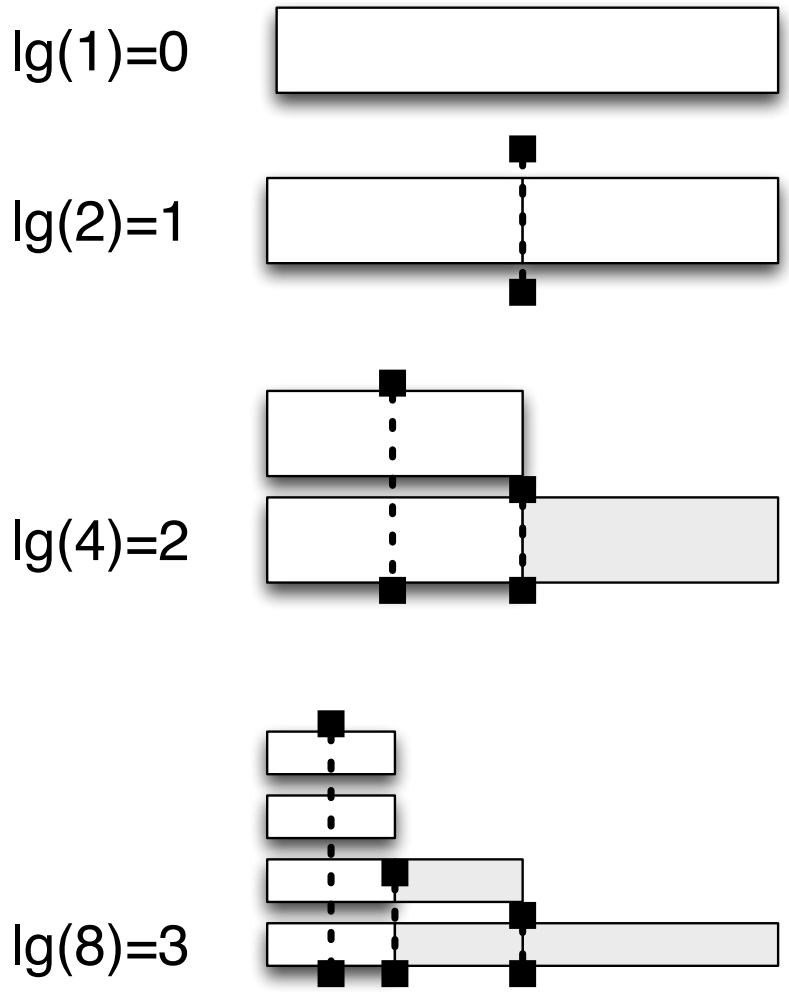
Entropy

- Measure of disorder in a system
- In the real world, entroy in a system tends to increase
- Can also be applied to probabilities:
  - Is one (or a few) outcomes certain (low entropy)
  - Are things equiprobable (high entropy)
- In data science
  - We look for features that allow us to reduce entropy (decision trees)
  - All else being equal, we seek models that have maximum entropy (Occam's razor)



Aside: Logarithms

- $\lg(x) = b \Leftrightarrow 2^b = x$
- Makes big numbers small
- Way to think about them: cutting a carrot



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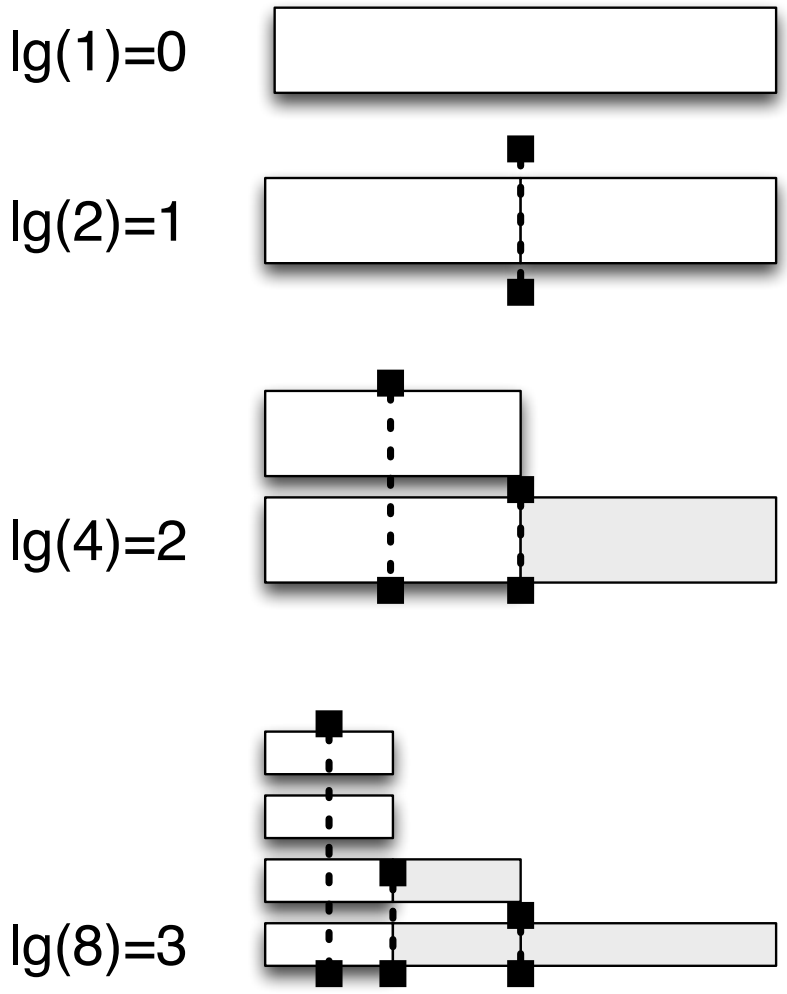

Aside: Logarithms

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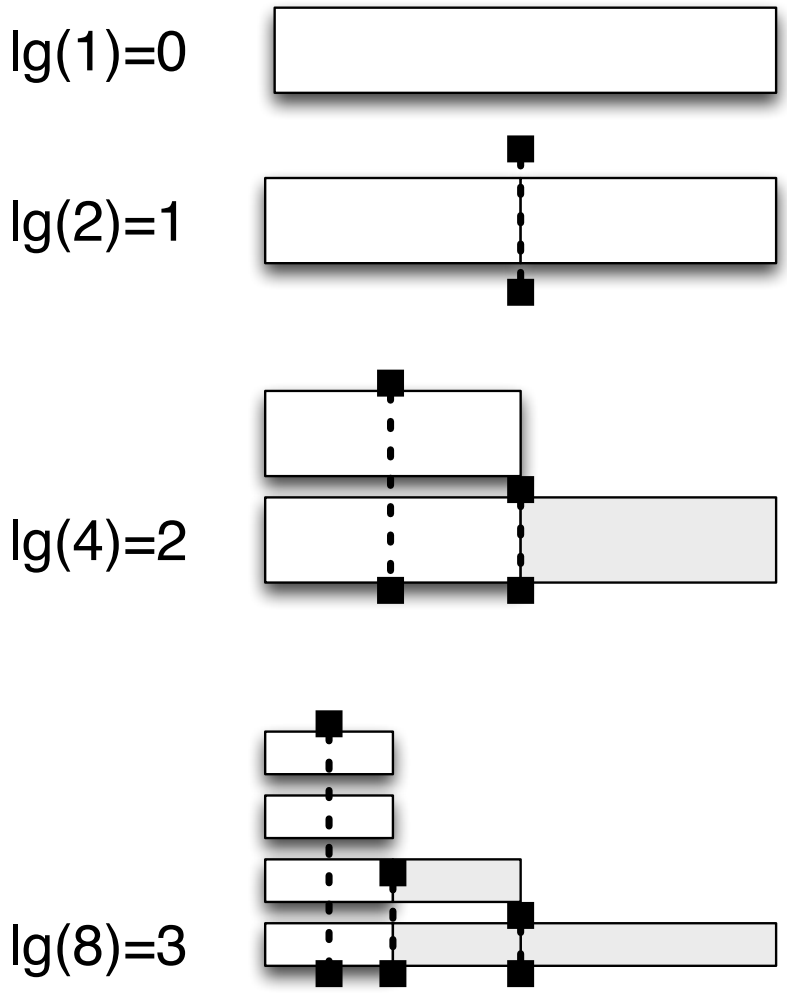
- Way to think about them: cutting a carrot
- Negative numbers?
- Non-integers?



Entropy

Entropy is a measure of uncertainty that is associated with the distribution of a random variable:

$$
\begin{aligned}
H(X) &= -\mathrm{E}\left[\lg(p(X))\right] \\
&= -\sum_{x} p(x)\lg(p(x)) && \text{(discrete)} \\
&= -\int_{-\infty}^{\infty} p(x)\lg(p(x))\,dx && \text{(continuous)}
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Does not account for the values of the random variable, only the spread of the distribution.

- $H(X) \geq 0$
- uniform distribution $=$ highest entropy, point mass $=$ lowest
- suppose $P(X=1)=p,\ P(X=0)=1-p$ and $P(Y=100)=p,\ P(Y=0)=1-p$: X and Y have the same entropy

Wrap up

- Probabilities are the language of data science
- You'll need to manipulate probabilities and understand marginalization and independence
- Thursday: Working through probability examples
- Next week: **Conditional** probabilities