

Information & Entropy

Comp 595 DM

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Information & Entropy

- Information Equation

$$I(p) = -\log_b(p)$$

p = probability of the event happening

b = base

(base 2 is mostly used in information theory)

*unit of information is determined by base

base 2 = bits

base 3 = trits

base 10 = Hartleys

base e = nats

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- Example of Calculating Information

Coin Toss

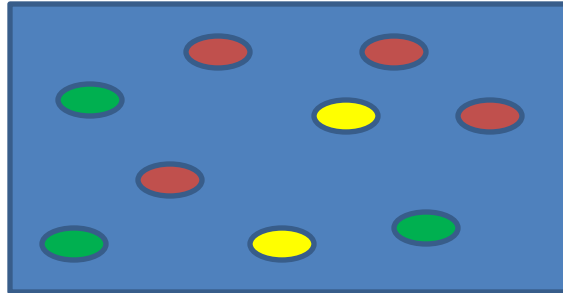
There are two probabilities in fair coin, which are head(.5) and tail(.5).

So if you get either head or tail you will get 1 bit of information through following formula.

$$I(\text{head}) = -\log(.5) = 1 \text{ bit}$$

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- Another Example
Balls in the bin



The information you will get by choosing a ball from the bin are calculated as following.

$$I(\text{red ball}) = -\log(4/9) = 1.1699 \text{ bits}$$

$$I(\text{yellow ball}) = -\log(2/9) = 2.1699 \text{ bits}$$

$$I(\text{green ball}) = -\log(3/9) = 1.58496 \text{ bits}$$

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- Then, what is Entropy?
 - Entropy is simply the average(expected) amount of the information from the event.
- Entropy Equation

$$\text{Entropy} = - \sum_{i=1}^n p_i \log_b(p_i)$$

n = number of different outcomes

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- How was the entropy equation is derived?

$$I = - \sum_{i=1}^n (N * p_i) * \log_b(p_i)$$

I = total information from N occurrences

N = number of occurrences

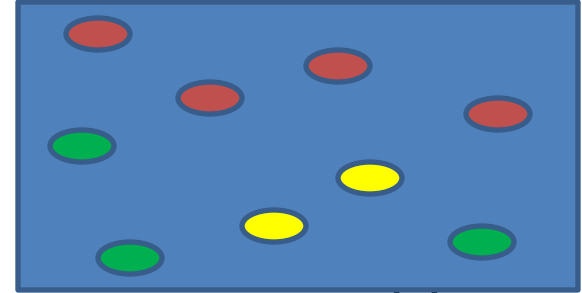
(N*Pi) = Approximated number that the certain result will come out in N occurrence

$$\text{Entropy} = - \sum_{i=1}^n p_i \log_b(p_i)$$

So when you look at the difference between the total Information from N occurrences and the Entropy equation, only thing that changed in the place of N. The N is moved to the right, which means that I/N is Entropy. Therefore, Entropy is the average(expected) amount of information in a certain event.

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- Let's look at this example again...
Calculating the entropy....



In this example there are three outcomes possible when you choose the ball, it can be either red, yellow, or green. ($n = 3$)

So the equation will be following.

$$\text{Entropy} = - \sum_{i=1}^3 p_i \log_b(p_i)$$
$$\begin{aligned} \text{Entropy} &= - (4/9) \log(4/9) + - (2/9) \log(2/9) \\ &\quad + - (3/9) \log(3/9) \\ &= 1.5304755 \end{aligned}$$

Therefore, you are expected to get 1.5304755 information each time you choose a ball from the bin

Clear things up.

- Does Entropy have range from 0 to 1?
 - No. However, the range is set based on the number of outcomes.
 - Equation for calculating the range of Entropy:
 $0 \leq \text{Entropy} \leq \log(n)$, where n is number of outcomes
 - Entropy 0(minimum entropy) occurs when one of the probabilities is 1 and rest are 0's
 - Entropy $\log(n)$ (maximum entropy) occurs when all the probabilities have equal values of $1/n$.

If you want more information...

- <http://csustan.csustan.edu/~tom/sfi-csss/info-theory/info-lec.pdf>
 - Look at pages from 15 to 34. This is what I read and prepared all the information that are on the current powerpoint slides. Very simple and easy for students to understand.
- <http://ee.stanford.edu/~gray/it.pdf>
 - Look at chapter two of this pdf file, it has very good detailed explanation of Entropy and Information theory.