# Information Entropy 

A walk-through

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- Increase as number of possible events increases.
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The uncertainty contained in a probability distribution is the average log-probability of an event.

## But why logarithms?

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- Can you guess which one?

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## But why logarithms?

I have selected a number between 0 and $31 .{ }^{1}$

- Can you guess which one?
- You can ask as many questions as you want.
- What is the minimum number of questions you have to ask to be $100 \%$ sure?

[^2]
## Let's do the math

We could pick between $0, \ldots, 31$, i.e., 32 numbers. Let's rewrite as an equation,

$$
\log _{2}(32)=x
$$

## Proof.

If $x$ and $b$ are positive real numbers, and $b$ does not equal 1 , then $\log _{b}(x)=y$ is equivalent to $b^{y}=x$,

$$
2^{x}=32
$$

Create equivalent expressions in the equation that all have equal bases,

$$
2^{x}=2^{5}
$$

Since the bases are the same, the two expressions are only equal if the exponents are also equal,

$$
x=5
$$

- In other words, we need to take the base two logarithm of 32 to get the number of questions required, i.e., 5 .
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- This logic applies to all numbers! If we pick a number between 0 and $n-1$, you need $\log (2, n)$ questions to find the number.


## Since we are nerds...

If we write down the answers in a row, we effectively encode the numbers in $n$ bits.
000000
100001
200010

3111111
Each 'code' is simply the number in base two.

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However, we have an assumption here,
Corollary
Each number is picked with an equal probability.
What if this is not the case?

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& P(X=1)=1 / 4 \\
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But we can be a bit more 'careful' with our 'bits', i.e., which question to ask.
Consider this,
1st $Q$ did you pick 0 ? If the answer is 'Yes', the 2 nd question is not needed. If not, we proceed! 2nd Q did you pick 1? No matter what the answer is, we know the solution! 'Yes' implies 1, 'No' implies 2.

## Let's do some math

avg. num. bits $=-\frac{1}{2} \log _{2}\left(\frac{1}{2}\right)-\frac{1}{4} \log _{2}\left(\frac{1}{4}\right)-\frac{1}{4} \log _{2}\left(\frac{1}{4}\right)$

$$
\begin{aligned}
& =\frac{1}{2} \log _{2} 2+\frac{1}{4} \log _{2} 4+\frac{1}{4} \log _{2} 4 \\
& =\frac{1}{2}+2 \frac{1}{4}+2 \frac{1}{4} \\
& =\frac{3}{2}
\end{aligned}
$$

The general case

- Suppose we pick between $x_{1}, x_{2}, \ldots, x_{n}$.
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Then the number of questions needed to find $k$ is the base two logarithm of $1 / p_{k}$,

$$
\text { Num. bits needed to find } \begin{aligned}
x_{k} & =\log _{2}\left(\frac{1}{p_{k}}\right) \\
& =-\log _{2} p_{k}
\end{aligned}
$$

## Conclusion

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- The entropy of a random variable is simply the average bits of information needed to guess its value successfully.
- Even though the formula might seem complicated, its meaning is simple.
- Please see [Shannon '48] for the original formulations.


## References

- Claude Shannon

A Mathematical Theory of Communication
1948, Bell Labs Journal


[^0]:    ${ }^{1} @$ TivadarDanka provided this thought experience.

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