# ILLC Project Course in Statistical Learning Theory 

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## Information Theory

## Problem

We are provided with a communication channel for binary information which tends to transmit as much as 3 in every 1000 bits wrongly ( 1 as 0 , or 0 as 1 ).

If we feed a sequence of 1000 bits into this
 channel, how many different sequences might come out?


## Information Theory

$$
\begin{aligned}
& \Phi(k, t)=\binom{t}{0}+\binom{t}{1}+\binom{t}{2}+\cdots+\binom{t}{k}
\end{aligned}
$$

## Information Theory

## Problem

We are provided with a communication channel which transmits each individual bit wrongly with probability 0.003 .

If we feed a sequence of 1000 bits into this channel, how many "reasonably likely"
 output sequences are there?
$000101110101 \ldots \longrightarrow$ Channel $\longrightarrow 01101110001 \ldots$

## Information Theory

## Problem

With the point probabilities

| $x$ | t | s | e |
| :---: | :---: | :---: | :---: |
| $p(x)$ | .25 | .50 | .25 |

1. What is $\operatorname{Pr}$ (stetsesses)?
2. What's the most probable sequence?

## Information Theory



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## Definition

The entropy of a random variable $X$ is

$$
H=E\left[\log \frac{1}{p(X)}\right]=-E[\log p(X)]
$$

## Definition

An $\varepsilon$-typical sequence of length $t$ is a sequence for which

$$
\left|\log \frac{1}{p\left(x_{1}, x_{2}, \ldots, x_{t}\right)}-H t\right|<\varepsilon
$$

## Information Theory

## The Weak Law of Large Numbers

For every $\varepsilon>0$ and $\alpha>0$ there is a $t$ such that

$$
\operatorname{Pr}\left\{\left|\frac{\sum_{i=1}^{t} X_{i}}{t}-E[X]\right|>\varepsilon\right\} \leq \alpha .
$$

## The Asymptotic Equipartition Property

Eventually, everything has the same probability.
The Source Coding Theorem
For large $t$, there are only $2^{H t}$ sequences worth caring about.

## Information Theory



## Information Theory

$$
\exp \left(H_{2}(q)\right)=\left(\frac{1}{q}\right)^{q}\left(\frac{1}{1-q}\right)^{1-q}
$$



## Information Theory

$$
\exp (D(q \| p))=\left(\frac{p}{q}\right)^{q}\left(\frac{1-p}{1-q}\right)^{1-q} \leq \exp \left(-2(q-p)^{2}\right)
$$



