

Information Theory 101

Bits

You are watching a set of independent random samples of X

You see that X has four possible values

$P(X=A) = 1/4$	$P(X=B) = 1/4$	$P(X=C) = 1/4$	$P(X=D) = 1/4$
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So you might see: BAACBADCDADDDA...

You transmit data over a binary serial link. You can encode each reading with two bits (e.g. $A = 00$, $B = 01$, $C = 10$, $D = 11$)

0100001001001110110011111100...

Fewer Bits

Someone tells you that the probabilities are not equal

$P(X=A) = 1/2$	$P(X=B) = 1/4$	$P(X=C) = 1/8$	$P(X=D) = 1/8$
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It's possible...

...to invent a coding for your transmission that only uses 1.75 bits on average per symbol. How?

Fewer Bits

Someone tells you that the probabilities are not equal

$P(X=A) = 1/2$	$P(X=B) = 1/4$	$P(X=C) = 1/8$	$P(X=D) = 1/8$
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It's possible...

...to invent a coding for your transmission that only uses 1.75 bits on average per symbol. How?

A	0
B	10
C	110
D	111

(This is just one of several ways)

General Case

Suppose X can have one of m values... V_1, V_2, \dots, V_m

$P(X=V_1) = p_1$	$P(X=V_2) = p_2$	$P(X=V_m) = p_m$
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What's the smallest possible number of bits, on average, per symbol, needed to transmit a stream of symbols drawn from X 's distribution? It's

$$\begin{aligned} H(X) &= -p_1 \log_2 p_1 - p_2 \log_2 p_2 - \dots - p_m \log_2 p_m \\ &= -\sum_{j=1}^m p_j \log_2 p_j \end{aligned}$$

$H(X)$ = The entropy of X

- “High Entropy” means X is from a uniform (boring) distribution
- “Low Entropy” means X is from varied (peaks and valleys) distribution

Specific Conditional Entropy $H(Y|X=v)$

Suppose I'm trying to predict output Y and I have input X

X = College Major

Y = Likes "Gladiator"

Let's assume this reflects the true probabilities

E.G. From this data we estimate

X	Y
Math	Yes
History	No
CS	Yes
Math	No
Math	No
CS	Yes
History	No
Math	Yes

- $P(\text{LikeG} = \text{Yes}) = 0.5$
- $P(\text{Major} = \text{Math} \ \& \ \text{LikeG} = \text{No}) = 0.25$
- $P(\text{Major} = \text{Math}) = 0.5$
- $P(\text{LikeG} = \text{Yes} \mid \text{Major} = \text{History}) = 0$

Note:

- $H(X) = 1.5$
- $H(Y) = 1$

Specific Conditional Entropy $H(Y|X=v)$

X = College Major

Y = Likes "Gladiator"

Definition of Specific Conditional Entropy:

$H(Y | X=v)$ = The entropy of Y among only those records in which X has value v

X	Y
Math	Yes
History	No
CS	Yes
Math	No
Math	No
CS	Yes
History	No
Math	Yes

Specific Conditional Entropy $H(Y|X=v)$

X = College Major

Y = Likes "Gladiator"

X	Y
Math	Yes
History	No
CS	Yes
Math	No
Math	No
CS	Yes
History	No
Math	Yes

Definition of Specific Conditional Entropy:

$H(Y | X=v)$ = The entropy of Y among only those records in which X has value v

Example:

- $H(Y|X=Math) = 1$
- $H(Y|X=History) = 0$
- $H(Y|X=CS) = 0$

Conditional Entropy $H(Y|X)$

X = College Major

Y = Likes "Gladiator"

X	Y
Math	Yes
History	No
CS	Yes
Math	No
Math	No
CS	Yes
History	No
Math	Yes

Definition of Conditional Entropy:

$H(Y|X)$ = The average specific conditional entropy of Y

= if you choose a record at random what will be the conditional entropy of Y , conditioned on that row's value of X

= Expected number of bits to transmit Y if both sides will know the value of X

$$= \sum_j \text{Prob}(X=v_j) H(Y | X = v_j)$$

Conditional Entropy

X = College Major

Y = Likes "Gladiator"

Definition of Conditional Entropy:

$H(Y|X)$ = The average conditional entropy of Y

$$= \sum_j \text{Prob}(X=v_j) H(Y | X = v_j)$$

X	Y
Math	Yes
History	No
CS	Yes
Math	No
Math	No
CS	Yes
History	No
Math	Yes

Example:

v_j	$\text{Prob}(X=v_j)$	$H(Y X = v_j)$
Math	0.5	1
History	0.25	0
CS	0.25	0

$$H(Y|X) = 0.5 * 1 + 0.25 * 0 + 0.25 * 0 = 0.5$$

Information Gain

X = College Major

Y = Likes "Gladiator"

Definition of Information Gain:

$IG(Y|X)$ = I must transmit Y .

How many bits on average
would it save me if both ends of
the line knew X ?

$$IG(Y|X) = H(Y) - H(Y|X)$$

Example:

- $H(Y) = 1$
- $H(Y|X) = 0.5$
- Thus $IG(Y|X) = 1 - 0.5 = 0.5$

X	Y
Math	Yes
History	No
CS	Yes
Math	No
Math	No
CS	Yes
History	No
Math	Yes

Relative Information Gain

X = College Major

Y = Likes "Gladiator"

Definition of Relative Information Gain:

$RIG(Y|X)$ = I must transmit Y , what fraction of the bits on average would it save me if both ends of the line knew X ?

$$RIG(Y|X) = H(Y) - H(Y|X) / H(Y)$$

Example:

- $H(Y|X) = 0.5$
- $H(Y) = 1$
- *Thus $IG(Y|X) = (1 - 0.5)/1 = 0.5$*

X	Y
Math	Yes
History	No
CS	Yes
Math	No
Math	No
CS	Yes
History	No
Math	Yes