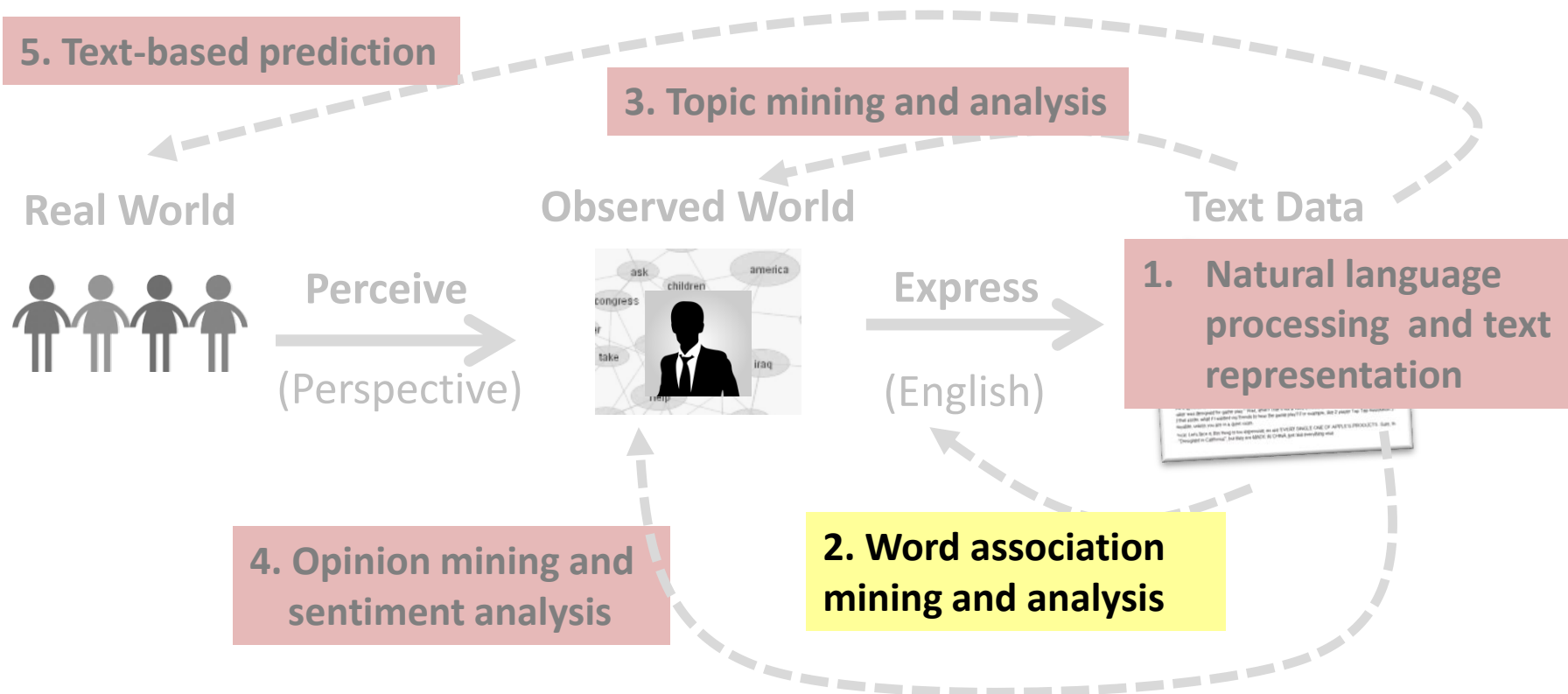


# Syntagmatic Relation Discovery: Entropy

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# Syntagmatic Relation Discovery: Entropy



# Syntagmatic Relation = Correlated Occurrences

Whenever “**eats**” occurs, what **other words** also tend to occur?

My cat **eats** fish on Saturday  
His cat **eats** turkey on Tuesday  
My dog **eats** meat on Sunday  
His dog **eats** turkey on Tuesday  
...

My	_____	<b>eats</b>	_____	on Saturday
His	_____	<b>eats</b>	_____	on Tuesday
My	_____	<b>eats</b>	_____	on Sunday
His	_____	<b>eats</b>	_____	on Tuesday
...	_____		_____	

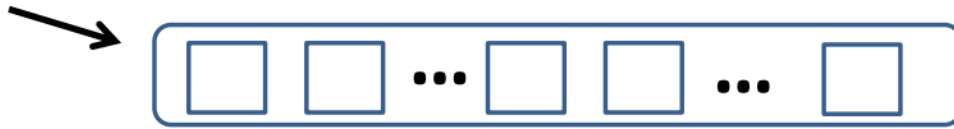
What words tend to occur  
to the **left** of “**eats**”?

What words  
are to the  
**right**?

# Word Prediction: Intuition

Prediction Question: Is word **W** present (or absent) in this segment?

Text Segment (any unit, e.g., sentence, paragraph, document)



Are some words easier to predict than others?

1)  $W = \text{"meat"}$

*medium*

2)  $W = \text{"the"}$

*frequent*

3)  $W = \text{"unicorn"}$

*rare*

# Word Prediction: Formal Definition

Binary Random Variable :  $X_w \in \{0, 1\}$   $X_w = \begin{cases} 1 & \text{w is present 存在} \\ 0 & \text{w is absent 不存在.} \end{cases}$

$$p(X_w = 1) + p(X_w = 0) = 1$$

*$X_w$  越随机, 则预测越困难.*

The more random  $X_w$  is, the more difficult the prediction would be.

How does one quantitatively measure the “randomness” of a random variable like  $X_w$ ?

# ★ Entropy $H(X)$ Measures Randomness of $X$

熵描述了随机程度。

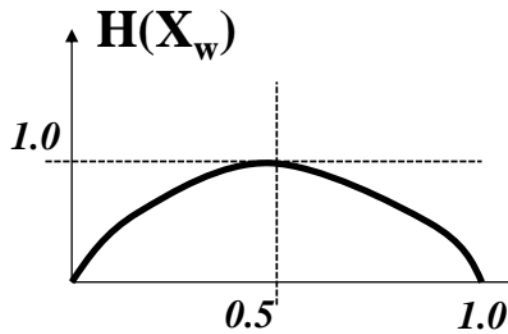
$$H(X_w) = \sum_{v \in \{0,1\}} -p(X_w = v) \log_2 p(X_w = v)$$

$$X_w = \begin{cases} 1 & \text{w is present} \\ 0 & \text{w is absent} \end{cases}$$

展开

$$= -p(X_w = 0) \log_2 p(X_w = 0) - p(X_w = 1) \log_2 p(X_w = 1)$$

Define  $0 \log_2 0 = 0$



For what  $X_w$ , does  $H(X_w)$  reach **maximum/minimum**?

E.g.,  $P(X_w=1)=1$ ?  $P(X_w=1)=0.5$ ?

or equivalently  $P(X_w=0)$  (Why?)

# Entropy $H(X)$ : Coin Tossing

$$H(X_{\text{coin}}) = -p(X_{\text{coin}} = 0) \log_2 p(X_{\text{coin}} = 0) - p(X_{\text{coin}} = 1) \log_2 p(X_{\text{coin}} = 1)$$

$X_{\text{coin}}$ : tossing a coin

$$X_{\text{coin}} = \begin{cases} 1 & \text{Head} \\ 0 & \text{Tail} \end{cases}$$

**Fair coin:  $p(X=1)=p(X=0)=1/2$**

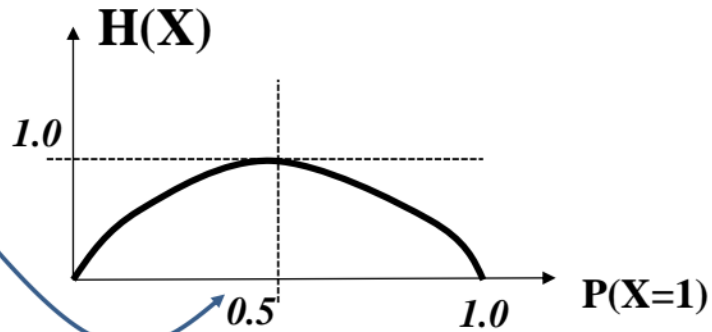
(此时熵最大, 随机程度最大, 但难预测.)

$$H(X) = -\frac{1}{2} \log_2 \frac{1}{2} - \frac{1}{2} \log_2 \frac{1}{2} = 1$$

**Completely biased:  $p(X=1)=1$**

(熵最小, 随机程度最小, 最容易预测.)

$$H(X) = -0 * \log_2 0 - 1 * \log_2 1 = 0$$



# Entropy for Word Prediction

Is word **W** present (or absent) in this segment?

☐☐ ... ☐☐ ... ☐

1)  $W = \text{"meat"}$

2)  $W = \text{"the"}$

3)  $W = \text{"unicorn"}$

Which is **high/low**?  $H(X_{\text{meat}})$ ,  $H(X_{\text{the}})$ , or  $H(X_{\text{unicorn}})$ ?

the 几乎很容易发现.

unicorn 熵很高.

$H(X_{\text{the}}) \approx 0 \rightarrow$  no uncertainty since  $p(X_{\text{the}}=1) \approx 1$

**High entropy words are harder to predict!**