

Natural Language Processing Applications

Week 6: Text Similarity



fit@hcmus

KHOA CÔNG NGHỆ THÔNG TIN
TRƯỜNG ĐẠI HỌC KHOA HỌC TỰ NHIÊN

- ❑ Introduction to Similarity
- ❑ Text Similarity
- ❑ Similarity Evaluation



Applied Natural Language Processing- Text Similarity

INTRODUCTION TO SIMILARITY



Introduction to Similarity

- ❑ Word similarity: finding similarities between words is a fundamental part of text similarity
- ❑ Words are considered similar if they :
 - ❑ Have the same meaning(Synonyms)
 - ❑ Have opposite meanings(Antonyms)
 - ❑ Are used in the same way(For example: red, green...)
 - ❑ Are used in the same context(For example: doctor, hospital...)
 - ❑ Are type of another word (For example: fluffy dog, dog, animal...)



Introduction to Similarity(continued)

❑ Text Similarity:

Hurricane Gilbert swept toward the Dominican Republic Sunday , and the Civil Defense alerted its heavily populated south coast to prepare for high **winds**, heavy **rains** and high seas.

The **storm** was approaching from the southeast with sustained **winds** of 75 mph gusting to 92 mph .

"There is no need for alarm," Civil Defense Director Eugenio Cabral said in a television alert shortly before midnight Saturday .

Cabral said residents of the province of Barahona should closely follow **Gilbert** 's movement .

An estimated 100,000 people live in the province, including 70,000 in the city of Barahona , about 125 miles west of Santo Domingo .

Tropical **Storm Gilbert** formed in the eastern Caribbean and strengthened into a **hurricane** Saturday night

The National **Hurricane** Center in Miami reported its position at 2a.m. Sunday at latitude 16.1 north , longitude 67.5 west, about 140 miles south of Ponce, Puerto Rico, and 200 miles southeast of Santo Domingo.

The National Weather Service in San Juan , Puerto Rico , said **Gilbert** was moving westward at 15 mph with a "broad area of cloudiness and heavy weather" rotating around the center of the **storm**.

The weather service issued a flash flood watch for Puerto Rico and the Virgin Islands until at least 6p.m. Sunday.

Strong **winds** associated with the **Gilbert** brought coastal flooding , strong southeast **winds** and up to 12 feet to Puerto Rico 's south coast.

Introduction to Similary(continued)

❑ Text Similary:

❑ Document1

- ❑ Gilbert: 3
- ❑ Hurricane: 2
- ❑ Rains: 1
- ❑ Storm: 2
- ❑ Winds: 2

❑ Document2

- ❑ Gilbert: 2
- ❑ Hurricane: 1
- ❑ Rains: 0
- ❑ Storm: 1
- ❑ Winds: 2

Cosine similarity: 0.9439



Introduction to Similary(continued)

- ❑ According to John Philip McCrae:
 - ❑ “**Semantic textual similarity** is the task of deciding if **two sentences** express a **similar or identical meaning** and requires a **deep understanding of a sentence and its meaning** in order to achieve high performance.”



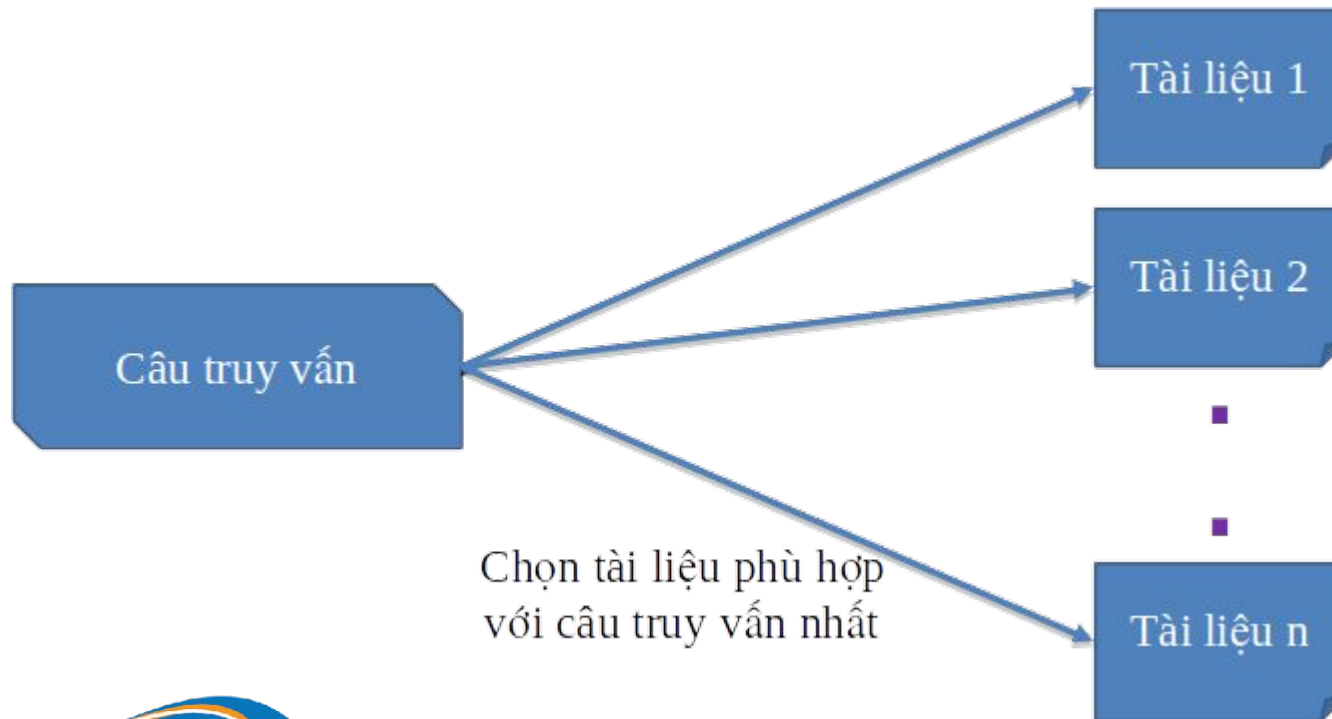
Introduction to Similarity(continued)

- ❑ Applications of Text Similarity:
 - ❑ Information Retrieval
 - ❑ Text Summarization
 - ❑ Machine Translation
 - ❑ Plagiarism Detection
 - ❑ ...



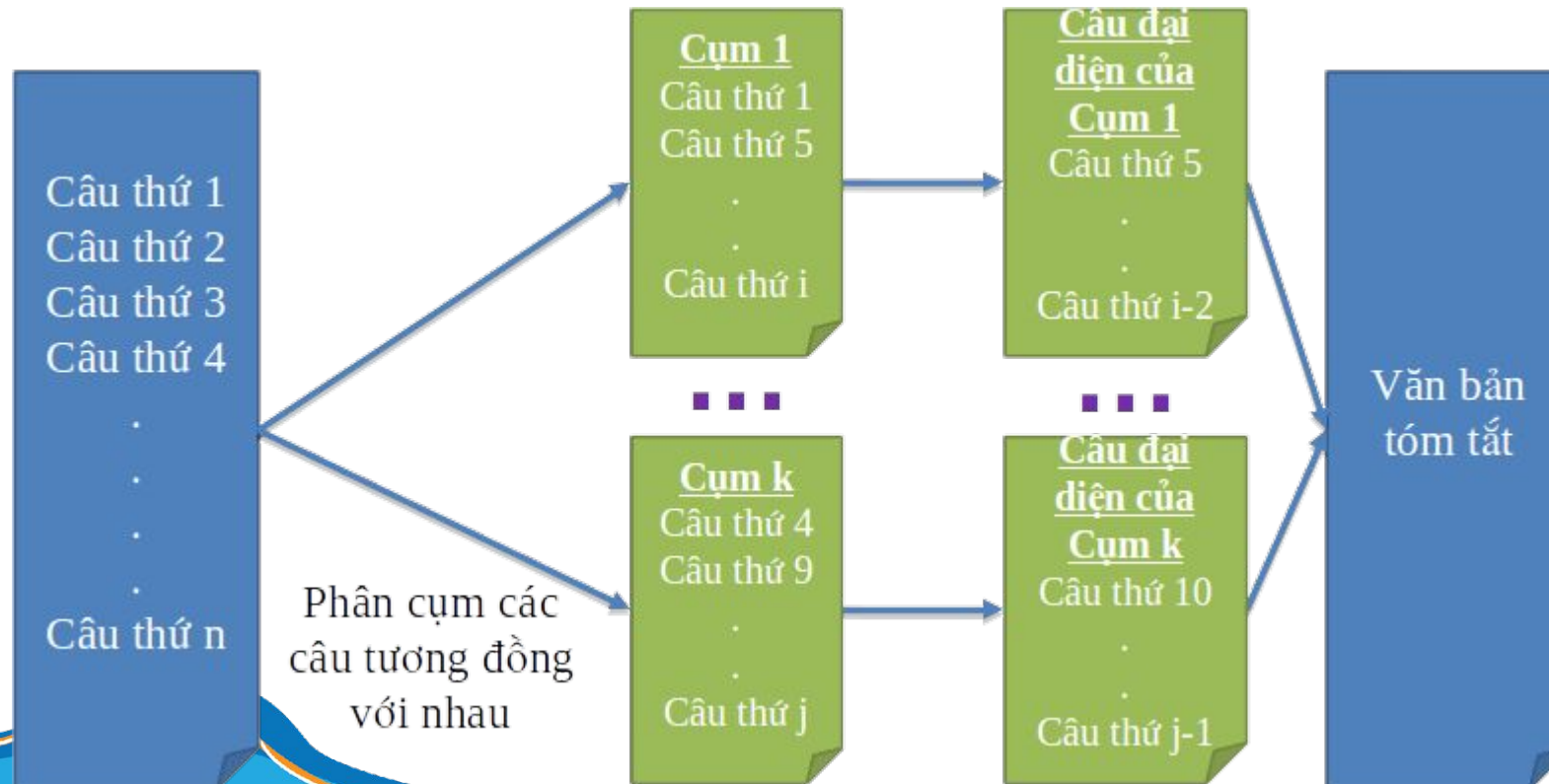
Introduction to Similarity(continued)

❑ Information Retrieval



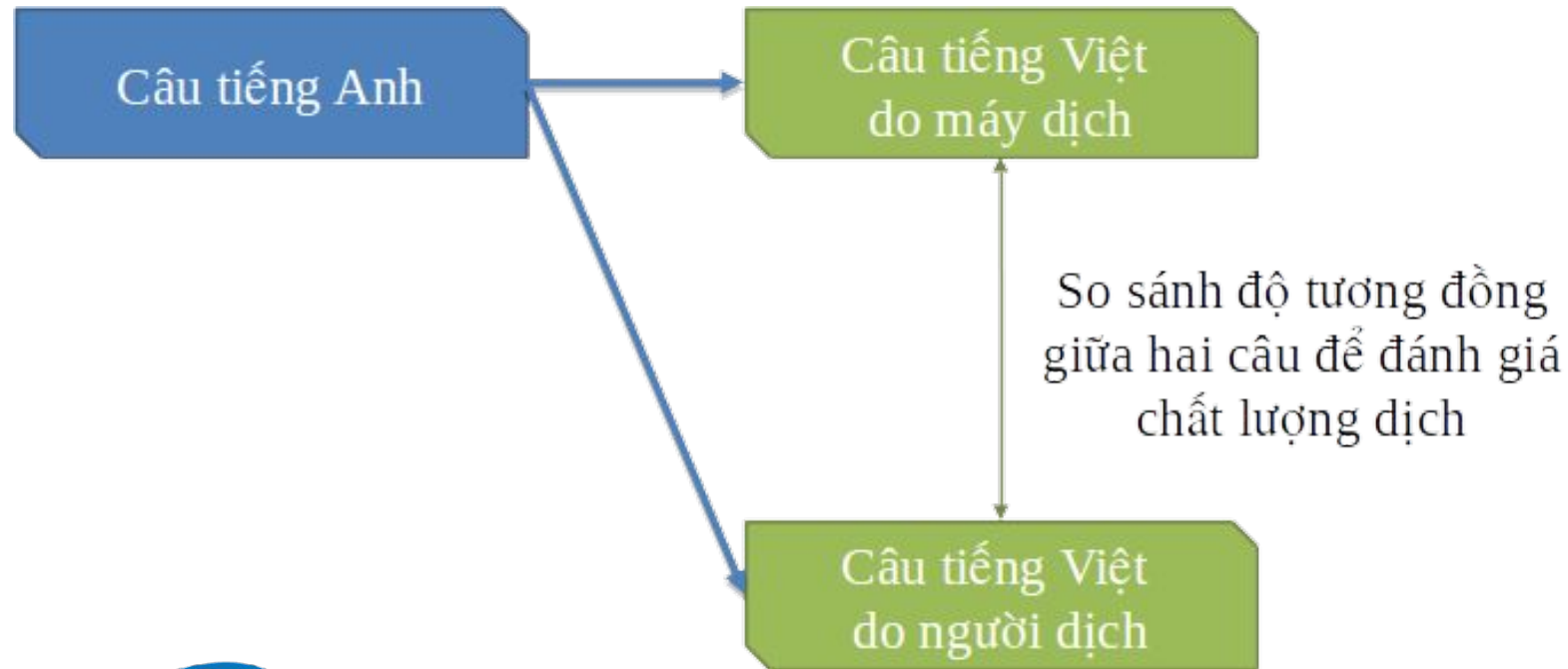
Introduction to Similary(continued)

Text Summarization



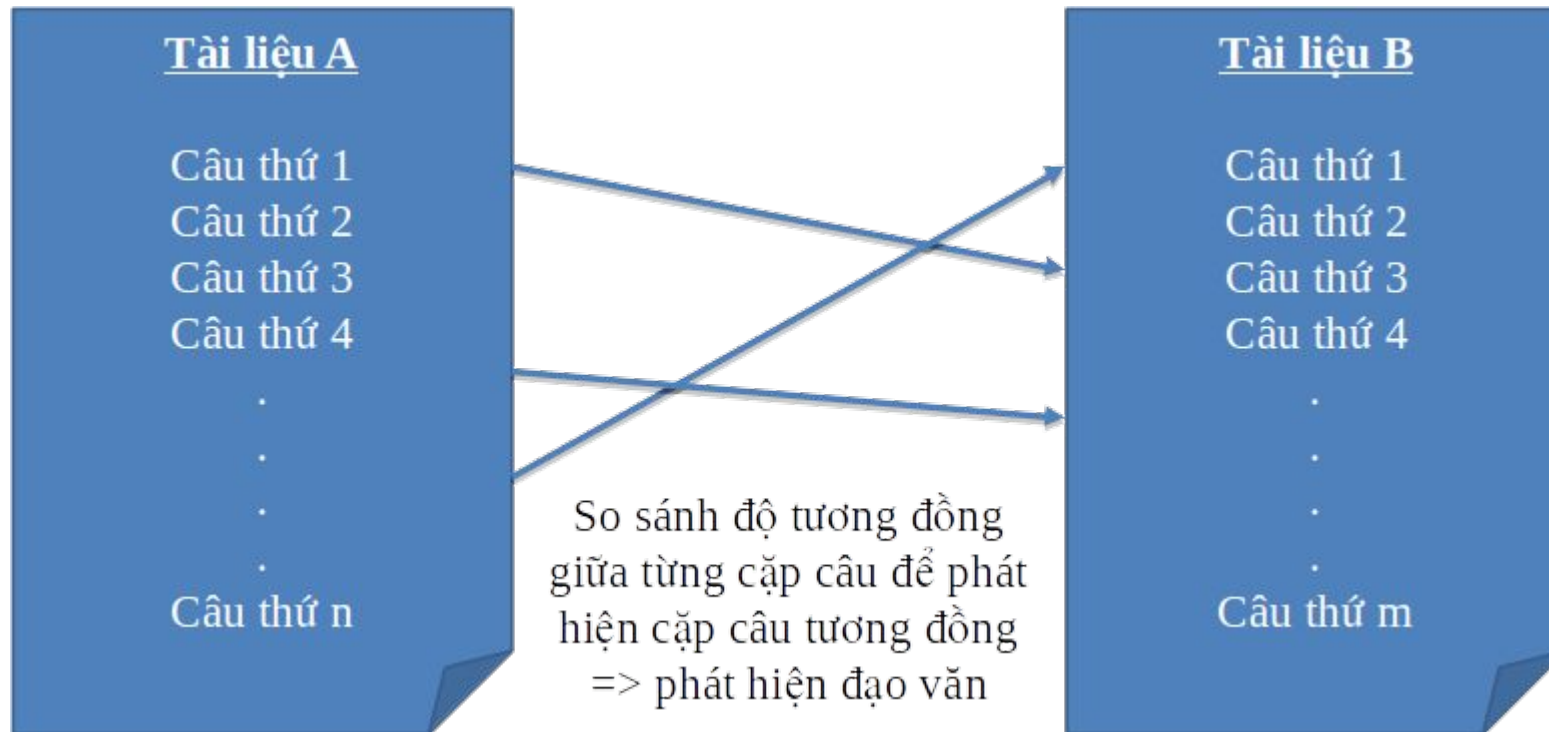
Introduction to Similarity(continued)

❑ Machine Translation (Quality evaluation)



Introduction to Similarity(continued)

❑ Plagiarism Detection



Introduction to Similary(continued)

- ❑ There are 2 popular scales :
 - ❑ **Scale 0 to 1** is used in detecting rewrites text while retaining full meaning(paraphrase identification)
 - ❑ **Scale 0 to 5:**



Introduction to Similary(continued)

Scale 0 to 1:

- ❑ The following pair of sentences is labeled as 1
 - ❑ Sentence 1: Customers will have to use a decoder card from the cable TV provider to plug in the set.
 - ❑ Sentence 2: To watch pay TV, customers will plug in the television set a decoder card provided by the cable TV provider.
- ❑ And the following pair of sentences is labeled as 0
 - ❑ Sentence 1: With an interpreter like you everything will be fine.
 - ❑ Sentence 2: Today's interpreter is Mr. Nam.

Introduction to Similary(continued)

□ Scale 0 to 5:

5	Hai câu tương đồng hoàn toàn Con chim đang tắm trong bồn rửa. Con chim non đang tắm trong bồn nước.
4	Hai câu tương đồng phần lớn, nhưng khác nhau vài chi tiết không quan trọng. Hai chàng trai đang chơi trò chơi điện tử trên một chiếc ghế dài. Hai chàng trai đang chơi trò chơi điện tử.
3	Hai câu gần tương đồng, nhưng khác nhau hoặc thiếu một vài thông tin quan trọng. John cho biết anh ấy được xem là một nhân chứng chứ không phải là một nghi phạm. “Anh ấy không phải là kẻ tình nghi nữa.” John nói.
2	Hai câu không tương đồng, nhưng có chung một vài thông tin. Chúng bay ra khỏi tổ theo từng nhóm. Chúng cùng bay vào tổ.
1	Hai câu không tương đồng, nhưng có cùng chung chủ đề. Người phụ nữ đang chơi đàn vĩ cầm. Người phụ nữ trẻ thích nghe đàn ghita.
0	Hai câu hoàn toàn khác nhau. Con chó đang chạy trên tuyết. Một người lái xe đua đang lái xe của mình qua bãi bùn.

Applied Natural Language Processing- Text Similarity

TEXT SIMILARITY METHODS



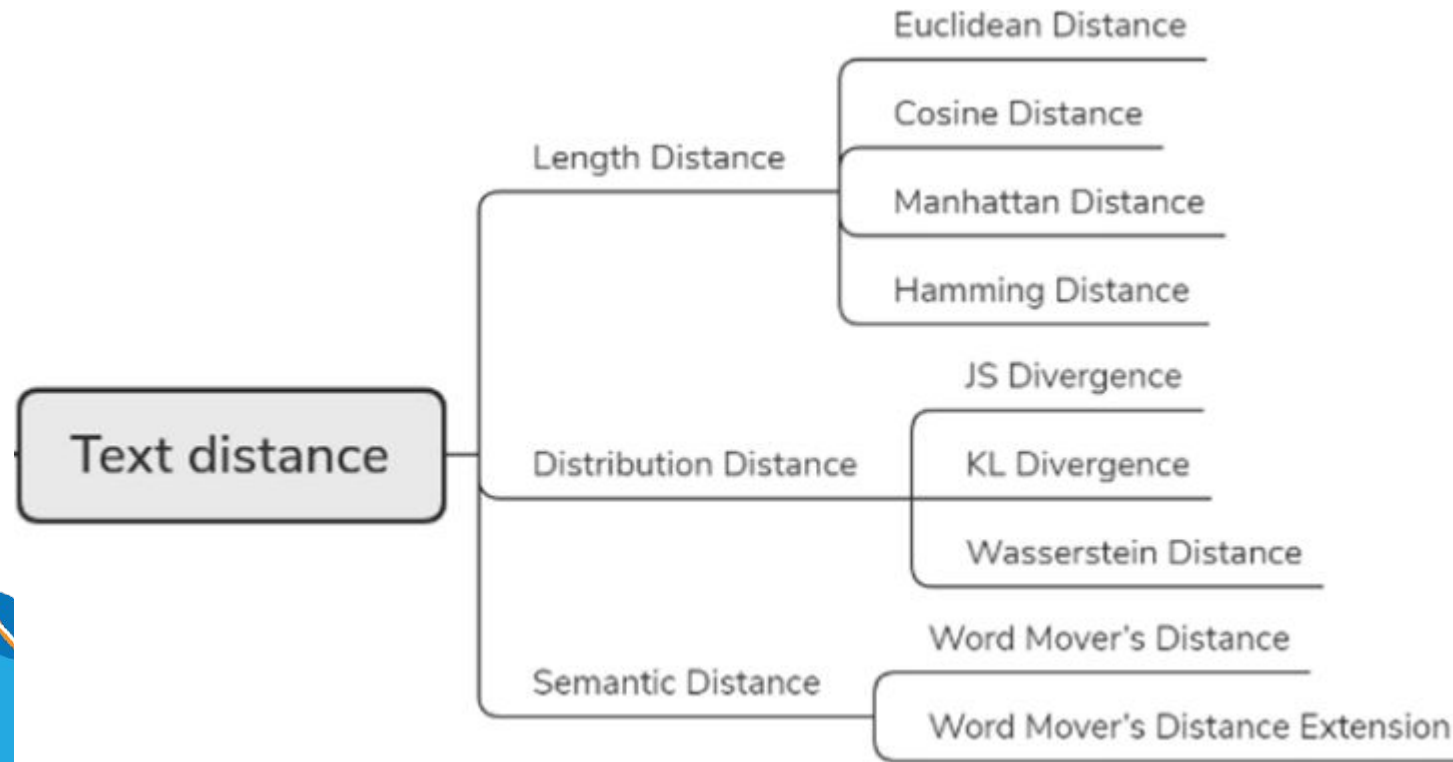
Text Similarity methods

- ❑ Text Distance
- ❑ Text Representation



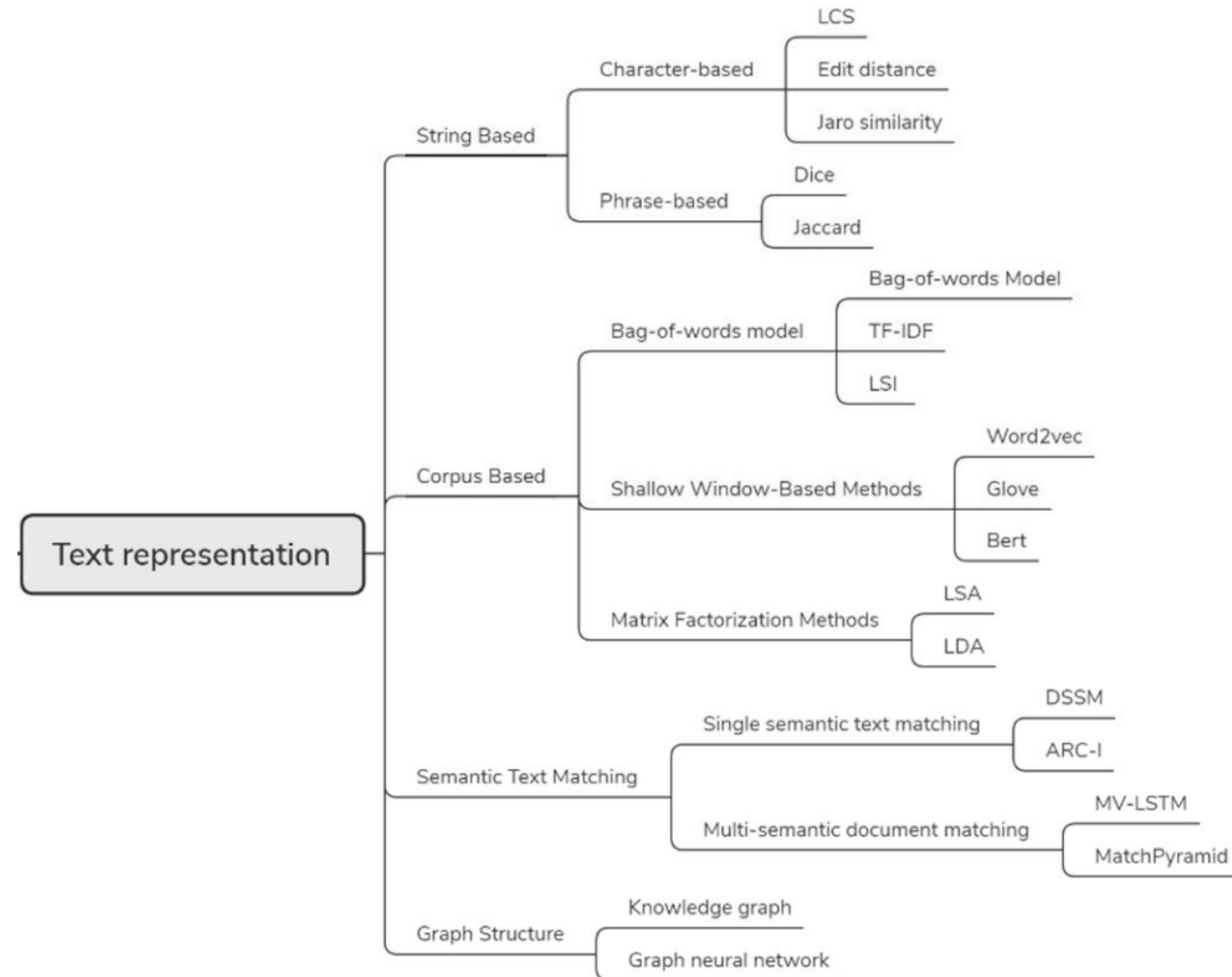
Text Similarity methods

- ❑ Text distance: describes the semantic proximity of two text words from the perspective of distance



Text Similarity methods

- ❑ Text representation:
numerically represent the
unstructured text
documents to make them
mathematically computable.



Text Distance

- ❑ Main methods:
 - ❑ Length Distance
 - ❑ Distribution Distance
 - ❑ Semantic Distance



Text Distance

- Length Distance:

- Euclidean Distance

$$d(S_a, S_b) = \sqrt{\sum_{i=1}^n (S_a^{(i)} - S_b^{(i)})^2}$$

- Cosine Distance

$$\text{Sim}(S_a, S_b) = \cos \Theta = \frac{\vec{S}_a \cdot \vec{S}_b}{\|S_a\| \cdot \|S_b\|}$$

- Manhattan Distance

$$\text{Sim}(x, y) = |x_1 - x_2| + |y_1 - y_2|$$



Text Distance

- ❑ Length Distance:
 - ❑ Suitable for symmetrical problems
 - ❑ $\text{Sim}(A, B) = \text{Sim}(B, A)$
 - ❑ \Rightarrow But for question Q to retrieve answer A, the corresponding similarity is not symmetrical.
 - ❑ Lack of statistical characteristics of the data



Text Distance

- ❑ Distribution Distance:
 - ❑ Kullback–Leibler Divergence
 - ❑ Jensen–Shannon Divergence
 - ❑ Wasserstein Distance

$$d(p||q) = \sum_{i=1}^n p(x) \log \frac{p(x)}{q(x)}$$

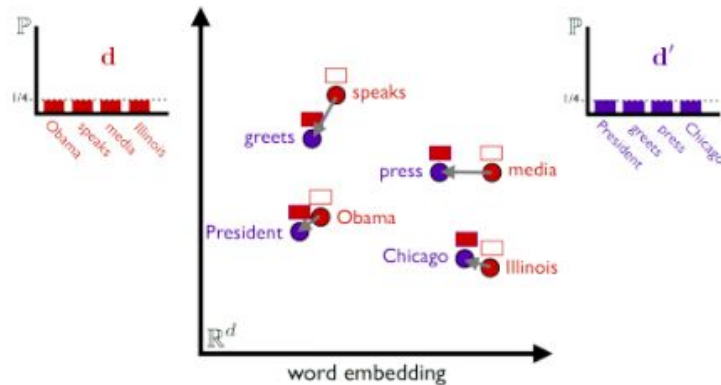
$$JS(P_1||P_2) = \frac{1}{2}KL(P_1||\frac{P_1+P_2}{2}) + \frac{1}{2}KL(P_2||\frac{P_1+P_2}{2})$$

$$W(p_r, p_g) = \inf_{\gamma \sim \Pi(p_r, p_g)} E_{(x,y) \sim \gamma} [\|x - y\|]$$



Text Distance

- ❑ Semantic Distance :
- ❑ Word Mover's Distance



- ❑ Word Mover's Distance Extension
- ❑ Use the Mahalanobis distance instead of the Euclidean distance

Text representation

- ❑ String-based:
 - ❑ Operate on string sequences and character composition
 - ❑ Includes 2 methods:
 - ❑ Character-based
 - ❑ Phrase-based



Text representation

- ❑ String based:

- ❑ Longest common substring:

$$\text{LCS}(S_a, S_b) = \begin{cases} 0, & \text{if } S_a = 0 \text{ or } S_b = 0 \\ 1 + \text{LCS}(S_a - 1, S_b - 1), & \text{if } x[S_a] == y[S_b] \\ \max \begin{cases} \text{LCS}(S_a, S_b - 1) \\ \text{LCS}(S_a - 1, S_b) \end{cases} & \text{if } x[S_a] \neq y[S_b] \end{cases}$$

- ❑ Jaro Similarity:

$$\text{Sim} = \begin{cases} 0, & \text{if } m = 0 \\ \frac{1}{3} \left(\frac{m}{|S_a|} + \frac{m}{|S_b|} + \frac{m-t}{m} \right) \end{cases}$$



Text representation

- ❑ String based:
 - ❑ Edit Distance:
 - ❑ Levenshtein distance (L distance)
 - ❑ The minimum number of single-character edits (insertions, deletions or substitutions) required to change S1 into S2.
 - ❑ Damerau–Levenshtein Distance (D distance)
 - ❑ Like L distance, but with the addition of the transposition operation
 - ❑ Optimal String Alignment
 - ❑ Like D distance
 - ❑ No substring/subsequence is edited more than once.

Text representation

- ❑ Corpus based:
 - ❑ Use data from corpus
 - ❑ Textual feature
 - ❑ Co-occurrence probability
 - ❑ Includes 3 methods
 - ❑ Bag-of-words
 - ❑ Distributed representation
 - ❑ Matrix factorization

Text representation

❑ BOW:

- ❑ Count the number of times each word appears

*It was the best of times,
it was the worst of times,
it was the age of wisdom,
it was the age of foolishness*

- "it"
- "was"
- "the"
- "best"
- "of"
- "times"
- "worst"
- "age"
- "wisdom"
- "foolishness"

1	"it was the best of times"	= [1, 1, 1, 1, 1, 1, 0, 0, 0, 0]
2	"it was the worst of times"	= [1, 1, 1, 0, 1, 1, 1, 0, 0, 0]
3	"it was the age of wisdom"	= [1, 1, 1, 0, 1, 0, 0, 1, 1, 0]
4	"it was the age of foolishness"	= [1, 1, 1, 0, 1, 0, 0, 1, 0, 1]



Text representation

- ❑ Term frequency–inverse document frequency (TF - IDF):
 - ❑ To measure how important a word is to a document in a collection (or corpus) of documents
 - ❑ TF: the ratio of a word's occurrence in a document

$$tf(w, d) = Freq(w, d)$$

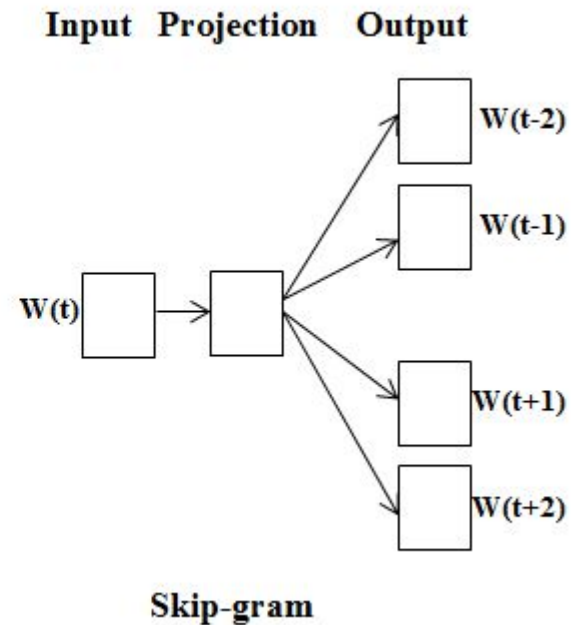
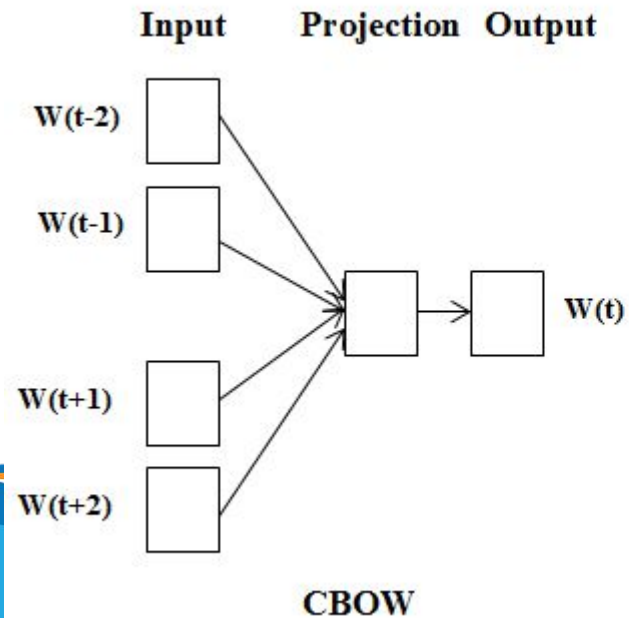
- ❑ IDF: indicates the amount of information provided

$$idf(w, D) = \log \frac{|D|}{N(w)}$$

$$tf-idf(w, d, D) = tf(w, d) \times idf(w, D)$$

Text representation

- ❑ word2vec:
 - ❑ Continuous Bag-of-words (BOW)
 - ❑ Word-skip grams (skip-gram)



Text representation

- ❑ Glove:
 - ❑ Words with similar meanings tend to appear in similar contexts
 - ❑ Encode the ratios of co-occurrence probabilities with vector differences
- ❑ Bert: Bidirectional Encoder Representations from Transformers
 - ❑ Already pre-trained on massive datasets



Corpus

- ❑ STS Benchmark (STSb)
 - ❑ Includes 8,628 sentence pairs :
 - ❑ Train: 5,749
 - ❑ Develop: 1,500
 - ❑ Test: 1,379
 - ❑ Three categories : captions, news, and forums



Corpus

Dataset Name	Sentence pairs	Similarity score range	Year
LiSent	65	0 - 4	2007
SRS	30	0 - 4	2007
STS2012	5250	0 - 5	2012
STS2013	2250	0 - 5	2013
STS2014	3750	0 - 5	2014
SICK	10000	1 - 5	2014
STS2015	3000	0 - 5	2015
STS2016	1186	0 - 5	2016
STS2017	1750	0 - 5	2017

Applied Natural Language Processing- Text Similarity

EVALUATING



Similarity Evaluation

- ❑ Pearson correlation coefficient:

$$r = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2 \sum_{i=1}^n (y_i - \bar{y})^2}}$$

