## 13. Searching & Midterm review

LING-351 Language Technology and LLMs

Instructor: Hakyung Sung

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\*Acknowledgment: These course slides are based on materials by Lelia Glass @ Georgia Tech (Course: Language & Computers)

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# Searching

Information need: What happened to my friend *Katie Smith* from elementary school? How to find out?

• Google search? (What queries to use?)

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Information need: What happened to my friend *Katie Smith* from elementary school? How to find out?

- Google search? (What queries to use?)
- · LinkedIn, Facebook, Instagram?
- (Probably not super helpful to ask ChatGPT...)
- Ask a mutual friend?

We need to understand that users interacting with technology have various intents.

• Set an alarm for tomorrow at 6.

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- Set an alarm for tomorrow at 6.
- · Book a flight to Guatemala.

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- Set an alarm for tomorrow at 6.
- Book a flight to Guatemala.
- · Intent recognition: identify user's intent from what they say
  - closely related to text classification problem

**Information needs** are the user's underlying intentions to find out or learn something. For example, if someone asks:

- What is the capital of Guatemala?
- · What's the weather there?
- · When did Mandarin and Cantonese diverge?
- · What happened to my friend Katie Smith?

all of these are information needs, because the user's goal is to know something

When these needs are turned into search queries, they often appear in a shorter or less natural form. Let's go back to the previous examples)

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· The information need is the question in the user's mind

When these needs are turned into search queries, they often appear in a shorter or less natural form. Let's go back to the previous examples)

- The information need is the question in the user's mind
- The query is the way they actually type it into a search engine

How to quantify success? Some common metrics.

- Precision
- Recall

**Precision**: Percentage of the documents returned that are relevant.

• Search engine gives you 400 documents, 200 of them are actually relevant: precision is 200/400 = 50%.

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- Search engine gives you 400 documents, 200 of them are actually relevant: precision is 200/400 = 50%.
- Aka positive predictive value (medicine!): what percent of positives are true positives?

**Recall**: Percentage of relevant documents that are returned.

• There are 1000 relevant documents out there; search engine gives 200 of them: precision is 200/1000 = 20%.

## **Evaluating search results: Summary**

$$\textbf{Precision} = \frac{\text{Number of relevant documents retrieved}}{\text{Total number of documents retrieved}}$$

$$\textbf{Recall} = \frac{\text{Number of relevant documents retrieved}}{\text{Total number of relevant documents in the collection}}$$

## **Evaluating search results: Summary**

$$\textbf{Precision} = \frac{\text{Number of relevant documents retrieved}}{\text{Total number of documents retrieved}}$$

$$\textbf{Recall} = \frac{\text{Number of relevant documents retrieved}}{\text{Total number of relevant documents in the collection}}$$

 $Precision \Rightarrow How \ accurate \ the \ results \ are$   $Recall \Rightarrow How \ complete \ the \ results \ are$ 

What would happen if a search engine aimed for 100%
 precision but didn't care about recall? → It would return very
 few (or even one) perfectly relevant documents.

- What would happen if a search engine aimed for 100%
   precision but didn't care about recall? → It would return very
   few (or even one) perfectly relevant documents.
- What would happen if a search engine aimed for 100% recall but didn't care about precision? → It would return everything, including all irrelevant documents.

**F-measure**: balances precision and recall (because both matter)

$$F_1 = \frac{2 \times P \times R}{P + R}$$

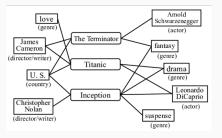
 In reality, search results are also ranked — people care most about the precision of the top-ranked results.

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- Precision@k: looks at the top k results and asks, what percent of these are relevant?

**Structured data** 

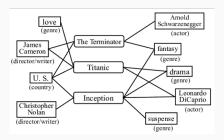
#### Structured data

• IMDB: links each film with year, director, producer, actor, etc.



#### Structured data

- **IMDB**: links each film with year, director, producer, actor, etc.
- Can visualize in a (limited-domain) knowledge graph (i.e., "ontology")



 Google Scholar: links each paper with year, author, who it cites, who cites it. (https://scholar.google.com/)

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- FrameNet: a lexical database of English that is both humanand machine-readable (https://framenet.icsi.berkeley.edu/)
- ConceptNet: a freely-available semantic network, designed to help computers understand the meanings of words that people use (https://conceptnet.io/)

Google's knowledge graph (https://www.youtube.com/watch?v=mmQ16VGvX-c)

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- Knowledge graph and searching (https://www.youtube.com/watch?v=Q5izD6Xlb8o)

**Midterm** 

- Start Date: October 9, 2025 9:00 AM
- End Date: October 9, 2025 5:00 PM
- Time Limit: 60 minutes
- Mode: Asynchronous (open within the given window)
- Attempt: 1
- Format: Open-book, but not Open-AI (PLEASE take this as an opportunity to review and apply what you've learned)

• Total 19 questions

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- Each class slide is related to 2-3 questions

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- Multiple choice (10 questions)

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- Short written response (8 questions)

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- Each class slide is related to 2-3 questions
- Multiple choice (10 questions)
- Short written response (8 questions)
- Longer written response (1 question)

• Introduction & Encoding

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  - What is language

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  - · What is language
  - Different type of languages (alphabetic, syllabic, logographic)

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  - · What is language
  - Different type of languages (alphabetic, syllabic, logographic)
  - language vs. writing (symbols)

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  - · What is language
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- · Spelling checkers

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  - · What is language
  - Different type of languages (alphabetic, syllabic, logographic)
  - language vs. writing (symbols)
- Spelling checkers
  - · Simple checkers

- · Introduction & Encoding
  - · What is language
  - Different type of languages (alphabetic, syllabic, logographic)
  - · language vs. writing (symbols)
- · Spelling checkers
  - · Simple checkers
  - More complex checkers

· Grammar checkers

- · Grammar checkers
  - Syntax

- · Grammar checkers
  - Syntax
  - POS

- · Grammar checkers
  - Syntax
  - POS
  - Dependency relations

- Grammar checkers
  - Syntax
  - POS
  - Dependency relations
- CALL

- · Grammar checkers
  - Syntax
  - POS
  - Dependency relations
- CALL
  - Language learning

- · Grammar checkers
  - Syntax
  - POS
  - Dependency relations
- CALL
  - · Language learning
  - L1 vs. L2

- · Grammar checkers
  - Syntax
  - POS
  - Dependency relations
- CALL
  - · Language learning
  - L1 vs. L2
  - transfer

- · Grammar checkers
  - Syntax
  - POS
  - Dependency relations
- CALL
  - · Language learning
  - L1 vs. L2
  - transfer
  - · social factors

• Text as data

- Text as data
  - knowledge-driven

- Text as data
  - · knowledge-driven
  - data-driven

- Text as data
  - · knowledge-driven
  - data-driven
  - (skip different previous studies)

- · Text as data
  - · knowledge-driven
  - data-driven
  - (skip different previous studies)
- Corpus, word distributions

- · Text as data
  - · knowledge-driven
  - data-driven
  - (skip different previous studies)
- · Corpus, word distributions
  - (skip English corpora)

- · Text as data
  - · knowledge-driven
  - data-driven
  - · (skip different previous studies)
- · Corpus, word distributions
  - (skip English corpora)
  - · Word distributions Zipf, Heap

- · Text as data
  - · knowledge-driven
  - · data-driven
  - · (skip different previous studies)
- · Corpus, word distributions
  - (skip English corpora)
  - · Word distributions Zipf, Heap
  - · Word vectors

• Text classification

- · Text classification
  - Different text classification tasks (e.g., spam filtering)

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  - How do we make text classifier?

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  - Different text classification tasks (e.g., spam filtering)
  - How do we make text classifier?
  - (skip the details of the perceptron)

# Group

### **Group assigned**

- Groups have been assigned and added to myCourses.
- Each group consists of 2–3 members.
- There will be two presentations per group:
  - One member presents in the first round.
  - Another member presents in the second round.
- We will discuss presentation preparation in more detail on October 16, after the fall break.
- For now, please stay in touch with your group members and start discussing your ideas.