

# Indexing

Presented By:  
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- [illegible]

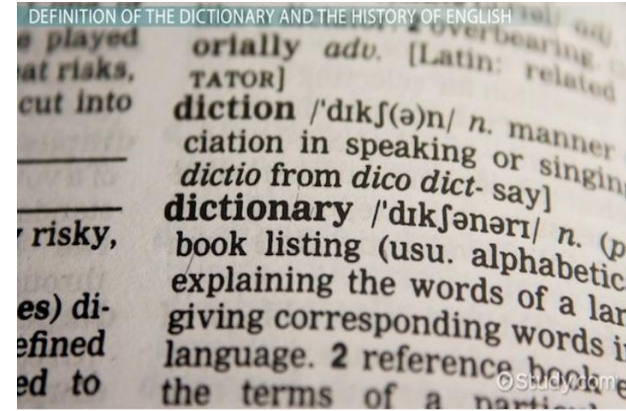
# Indexing

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- Problem with this approach?
- Reading each and every single document in a collection of documents is time and resource consuming.
- That is not how every search engine is implemented.
- Take into consideration an example of google search engine, it does not search in its database from start to end.

# Indexing

- So what is the solution for this problem?
- Building and using an **Index**
- Take an example of dictionary.
- Each and every word is listed in alphabetical order, page numbers are mentioned against a word.
- When you visit that page in dictionary, you would see it has different word forms such as alter, altered etc.



# Indexing

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Formal Definition of Indexing:

“A data structure technique that is used for quickly retrieving entries from database files using some attributes that have been indexed”.

# Assignment 1

Information Retrieval

Supervised By: Dr. Syed Khaldoon Khurshid

# Introduction

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- Build a document search engine.
- Use indexing to implement a simple search engine.

# Objectives

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- Goal:
  - Develop a basic search engine to search documents by title and content.
- Important Tasks:
  - Implement Indexing for fast lookup.
  - Implement scoring to setup important terms for indexing.



# Scope

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- Scope:
  - Build a simple, web based search engine.
  - Provide title and content based search.

# Assignment modules

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- Text Preprocessing
- Expansion with Synonyms
- Gather Data
- Term Frequency - Inverse Document Frequency Implementation
- Indexing
- Searching

# Pre-Processing

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Preprocessing is a crucial first step in building a search engine or any system dealing with large text data. It involves cleaning and organizing text to make it more "search-friendly."

Pre-Processing Techniques:

- Tokenization
- Lemmatization

# Pre-Processing (Continued...)

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## Tokenization:

Split text into individual words or "tokens" that can be indexed.

Lets understand with an example:

“Ali plays video games in evening”  $\Rightarrow$  [“Ali”, “plays”, “video”, “games”, “in”, “evening”]

# Pre-Processing (Continued...)

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## Lemmatization:

Reduce words to their root form.

Lets understand with few examples:

- “Running” ➡ “Run”
- “Children” ➡ “Child”
- “Better” ➡ “Good”

# Expansion with Synonyms

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Query expansion with synonyms helps in better searching in documents even if exact matches are not found.

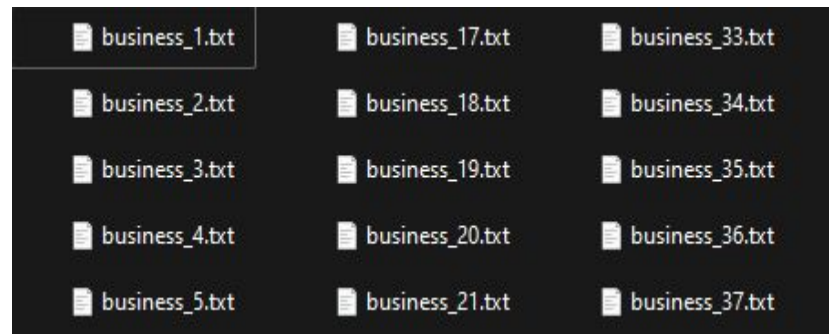
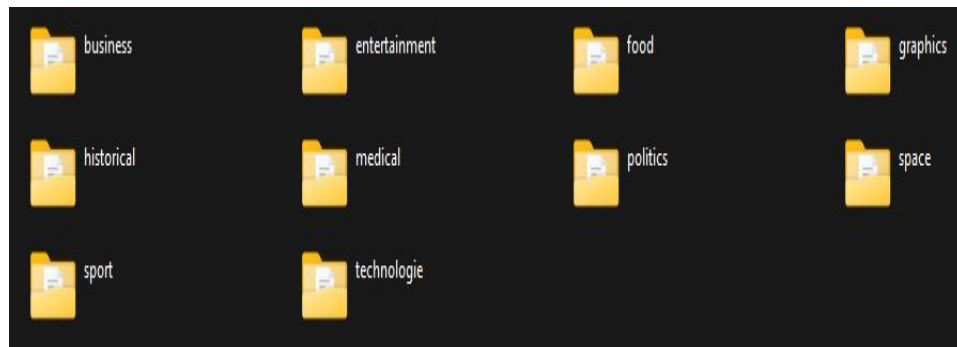
Practical Scenario:

1. **User Query:** “treatment”.
2. **Expanded Query:** “treatment, therapy, medication, cure”
3. **Result:** Documents with “treatment”, “therapy”, “medication”, “cure” will also be retrieved by search engine.

# Gather Data

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- Text files stored in a directory, containing different folders each representing a category.
- Each folder contains 100 text documents.



# Synonyms & Query Expansion

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- Retrieves synonyms for a given word from WordNet.
- Expands search queries to increase matches, even if the exact words is not used in the document.
- Example:
  - Search for “hope” includes results for synonyms like “aspiration” and “wish” etc.



# TF-IDF Implementation

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- TF (Term Frequency):
  - Measures how often a term appears in a document relative to total words.
  - Formula:
    - $\text{Term} = \text{count of Term} / \text{total words in document}.$

# TF-IDF Implementation (Continued...)

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- TF(Example):
  - Suppose you have  $X$  number of coins, and you want to rank similar coins in ascending orders.
  - Coin  $Y$  appears  $Z$  times in the coin set, so its term frequency would be:
    - $\text{Coins}[Y] = Z \text{ (count of } Y \text{ coins)} / X \text{ (total coins)}$



# TF-IDF Implementation (Continued...)

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- IDF (Inverse Document Frequency):
  - Measures term importance across documents by giving higher scores to terms with fewer appearance in dataset.
  - Formula:
    - $\text{Term} = \log(\text{Total Documents} / (1 + \text{Documents Containing Term}))$ .

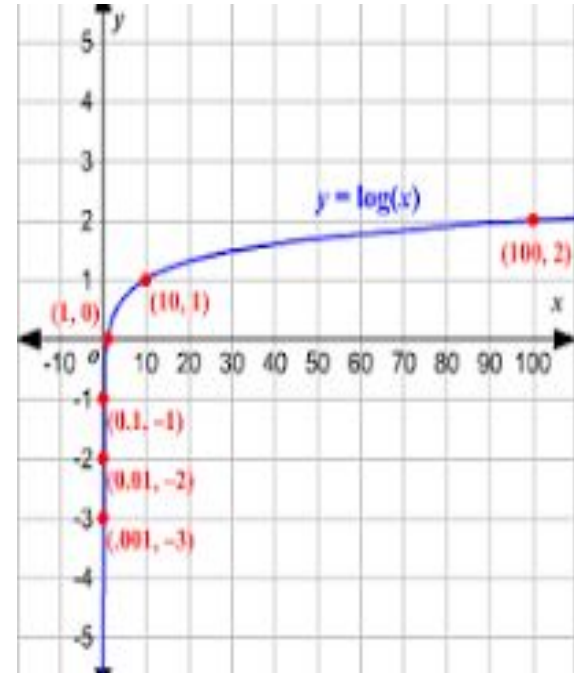
# TF-IDF Implementation (Continued...)

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- IDF (Formula Explanation):
  - **Total Documents:** Total documents in corpus.
  - **Documents Containing Term:** Simply it is count of terms in the complete corpus.
  - **Adding 1:** To avoid division by 0.
  - **Logarithmic Function:** To penalize common terms and reward rare terms.

# TF-IDF Implementation (Continued...)

- IDF (Logarithm Function):
  - It penalizes higher values more and lower values less.
  - Example:
    - $\log(100) = 2$
    - $\log(10) = 1$



# TF-IDF Implementation (Continued...)

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- IDF (Calculated Example):
  - Consider two words “machine” and “data”.
  - Documents with “machine” = 9.
  - Documents with “data” = 499.
  - $\text{IDF}[\text{machine}] = \log(1000 / (1 + 9)) \Rightarrow \log(100) \Rightarrow 2$ .
  - $\text{IDF}[\text{data}] = \log(1000 / (1 + 499)) \Rightarrow \log(2) \Rightarrow 0.3$ .

So, word “machine” is more relevant to be retrieved because of its higher relevancy score.

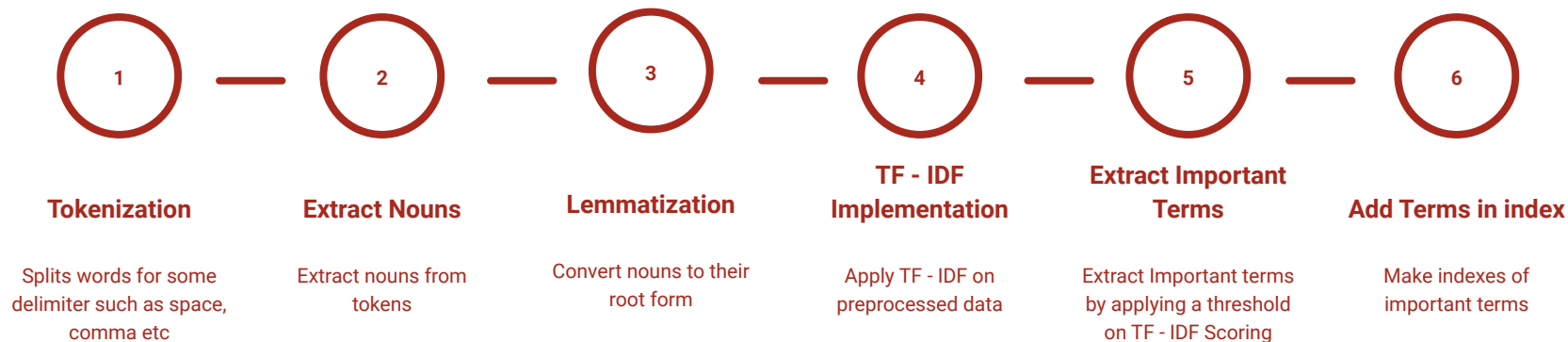
# TF-IDF Implementation (Continued...)

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- TF-IDF (Term Frequency - Inverse Document Frequency):
  - $TF\text{-}IDF = TF * IDF$

# Indexing

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# How Searching is Performed

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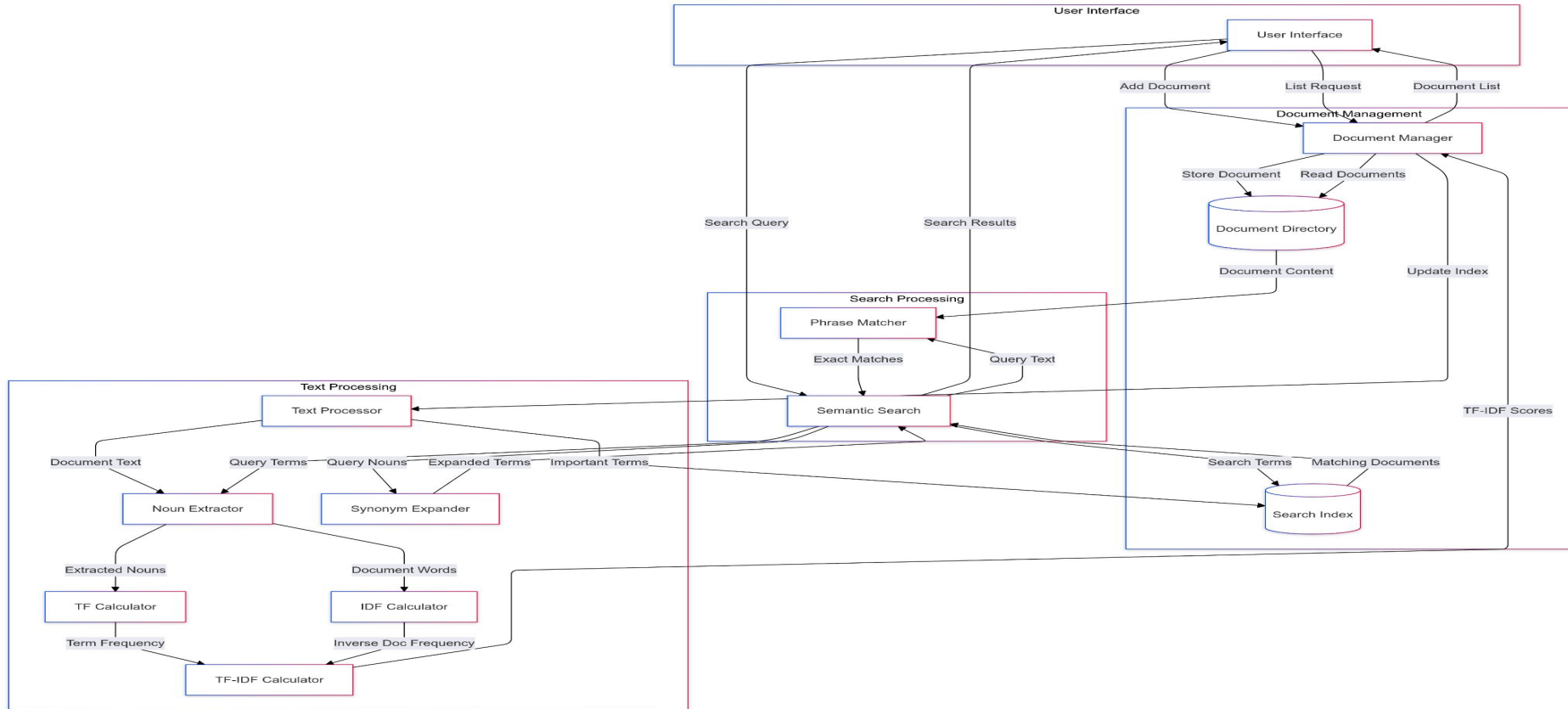
- Title-Based Search:
  - Checks if title contains the query term.
  - Returns exact matches for document titles.
- Content-Based Search:
  - Matches phrases directly or uses expanded terms with synonyms.
  - Looks up each term in the index to retrieve documents with matching terms.

# Summary & Conclusion

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- Summary:
  - Basic document search engine implemented with indexing and synonym expansion.
  - TF-IDF used for relevance scoring to improve search accuracy.
- Content-Based Search:
  - This assignment demonstrates foundational IR concepts with indexing and search.
  - Offers navigation for adding and retrieving documents.

# Data Flow Diagram



Thank You

Any Questions?