

# Entities

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## Named Entity Recognition (NER)

- This is the task of identifying and classifying named entities in a text.
  - Named entities are names of persons, organizations, locations, dates, etc.
  - This is typically approached as a sequence labeling task, where each word in a sentence is labeled with the type of named entity it represents.
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## Named Entity Recognition (NER)

- Task was first introduced at the Message Understanding Conference (MUC) in 1995.
  - Most versions identify proper nouns, such as names of people, places, and organizations.
  - Some also identify temporal expressions, such as dates and times.
  - Domain-specific versions exist such as GENIA for biomedical text, which has 36 classes!
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## CoNLL 2003 Dataset

- This dataset is a popular benchmark for NER.
  - It contains news articles from Reuters, and is annotated with named entities.
  - The entities are divided into four categories: persons, organizations, locations, and miscellaneous.
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## IOB Annotation

- In the CoNLL 2003 dataset, named entities are annotated using the IOB format.
    - I: Inside a named entity
    - O: Outside a named entity
    - B: Beginning of a named entity
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## Example

- Consider the sentence: “Lady Gaga is from New York.”

- The IOB annotation for this sentence is:
    - Lady B-PER
    - Gaga I-PER
    - is O
    - from O
    - New B-LOC
    - York I-LOC
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### **Methods for Named Entity Recognition**

- Rule-based methods:
    - Rule-based approach rely on hand-crafted rules to identify named entities.
    - Use domain specific gazetteers (list of names)
    - Combined with regular expressions
  - Unsupervised methods:
    - Use syntactic patterns to identify entities
    - Clustering of words based on context
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### **Feature-based Approaches**

- Use features such as:
    - Part-of-speech tags
    - Word embeddings
    - Contextual embeddings
    - Word shapes
    - Capitalization
    - Prefixes and suffixes
    - etc.
  - Solved with “Good Old Machine Learning” (e.g., SVM)
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### **Deep Learning Approach**

- Input features are extracted for each word:
    - Word-level features (e.g., word embeddings)
    - Character-level features (often using CNNs)
    - Hybrid features (e.g., gazetteers, lexical similarity or POS tags)
  - As named entities often use unusual words, character-level features are helpful
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## Encoding and Decoding

- Context Encoder:
    - A model (e.g., BiLSTM, Transformer) is used to encode the context of each word.
  - Tag Decoder:
    - Probabilities of tags can be calculated using a softmax layer.
    - The most likely sequence of tags can be predicted with a Viterbi decoder.
    - Conditional Random Fields (CRF) can also be used to model the dependencies between tags.
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## Evaluation

- Precision, Recall, and F1-score are used to evaluate NER systems.
  - This is done with “exact-match evaluation”
    - A named entity is considered correct only if its boundaries are correctly predicted.
  - Some evaluations (e.g. MUC-6) used partial match, however this makes evaluation unintuitive and is not commonly used.
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## Automated Term Extraction

- Automated term extraction is the task of identifying terms (e.g., multi-word expressions) in a text.
  - Terms are domain-specific, and can be used to build a domain-specific ontology.
  - Named entities are often terms, but not all terms are named entities.
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## Key-phrase Extraction

- Key-phrase extraction is the task of identifying important phrases in a text.
  - These are typically used to summarize and index a document.
  - Key-phrases are not necessarily terms or named entities, and can be single words or multi-word expressions.
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## Example

Within the cutting-edge oncology department of St. Mary’s Hospital, groundbreaking research is underway, focusing on innovative

treatments for aggressive brain tumors like glioma.

- Named entities: St. Mary's Hospital
  - Terms: oncology, brain tumors, glioma
  - Key-phrases: groundbreaking research, innovative treatments
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## Automated Term Extraction

- Unsupervised methods for term extraction:
    - Use statistical measures such as frequency, mutual information, or pointwise mutual information.
    - Use syntactic patterns to identify terms.
  - The state-of-the-art approaches for term extraction are similar to NER:
    - Deep-learning methods using pretrained language models (e.g., RoBERTa)
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## Key-phrase Extraction

- Unsupervised methods are more successful for key-phrase extraction.
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## KeyBERT

- N-grams are extracted from the text.
  - For each n-gram, BERT is used to calculate the embedding.
  - An embedding is also calculated for the entire document.
  - The cosine similarity between the n-gram embedding and the document embedding is used to rank the n-grams.
  - The top n-grams are selected as key-phrases.
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## Entity Linking

- Entity linking is the task of linking named entities to a knowledge base.
  - Typically a knowledge base such as Wikipedia or Wikidata is used.
  - This is useful for disambiguating named entities, and for enriching the text with additional information.
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## Entity Linking - Example

**Swift** began writing songs professionally at age 14 and signed with Big Machine Records in 2005.

Candidate entities for Swift: \* Taylor Swift (singer) \* Swift (programming language) \* Society for Worldwide Interbank Financial Telecommunication (SWIFT)

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## Entity Linking

- Entity linking is typically approached as a multi-step process:
    - Named entity recognition
    - Candidate generation
    - Entity Ranking
  - State-of-the-art approaches create an embedding for the word in context and an embedding for each candidate.
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## Candidate Generation

- Candidate generation is the task of generating a list of candidate entities for a named entity.
  - Entities are found in a knowledge base, such as Wikipedia or Wikidata.
  - Mentions of entities may be shortened.
  - Entities may also have aliases (e.g., “The Beatles” and “The Fab Four”).
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## Context embedding

- The context embedding is generally the word embedding of the named entity in the context.
    - Methods like attention or pooling are used to combine the word embeddings of multiword entities.
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## Candidate embedding

- The candidate embedding is dependent on the knowledge base:
    - First sentence of the Wikipedia page
    - Graph neural network for a knowledge graph
  - Similarity is calculated using a Siamese network or cosine similarity.
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## Summary

- Named Entity Recognition (NER) is the task of identifying and classifying named entities in a text.

- Named entities are typically names of persons, organizations, locations, dates, etc.
- NER is typically approached as a sequence labeling task.
- Automated term extraction and key-phrase extraction are related tasks, but are not the same as NER.
- Entity linking is the task of linking named entities to a knowledge base.