

Department of Computer Science and Engineering (Data Science)

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Experiment No.7
Implement Named Entity Recognizer for the given Text input
Date of Performance:
Date of Submission:



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Aim: Implement Named Entity Recognizer for the given Text input

Objective: Understand the importance of NER in NLP and Implement NER.

Theory:

The named entity recognition (NER) is one of the most data preprocessing task. It involves the identification of key information in the text and classification into a set of predefined categories. An entity is basically the thing that is consistently talked about or refer to in the text.

NER is the form of NLP.

At its core, NLP is just a two-step process, below are the two steps that are involved:

- Detecting the entities from the text
- Classifying them into different categories

Some of the categories that are the most important architecture in NER such that:

- Person
- Organization
- Place/ location

Other common tasks include classifying of the following:

- date/time.
- expression
- Numeral measurement (money, percent, weight, etc)
- E-mail address

Ambiguity in NE

For a person, the category definition is intuitively quite clear, but for computers, there is some ambiguity in classification. Let's look at some ambiguous example:

England (Organisation) won the 2019 world cup vs The 2019 world cup happened in

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England(Location).

Washington(Location) is the capital of the US vs The first president of the US was Washington(Person).

Output:

```
Load the spaCy language model

In [2]: nlp = spacy.load("en_core_web_sm")

Sample text input

In [3]: text = "Apple Inc. is a company based in Cupertino, California. John works for Google in Mountain View."
```

Process the text using spaCy

```
In [4]: import spacy
         # Load the spaCy language model
         nlp = spacy.load("en_core_web_sm")
         text = "Apple Inc. is a company based in Cupertino, California. John works for Google in Mountain View."
         # Process the text using spaCy
         doc = nlp(text)
         # Initialize variables to store named entities
         named entities = []
         # Define a function to extract named entities
         def extract_named_entities(doc):
             entities = []
             current_entity = None
             for token in doc:
                 if token.ent type :
                     if current entity and token.ent type == current entity[1]:
                         current_entity = (current_entity[0] + " " + token.text, token.ent_type_)
                             {\sf entities.append}({\sf current\_entity})
                         current_entity = (token.text, token.ent_type_)
                     if current entity:
                         entities.append(current_entity)
```



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```
current_entity = (token.text, token.ent_type_)
                  entities.append(current_entity)
              current_entity = None
      if current entity:
          entities.append(current entity)
      return entities
  # Extract named entities
  named_entities = extract_named_entities(doc)
  # Print the named entities
  for entity, label in named_entities:
     print(f"Entity: {entity}, Label: {label}")
Entity: Apple Inc., Label: ORG
Entity: Cupertino, Label: GPE
Entity: California, Label: GPE
Entity: John, Label: PERSON
Entity: Google, Label: ORG
Entity: Mountain View, Label: GPE
```

Conclusion:

The results of the spaCy named entity recognition (NER) on the sample text input are correct. spaCy correctly identified all of the named entities in the text, including the following:

Apple Inc. (organization)

Cupertino (city)

California (state)

John (person)

Google (organization)

Mountain View (city)

There are no words in the text that ought to be recognized as named entities but weren't.