Vidyavardhini's College of Engineering and Technology Department of Artificial Intelligence & Data Science

| 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 England(Location).

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

Code:

import spacy

Load the spaCy language model

nlp = spacy.load("en_core_web_sm")

Sample text input

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)

Extract named entities

named_entities = [(ent.text, ent.label_) for ent in doc.ents]

Print the named entities

for entity, label in named_entities:

print(f"Entity: {entity}, Label: {label}")



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Output:

(venv) PS D:\Vartak college\sem 7\NLP\EXP\New folder> python .\exp7.py

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:

Implementation of named entity recognition using spaCy successfully extracts various entities from the given text. Here's a breakdown of the results and comments on the identified named entities:

Extracted Named Entities

- 1. Apple Inc. Label: ORG
 - a. Correctly identified as an organization. No issues here.
- 2. Cupertino Label: GPE
 - a. Correctly identified as a geographical entity (GPE). Cupertino is a city in California.
- 3. California Label: GPE
 - a. Correctly identified as a geographical entity (GPE). California is a state in the U.S.
- 4. John Label: PERSON
 - a. Correctly identified as a person. This is accurate based on the context provided.
- 5. Google Label: ORG
 - a. Correctly identified as an organization. This is also accurate.
- 6. Mountain View Label: GPE
 - a. Correctly identified as a geographical entity (GPE). Mountain View is another city in California.

Comments on Results

- The implementation appears to function correctly in recognizing all relevant named entities in the provided text. The entities have been classified accurately according to spaCy's entity types.
- There are no misclassifications or omissions in this instance. All key organizations, persons, and geographical locations mentioned in the text are correctly labeled.