

Lab No. 13

Natural Language Processing (NLP)

This laboratory session introduces students to Word2Vec, a popular technique in Natural Language Processing (NLP) used to represent words as meaningful dense vectors. Using textual data from the *Game of Thrones* series, students will learn how word embeddings capture semantic relationships between words based on their context. Through preprocessing, model training, and similarity analysis, this lab helps students understand how machines learn word meanings and relationships from large text corpora in an unsupervised manner.

LAB Objectives:

- Understand **distributional semantics**
- Learn how **Word2Vec** converts words into dense vectors
- Apply **Word2Vec (Skip-Gram / CBOW)** on real textual data (*Game of Thrones*)
- Explore **word similarity, analogy, and visualization**

game-of-thrones-word2vec

word2vec applied on game of thrones data

Dataset Link: <https://www.kaggle.com/khulasasndh/game-of-thrones-books>

Download the data set from Kaggle

Game Of Thrones books

Data Card Code (47) Discussion (0) Suggestions (0)

29 Code Download

001ssb.txt (1.63 MB)

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About this file

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Create Notebook

A Game Of Thrones

Book One of A Song of Ice and Fire

By George R. R. Martin

PROLOGUE

"We should start back," Gared urged as the woods began to grow dark around them. "The wildlings are dead."

Data Explorer

Version 1 (9.9 MB)

001ssb.txt
002ssb.txt
003ssb.txt
004ssb.txt
005ssb.txt

Summary

5 files

Add the dataset in folder data where VS code directory present

VScode Examples				
File	Home	Share	View	
	← → ↻ ↺	📁	This PC	C:\Users\Syed Hamedoon\VScode Examples
Quick access				
Desktop				Name Date modified Type Size
Downloads				📁 .vscode 8/25/2025 3:54 PM File folder
Documents				📁 data 12/22/2025 12:03 PM File folder
Pictures				📁 myenv 9/10/2025 10:10 AM File folder
AI course				ZIP archive (1) 12/22/2025 12:02 PM WinRAR ZIP archive 3,801 KB
FALL 2025				.ArraysP 9/10/2025 10:33 AM Python Source File 1 KB
System network adr				.cardpolarrunner 12/17/2025 11:55 AM Jupyter Source File 1 KB
VScode Examples				.cartpolarun 12/16/2025 4:33 PM Python Source File 1 KB
OneDrive				.chatbot 12/14/2025 2:11 PM Python Source File 1 KB
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3D Objects				.DTandReg 10/8/2025 9:35 AM Jupyter Source File 76 KB
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Code in jupiter VS code:

```
import numpy as np  
import pandas as pd
```

```
!pip install gensim
```

```
import gensim  
import os
```

```
!pip install nltk
```

```
data = "C:/Users/Syed Hamedoon/VScode Examples/data"
```

```
import nltk  
  
nltk.download('punkt')  
nltk.download('punkt_tab')
```

```
import os  
from nltk import sent_tokenize  
from gensim.utils import simple_preprocess  
  
DATA_PATH = r"C:\Users\Syed Hamedoon\VScode Examples\data"  
  
story = []  
  
for filename in os.listdir(DATA_PATH):  
    if filename.endswith(".txt"):  
        file_path = os.path.join(DATA_PATH, filename)  
  
        try:  
            with open(file_path, "r", encoding="utf-8") as f:  
                corpus = f.read()  
        except UnicodeDecodeError:  
            with open(file_path, "r", encoding="cp1252") as f:  
                corpus = f.read()  
  
        for sent in sent_tokenize(corpus):  
            story.append(simple_preprocess(sent))
```

```
print(len(story))
print(story[:2])
```

```
model = gensim.models.Word2Vec(
    window=10,
    min_count=2
)
```

```
model.build_vocab(story)
```

```
model.train(story, total_examples=model.corpus_count, epochs=model.epochs)
```

```
model.wv.most_similar('daenerys')
```

```
model.wv.doesnt_match(['jon','rikon','robb','arya','sansa','bran'])
```

```
model.wv.doesnt_match(['cersei', 'jaime', 'bronn', 'tyrion'])
```

```
model.wv['king']
```

```
model.wv.similarity('arya','sansa')
```

```
model.wv.similarity('tywin','sansa')
```

```
model.wv.get_normed_vectors()
```

```
y = model.wv.index_to_key
```

```
len(y)
```

```
y
```

```
from sklearn.decomposition import PCA
```

```
pca = PCA(n_components=3)
```

```
X = pca.fit_transform(model.wv.get_normed_vectors())
```

```
X.shape
```

```
!pip install --upgrade nbformat
```

```
import pandas as pd
import plotly.express as px
import plotly.io as pio

pio.renderers.default = "browser" # ← IMPORTANT

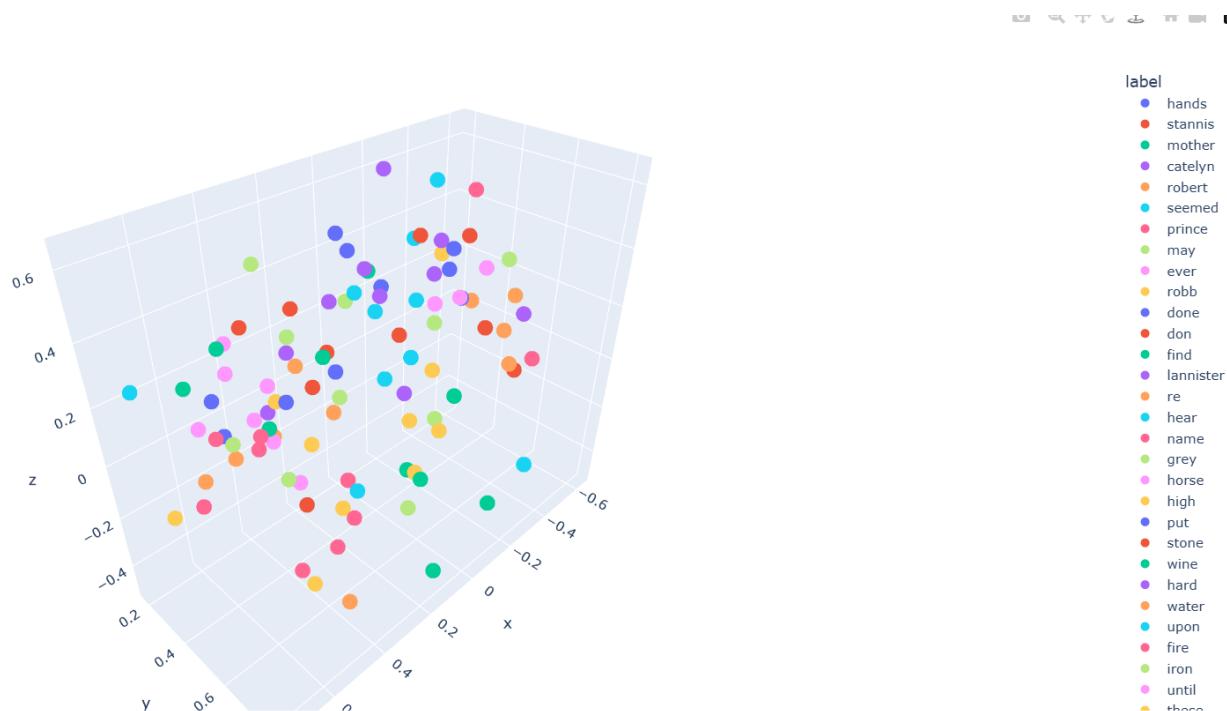
df = pd.DataFrame(X[200:300], columns=["x", "y", "z"])
df["label"] = y[200:300]
```

```

fig = px.scatter_3d(
    df,
    x="x",
    y="y",
    z="z",
    color="label"
)
fig.show()

```

Output:



LAB Questions

1. What is the core idea behind Word2Vec?
2. Difference between CBOW and Skip-Gram?
3. Why is one-hot encoding inefficient?

4. Why do character names appear close in vector space?

5. How does window size affect semantic learning?

6. Why might rare characters have poor embeddings?

7. Which model performed better: CBOW or Skip-Gram? Why?

8. What happens if vector size is too small or too large?

9. Can Word2Vec understand word meaning without labels? Explain.

LAB Assessment

Student Name		LAB Rubrics	CLO3 , P5, PLO5
		Total Marks	10
Registration No		Obtained Marks	
		Teacher Name	Dr. Syed M Hamedoon
Date		Signature	