

```
!pip install gensim
```

```
Collecting gensim
  Downloading gensim-4.4.0-cp312-cp312-manylinux_2_24_x86_64.manylinux_2_28_x86_64.whl.metadata (8.4 kB)
Requirement already satisfied: numpy>=1.18.5 in /usr/local/lib/python3.12/dist-packages (from gensim) (2.0.2)
Requirement already satisfied: scipy>=1.7.0 in /usr/local/lib/python3.12/dist-packages (from gensim) (1.16.3)
Requirement already satisfied: smart_open>=1.8.1 in /usr/local/lib/python3.12/dist-packages (from gensim) (7.5.0)
Requirement already satisfied: wrapt in /usr/local/lib/python3.12/dist-packages (from smart_open>=1.8.1->gensim) (2.1.1)
Downloading gensim-4.4.0-cp312-cp312-manylinux_2_24_x86_64.manylinux_2_28_x86_64.whl (27.9 MB)
----- 27.9/27.9 MB 21.4 MB/s eta 0:00:00
Installing collected packages: gensim
Successfully installed gensim-4.4.0
```

```
import gensim.downloader as api
import numpy as np
import matplotlib.pyplot as plt
from sklearn.manifold import TSNE
```

```
model = api.load("glove-wiki-gigaword-100")
```

```
[=====] 100.0% 128.1/128.1MB downloaded
```

```
print("Vocabulary size:", len(model))
```

```
Vocabulary size: 400000
```

```
print("Vector for 'king':\n", model['king'])
```

```
Vector for 'king':
[-0.32307 -0.87616  0.21977  0.25268  0.22976  0.7388  -0.37954
 -0.35307 -0.84369 -1.1113  -0.30266  0.33178 -0.25113  0.30448
 -0.077491 -0.89815  0.092496 -1.1407  -0.58324  0.66869 -0.23122
 -0.95855  0.28262 -0.078848  0.75315  0.26584  0.3422  -0.33949
  0.95608  0.065641  0.45747  0.39835  0.57965  0.39267 -0.21851
  0.58795 -0.55999  0.63368 -0.043983 -0.68731 -0.37841  0.38026
  0.61641 -0.88269 -0.12346 -0.37928 -0.38318  0.23868  0.6685
 -0.43321 -0.11065  0.081723  1.1569  0.78958 -0.21223 -2.3211
 -0.67806  0.44561  0.65707  0.1045  0.46217  0.19912  0.25802
  0.057194  0.53443 -0.43133 -0.34311  0.59789 -0.58417  0.068995
  0.23944 -0.85181  0.30379 -0.34177 -0.25746 -0.031101 -0.16285
  0.45169 -0.91627  0.64521  0.73281 -0.22752  0.30226  0.044801
 -0.83741  0.55006 -0.52506 -1.7357  0.4751 -0.70487  0.056939
 -0.7132  0.089623  0.41394 -1.3363 -0.61915 -0.33089 -0.52881
  0.16483 -0.98878 ]
```

```
words = [
    # Animals
    "cat", "dog", "lion", "tiger", "elephant", "horse",

    # Fruits
    "apple", "banana", "mango", "grape", "orange",

    # Countries
    "india", "china", "france", "germany", "japan",

    # Technology
    "computer", "laptop", "keyboard", "mouse", "internet",

    # Royalty
    "king", "queen", "prince", "princess",

    # Vehicles
    "car", "bus", "train", "bike", "truck"
]
```

```
vectors = np.array([model[word] for word in words])
```

```
tsne = TSNE(n_components=2, random_state=42, perplexity=5)
reduced_vectors = tsne.fit_transform(vectors)
```

```
plt.figure(figsize=(10,8))

for i, word in enumerate(words):
    x, y = reduced_vectors[i]
    plt.scatter(x, y)
    plt.text(x+0.5, y+0.5, word)
```

```
plt.title("t-SNE Visualization of Word Embeddings")  
plt.show()
```

