🛆 Autonomous Agent Creation Mind Masters Oren Moreno 2024Spr.ipynb 🔯 🔲 Comment 🛮 😃 Share 🌼 File Edit View Insert Runtime Tools Help All changes saved + Code + Text @ Colab AI := Q # Import libraries import pandas as pd from sklearn.datasets import fetch\_20newsgroups  $\{x\}$ from nltk.corpus import stopwords import nltk **⊙** nltk.download('stopwords') from sklearn.feature\_extraction.text import TfidfVectorizer  $\Gamma$ from gensim.models import Word2Vec import gensim.downloader as api [inltk\_data] Downloading package stopwords to /root/nltk\_data... [nltk\_data] Unzipping corpora/stopwords.zip. [ ] # Load the Word2Vec model model = api.load('word2vec-google-news-300') [======] 100.0% 1662.8/1662.8MB downloaded [ ] # Load the 20newsgroups dataset newsgroups\_train = fetch\_20newsgroups(subset='train') [ ] # Takes lists of texts and their titles then finds the most unique word (keyword) in each text according to TF-IDF def get top tfidf word(texts, category name): <> stop\_words = stopwords.words('english' vectorizer = TfidfVectorizer(stop\_words=stop\_words) # Creates vectorizer including stopwords to be ignored vectors = vectorizer.fit transform(texts) # Fits text to vectorizer by transforming to TF-IDF matrix  $\equiv$ feature\_names = vectorizer.get\_feature\_names\_out() # Gets words of associated TF-IDF scores >\_ dense = vectors.todense() # Turns sparse TF-IDF matrix into dense
denselist - dense.tolist() # Makes dense matrix a list of lists df = pd.DataFrame(denselist, columns=feature\_names) # Creates dataframe with words from dataset as columns top\_word\_idx = df.sum().argmax() # Finds the index position with the highest TF-IDF score summed up across all documents in the category top\_word = feature\_names[top\_word\_idx] # Gets the word at that position top\_tfidf = df.sum().max() # Gets the TF-IDF score for that word return top\_word, top\_tfidf [ ] top\_words = [] # Empty list to store keywords # Loops through each category in dataset and outputs the unique keyword with its for category in newsgroups\_train.target\_names: category\_index = newsgroups\_train.target\_names.index(category) # Locates index position of category category\_mask = (newsgroups\_train.target == category\_index) # Specifies the documents that belong to category using boolean mask category\_texts = [text for text, mask in zip(newsgroups\_train.data, category\_mask) if mask] # Filters out the documents that do not belong to category top\_word, top\_tfidf = get\_top\_tfidf, word(category\_texts, category) # Calls keyword function
top\_words.append((top\_word, top\_tfidf, category)) # Adds the word, its TF-IDF score, and category to list of keywords print(f"Top word for category '{category}': {top\_word} (TF-IDF score: {top\_tfidf:.4f})") Top word for category 'alt.atheism': edu (TF-IDF score: 24.0282)
Top word for category 'comp.graphics': edu (TF-IDF score: 22.0857)
Top word for category 'comp.sys.ibm.pc.hardware': edu (TF-IDF score: 23.7632)
Top word for category 'comp.sys.ibm.pc.hardware': edu (TF-IDF score: 20.7596)
Top word for category 'comp.sys.mac.hardware': edu (TF-IDF score: 24.8660)
Top word for category 'comp.windows.x': edu (TF-IDF score: 19.3980)
Top word for category 'misc.forsale': edu (TF-IDF score: 23.9920)
Top word for category 'rec.autos': edu (TF-IDF score: 23.1517)
Top word for category 'rec.sport.baseball': edu (TF-IDF score: 29.9561)
Top word for category 'rec.sport.baseball': edu (TF-IDF score: 29.9561)
Top word for category 'rec.sport.hockey': edu (TF-IDF score: 25.9939)
Top word for category 'sci.electronics': edu (TF-IDF score: 21.593)
Top word for category 'sci.electronics': edu (TF-IDF score: 21.593)
Top word for category 'sci.space': edu (TF-IDF score: 26.1149)
Top word for category 'talk.politics.guns': edu (TF-IDF score: 24.3857)
Top word for category 'talk.politics.mideast': edu (TF-IDF score: 24.3859)
Top word for category 'talk.politics.misc': edu (TF-IDF score: 24.3859)
Top word for category 'talk.politics.misc': edu (TF-IDF score: 24.3859)
Top word for category 'talk.politics.misc': edu (TF-IDF score: 24.3859) ↑ ↓ ⇔ 🗏 🗘 🗓 ᠄ print("\nKeyword Similarities:") # Calculate similarities between top words
for i in range(len(top\_words)): word1, tfidf1, category1 = top\_words[i] most\_similar\_word = None
max\_similarity = -1 for j in range(len(top\_words)):
 if i != j: # Makes sure word is from different category word2, tfidf2, category2 = top words[j] similarity = model.similarity(word1, word2) # Calculates similarity using word2vec representations
if similarity > max\_similarity: # Checks to find the most similar keyword in list of keywords
 max\_similarity = similarity most\_similar\_word = (word2, category2) except KeyError: if most\_similar\_word:  $print(f'''\{word1\}'' (\{category1\}) \ is \ most \ similar \ to \ '\{most\_similar\_word[0]\}'' (\{most\_similar\_word[1]\}) \ with \ similarity: \{max\_similarity: .4f\}'') \ is \ most \ similarity: \{max\_similarity: .4f\}''' \ (\{category1\}) \ is \ most \ similarity: .4f\}''' \ (\{category1\}) \ is \ most \ similarity: \{max\_similarity: .4f\}''' \ (\{category1\}) \ is \ most \ similarity: \{max\_similarity: .4f\}''' \ (\{category1\}) \ is \ most \ similarity: .4f\}'''$ print(f"No similar word found for '{word1}' ({category1})") Keyword Similarities: Keyword Similarities:
'edu' (alt.atheism) is most similar to 'edu' (comp.graphics) with similarity: 1.0000
'edu' (comp.graphics) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (comp.sy.sheim.pc.hardware) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (comp.sy.sheim.pc.hardware) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (comp.sys.mac.hardware) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (comp.windows.x) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (misc.forsale) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (rec.autos) is most similar to 'edu' (alt.atheism) with similarity: 0.0000
'com' (rec.motorcycles) is most similar to 'edu' (alt.atheism) with similarity: 1.0000

```
'edu' (rec.sport.hockey) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'key' (sci.crypt) is most similar to 'edu' (alt.atheism) with similarity: 0.0109
'edu' (sci.electronics) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (sci.med) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (sci.space) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'god' (soc.religion.christian) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (talk.politics.guns) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (talk.politics.mideast) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (talk.politics.misc) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
'edu' (talk.religion.misc) is most similar to 'edu' (alt.atheism) with similarity: 1.0000
```

Colab paid products - Cancel contracts here

✓ Connected to Python 3 Google Compute Engine backend