Import Packages

```
# Run in python console
import nltk; nltk.download('stopwords')
# Run in terminal or command prompt
# !python3 -m spacy download en
    [nltk_data] Downloading package stopwords to /root/nltk_data...
                 Unzipping corpora/stopwords.zip.
!pip install pyLDAvis
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
    Collecting pyLDAvis
      Downloading pyLDAvis-3.4.0-py3-none-any.whl (2.6 MB)
                                                  · 2.6/2.6 MB 40.1 MB/s eta 0:00:00
    Requirement already satisfied: numexpr in /usr/local/lib/python3.9/dist-packages (from pyLDAvis) (2.8.4)
      Downloading funcy-2.0-py2.py3-none-any.whl (30 kB)
    Requirement already satisfied: scikit-learn>=1.0.0 in /usr/local/lib/python3.9/dist-packages (from pyLDAvis) (1.2.2)
    Collecting joblib>=1.2.0
      Downloading joblib-1.2.0-py3-none-any.whl (297 kB)
                                                - 298.0/298.0 KB 41.0 MB/s eta 0:00:00
    Requirement already satisfied: setuptools in /usr/local/lib/python3.9/dist-packages (from pyLDAvis) (67.6.1)
    Requirement already satisfied: scipy in /usr/local/lib/python3.9/dist-packages (from pyLDAvis) (1.10.1)
    Requirement already satisfied: pandas>=1.3.4 in /usr/local/lib/python3.9/dist-packages (from pyLDAvis) (1.4.4)
    Requirement already satisfied: jinja2 in /usr/local/lib/python3.9/dist-packages (from pyLDAvis) (3.1.2)
    Requirement already satisfied: gensim in /usr/local/lib/python3.9/dist-packages (from pyLDAvis) (4.3.1)
    Requirement already satisfied: numpy>=1.22.0 in /usr/local/lib/python3.9/dist-packages (from pyLDAvis) (1.22.4)
    Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.9/dist-packages (from pandas>=1.3.4->pyLDAvis
    Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.9/dist-packages (from pandas>=1.3.4->pyLDAvis) (2022.7.
    Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.9/dist-packages (from scikit-learn>=1.0.0->pyLI
    Requirement already satisfied: smart-open>=1.8.1 in /usr/local/lib/python3.9/dist-packages (from gensim->pyLDAvis) (6.3.0)
    Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.9/dist-packages (from jinja2->pyLDAvis) (2.1.2)
    Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.9/dist-packages (from python-dateutil>=2.8.1->pandas>=1.3.4
    Installing collected packages: funcy, joblib, pyLDAvis
      Attempting uninstall: joblib
        Found existing installation: joblib 1.1.1
        Uninstalling joblib-1.1.1:
          Successfully uninstalled joblib-1.1.1
    ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is
    pandas-profiling 3.2.0 requires joblib~=1.1.0, but you have joblib 1.2.0 which is incompatible.
    Successfully installed funcy-2.0 joblib-1.2.0 pyLDAvis-3.4.0
import re
import numpy as np
import pandas as pd
from pprint import pprint
# Gensim
import gensim
import gensim.corpora as corpora
from gensim.utils import simple preprocess
from gensim.models import CoherenceModel
# spacy for lemmatization
import spacy
# Plotting tools
import pyLDAvis
import pyLDAvis.gensim # don't skip this
import matplotlib.pyplot as plt
%matplotlib inline
# Enable logging for gensim - optional
import logging
logging.basicConfig(format='%(asctime)s : %(levelname)s : %(message)s', level=logging.ERROR)
import warnings
warnings.filterwarnings("ignore",category=DeprecationWarning)
```

/usr/local/lib/python3.9/dist-packages/torch/cuda/__init__.py:497: UserWarning: Can't initialize NVML warnings.warn("Can't initialize NVML")

key factors to obtaining good segregation topics:

- 1. The quality of text processing.
- 2. The variety of topics the text talks about.
- 3. The choice of topic modeling algorithm.
- 4. The number of topics fed to the algorithm.
- 5. The algorithms tuning parameters.

Prepare Stopwords

```
# NLTK Stop words
from nltk.corpus import stopwords
stop_words = stopwords.words('english')
stop_words.extend(['from', 'subject', 're', 'edu', 'use'])
```

Import Newsgroups Data

```
# Import Dataset
df = pd.read json('https://raw.githubusercontent.com/selva86/datasets/master/newsgroups.json')
print(df.target_names.unique())
df.head()
     ['rec.autos' 'comp.sys.mac.hardware' 'comp.graphics' 'sci.space'
      'talk.politics.guns' 'sci.med' 'comp.sys.ibm.pc.hardware'
      'comp.os.ms-windows.misc' 'rec.motorcycles' 'talk.religion.misc'
      'misc.forsale' 'alt.atheism' 'sci.electronics' 'comp.windows.x'
      'rec.sport.hockey' 'rec.sport.baseball' 'soc.religion.christian'
      'talk.politics.mideast' 'talk.politics.misc' 'sci.crypt']
                                            content target
                                                                    target_names
         From: lerxst@wam.umd.edu (where's my thing)\nS...
                                                                          rec autos
      1 From: quvkuo@carson.u.washington.edu (Guv Kuo)...
                                                           4 comp.sys.mac.hardware
           From: twillis@ec.ecn.purdue.edu (Thomas E Will...
                                                           4 comp.sys.mac.hardware
          From: jgreen@amber (Joe Green)\nSubject: Re: W...
                                                                      comp.graphics
                                                           1
      4 From: jcm@head-cfa.harvard.edu (Jonathan McDow...
                                                                          sci.space
```

Remove emails and newline characters

```
# Convert to list
data = df.content.values.tolist()
# Remove Emails
data = [re.sub('\S*@\S*\s?', '', sent) for sent in data]
# Remove new line characters
data = [re.sub('\s+', ' ', sent)] for sent in data]
# Remove distracting single quotes
data = [re.sub("\'", "", sent) for sent in data]
pprint(data[:1])
    ['From: (wheres my thing) Subject: WHAT car is this!? Nntp-Posting-Host: '
      'rac3.wam.umd.edu Organization: University of Maryland, College Park Lines:
      '15 I was wondering if anyone out there could enlighten me on this car I saw '
      'the other day. It was a 2-door sports car, looked to be from the late 60 \, \mathrm{s}/
      'early 70s. It was called a Bricklin. The doors were really small. In
      addition, the front bumper was separate from the rest of the body. This is '
      'all I know. If anyone can tellme a model name, engine specs, years of '
```

```
production, where this car is made, history, or whatever info you have on
'this funky looking car, please e-mail. Thanks, - IL ---- brought to you by
'your neighborhood Lerxst ---- ']
```

▼ Tokenize words and Clean-up text

```
def sent_to_words(sentences):
    for sentence in sentences:
       yield(gensim.utils.simple_preprocess(str(sentence), deacc=True)) # deacc=True removes punctuations
data_words = list(sent_to_words(data))
print(data_words[:1])
    [['from', 'wheres', 'my', 'thing', 'subject', 'what', 'car', 'is', 'this', 'nntp', 'posting', 'host', 'rac', 'wam', 'umd', 'e
```

Creating Bigram and Trigram Models

```
# Build the bigram and trigram models
bigram = gensim.models.Phrases(data_words, min_count=5, threshold=100) # higher threshold fewer phrases.
trigram = gensim.models.Phrases(bigram[data_words], threshold=100)
# Faster way to get a sentence clubbed as a trigram/bigram
bigram_mod = gensim.models.phrases.Phraser(bigram)
trigram_mod = gensim.models.phrases.Phraser(trigram)
# See trigram example
print(trigram mod[bigram mod[data words[0]]])
     ['from', 'wheres', 'my', 'thing', 'subject', 'what', 'car', 'is', 'this', 'nntp posting host', 'rac wam umd edu', 'organizati
```

Remove Stopwords, Make Bigrams and Lemmatize

```
# Define functions for stopwords, bigrams, trigrams and lemmatization
def remove stopwords(texts):
   return [[word for word in simple_preprocess(str(doc)) if word not in stop_words] for doc in texts]
def make_bigrams(texts):
   return [bigram_mod[doc] for doc in texts]
def make trigrams(texts):
    return [trigram mod[bigram mod[doc]] for doc in texts]
def lemmatization(texts, allowed postags=['NOUN', 'ADJ', 'VERB', 'ADV']):
    """https://spacy.io/api/annotation"'
   texts out = []
    for sent in texts:
       doc = nlp(" ".join(sent))
       texts out.append([token.lemma for token in doc if token.pos in allowed postags])
    return texts_out
# Remove Stop Words
data words nostops = remove stopwords(data words)
# Form Bigrams
data_words_bigrams = make_bigrams(data_words_nostops)
# Initialize spacy 'en' model, keeping only tagger component (for efficiency)
# python3 -m spacy download en
nlp = spacy.load('en_core_web_sm', disable=['parser', 'ner'])
# Do lemmatization keeping only noun, adj, vb, adv
data_lemmatized = lemmatization(data_words_bigrams, allowed_postags=['NOUN', 'ADJ', 'VERB', 'ADV'])
```

```
print(data_lemmatized[:1])
    [['s', 'thing', 'car', 'nntp_poste', 'host', 'rac_wam', 'university', 'park', 'line', 'wonder', 'enlighten', 'car', 'see', 'c
```

Create the Dictionary and Corpus needed for Topic Modeling

```
# Create Dictionary
id2word = corpora.Dictionary(data_lemmatized)
# Create Corpus
texts = data_lemmatized
# Term Document Frequency
corpus = [id2word.doc2bow(text) for text in texts]
# View
print(corpus[:1])
     [[(0, 1), (1, 1), (2, 1), (3, 1), (4, 5), (5, 1), (6, 2), (7, 1), (8, 1), (9, 1), (10, 1), (11, 1), (12, 1), (13, 1), (14, 1)]
id2word[0]
     'addition'
[[(id2word[id], freq) for id, freq in cp] for cp in corpus[:1]]
     [[('addition', 1),
       ('body', 1),
('bring', 1),
       ('call', 1),
       ('car', 5),
('day', 1),
('door', 2),
       ('early', 1),
('engine', 1),
        ('enlighten', 1),
        ('funky', 1),
       ('history', 1),
       ('host', 1),
('info', 1),
('know', 1),
        ('late', 1),
        ('lerxst', 1),
       ('line', 1),
        ('look', 2),
       ('mail', 1),
       ('make', 1),
('model', 1),
        ('name', 1),
       ('neighborhood', 1),
       ('nntp_poste', 1),
       ('park', 1),
       ('production', 1),
       ('rac_wam', 1),
        ('really', 1),
       ('rest', 1),
        ('s', 1),
        ('see', 1),
        ('separate', 1),
        ('small', 1),
       ('spec', 1),
('sport', 1),
('thank', 1),
        ('thing', 1),
        ('university', 1),
        ('wonder', 1),
        ('year', 1)]]
```

→ Build the Topic Model

```
# Build LDA model
lda model = gensim.models.ldamodel.LdaModel(corpus=corpus,
                                           id2word=id2word,
                                           num topics=20,
                                           random state=100,
                                           update every=1,
                                           chunksize=100.
                                           passes=10,
                                           alpha='auto',
                                           per_word_topics=True)
# Print the Keyword in the 10 topics
pprint(lda model.print topics())
doc_lda = lda_model[corpus]
       '0.065*"cost" + 0.059*"model" + 0.039*"character" + 0.036*"picture" + '
      '0.036*"format" + 0.032*"quality" + 0.032*"associate" + 0.028*"handle" + '
      '0.023*"hole" + 0.023*"gift"'),
      '0.032*"system" + 0.028*"use" + 0.024*"program" + 0.023*"file" + '
      '0.018*"card" + 0.016*"run" + 0.014*"software" + 0.014*"bit" +
      '0.013*"machine" + 0.013*"problem"'),
       '0.022*"positively" + 0.021*"intent" + 0.018*"alarm" + 0.012*"converter" + '
      '0.011*"unnecessary" + 0.007*"provision"'),
     (8,
      '0.249*"window" + 0.057*"monitor" + 0.055*"normal" + 0.041*"do" + '
      '0.032*"font" + 0.023*"left" + 0.020*"widget" + 0.019*"please_respond" + '
      '0.017*"environment" + 0.017*"trivial"'),
       '0.061*"child" + 0.028*"church" + 0.027*"woman" + 0.025*"armenian" + '
      '0.022*"authority" + 0.020*"community" + 0.019*"greek" + 0.017*"period" + '
      '0.017*"turk" + 0.016*"soldier"'),
     (10,
       '0.765*"ax" + 0.035*"physical" + 0.024*"graphic" + 0.014*"direct" + '
      '0.011*"convert" + 0.006*"daughter" + 0.006*"capture" + 0.005*"human being" '
      '+ 0.004*"split" + 0.003*"accomplish"'),
       '0.130*"line" + 0.076*"organization" + 0.074*"write" + 0.063*"article" + '
      '0.056*"nntp_poste" + 0.050*"host" + 0.029*"reply" + 0.024*"thank" + '0.018*"university" + 0.013*"post"'),
       '0.072*"plane" + 0.030*"hi" + 0.021*"subscription" + 0.020*"steve" + '
      '0.015*"divide" + 0.011*"evolve" + 0.010*"intersection" + 0.010*"rip" + '
      '0.008*"upcoming" + 0.007*"script"'),
     (13.
       '0.031*"people" + 0.028*"state" + 0.018*"gun" + 0.017*"government" + '
      '0.017*"law" + 0.016*"right" + 0.015*"kill" + 0.013*"death" + 0.011*"live" + '
      '0.011*"force"'),
       '0.141*"drug" + 0.029*"film" + 0.026*"movie" + 0.025*"stereo" + '
      '0.024*"japanese" + 0.022*"deficit" + 0.020*"plot" + 0.014*"mad" + '
      '0.009*"harley" + 0.007*"deck"'),
       '0.061*"box" + 0.050*"club" + 0.041*"modem" + 0.041*"status" + '
      '0.030*"primary" + 0.029*"routine" + 0.029*"spec" + 0.026*"sufficient" + '
      '0.023*"public_access" + 0.023*"automatically"'),
       '0.152*"drive" + 0.091*"car" + 0.036*"bike" + 0.024*"engine" + 0.023*"nhl" + '
      '0.022*"ride" + 0.018*"road" + 0.017*"weight" + 0.016*"mile" +
      '0.015*"ground"'),
       '0.113*"patient" + 0.060*"disease" + 0.054*"scientific" + '
      '0.050*"computer science" + 0.043*"animal" + 0.041*"health" + '
      '0.040*"treatment" + 0.037*"medical" + 0.033*"dog" + 0.030*"study"'),
       '0.023*"get" + 0.018*"go" + 0.015*"good" + 0.015*"time" + 0.015*"know" + '
      '0.014*"make" + 0.013*"well" + 0.013*"think" + 0.012*"see" + 0.010*"take"'),
       '0.106*"key" + 0.043*"test" + 0.032*"public" + 0.031*"encryption" + '
      '0.028*"security" + 0.028*"server" + 0.022*"clipper" + 0.021*"chip" + '
      '0.018*"secure" + 0.018*"message"')]
```

Compute Model Perplexity and Coherence Score

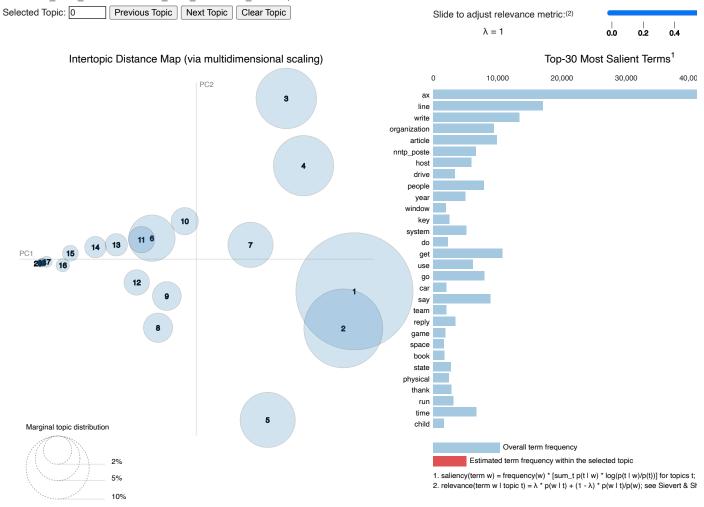
```
# Compute Perplexity
print('\nPerplexity: ', lda_model.log_perplexity(corpus)) # a measure of how good the model is. lower the better.
```

```
# Compute Coherence Score
\verb|coherence_model_lda| = CoherenceModel(model=lda_model, texts=data_lemmatized, dictionary=id2word, coherence='c_v')|
coherence_lda = coherence_model_lda.get_coherence()
print('\nCoherence Score: ', coherence_lda)
    Perplexity: -13.32461333694394
    Coherence Score: 0.483541481988623
```

→ Viz the topic-keywords

```
# Visualize the topics
pyLDAvis.enable_notebook()
vis = pyLDAvis.gensim.prepare(lda_model, corpus, id2word)
```

/usr/local/lib/python3.9/dist-packages/pyLDAvis/_prepare.py:243: FutureWarning: In a future version of pandas all arguments (default_term_info = default_term_info.sort_values(



✓ 12s completed at 3:28 PM

• x