

```

1 data = ""o
2 Sr. NO 123 - XYZ Data
3 XYZ Data is a no-code and highly intuitive data pipeline platform. was founded by Mr. Som
4
5 The built-in CDC for database sources allows you to get up-to-date data from any database
6 The current CEO is Mr. Pankaj Gupta
7 ""

```

```

1 import nltk
2 import pandas as pd
3 import seaborn as sns
4 import matplotlib.pyplot as plt
5 import spacy
6 from spacy import displacy
7 nltk.download('maxent_ne_chunker')
8 from nltk import word_tokenize
9 nltk.download('punkt')
10 from nltk.probability import FreqDist
11 from nltk.util import bigrams, trigrams, ngrams
12 nltk.download('wordnet')
13 from nltk.stem import wordnet
14 nltk.download('words')
15 from nltk.corpus import stopwords
16 nltk.download('averaged_perceptron_tagger')

[nltk_data] Downloading package maxent_ne_chunker to
[nltk_data] /root/nltk_data...
[nltk_data] Package maxent_ne_chunker is already up-to-date!
[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data] Package punkt is already up-to-date!
[nltk_data] Downloading package wordnet to /root/nltk_data...
[nltk_data] Package wordnet is already up-to-date!
[nltk_data] Downloading package words to /root/nltk_data...
[nltk_data] Package words is already up-to-date!
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data] /root/nltk_data...
[nltk_data] Package averaged_perceptron_tagger is already up-to-
[nltk_data] date!
True

```

## ▼ Statistics

```

1 data_tokens = word_tokenize(data) #Converting the dataset into sing
2 data_tokens
    services ,
    '.',
    'You',
    'can',
    '~'

```

```
'also',  
'customized',  
'your',  
'data',  
'connections',  
'using',  
'Rest',  
'API',  
'and',  
'Webhook',  
'integration',  
'to',  
'interact',  
'with',  
'all',  
'kinds',  
'of',  
'applications',  
'.',  
'The',  
'built-in',  
'CDC',  
'for',  
'database',  
'sources',  
'allows',  
'you',  
'to',  
'get',  
'up-to-date',  
'data',  
'from',  
'any',  
'database',  
'to',  
'your',  
'destination',  
'in',  
'real-time',  
'without',  
'putting',  
'additional',  
'load',  
'on',  
'your',  
'database',  
'.',  
'The',  
'current',  
'CEO',  
'is',  
'Mr.',  
'Pankaj',  
'Gupta']
```

```
1 fdist = FreqDist()
```

```
1 for word in data_tokens:
```

```
2     fdist[word.lower()]+=1
```

```
3 fdist
```

#FreqDist is a dict with tokens as keys and the

```
current': 1,  
'customized': 1,  
'data': 6,  
'database': 3,  
'databases': 1,  
'destination': 1,  
'for': 1,  
'founded': 1,  
'from': 2,  
'get': 1,  
'gupta': 1,  
'highly': 1,  
'in': 2,  
'integration': 1,  
'integrations': 1,  
'interact': 1,  
'intuitive': 1,  
'is': 2,  
'it': 1,  
'kinds': 1,  
'load': 2,  
'minutes': 1,  
'mr.': 2,  
'no': 1,  
'no-code': 1,  
'o': 1,  
'of': 1,  
  
'on': 1,  
'pankaj': 1,  
'pipeline': 1,  
'platform': 1,  
'provides': 1,  
'putting': 1,  
'ready-to-use': 1,  
'real-time': 1,  
'reliably': 1,  
'rest': 1,  
'saas': 1,  
'sdks': 1,  
'services': 1,  
'sharma': 1,  
'somnath': 1,  
'sources': 2,  
'sr.': 1,  
'storage': 1,  
'streaming': 1,  
'the': 3,  
'to': 4,  
'up-to-date': 1,  
'using': 1,
```

```
'warehouse': 1,
'was': 1,
'webhook': 1,
'with': 1,
'without': 1,
'xyz': 2,
'you': 3,
'your': 4})
```

```
1 len(fdist)
```

```
80
```

This implies we have 80 distinct features in the given data.

```
1 fdist['mr.']
```

```
2
```

```
1 fdist_top20 = fdist.most_common(20)
```

```
#To extract the top 20 occuring f
```

```
2 fdist_top20
```

```
[('data', 6),
 ('.', 6),
 ('your', 4),
 ('to', 4),
 (',', 4),
 ('and', 3),
 ('you', 3),
 ('the', 3),
 ('database', 3),
 ('xyz', 2),
 ('is', 2),
 ('mr.', 2),
 ('can', 2),
 ('load', 2),
 ('from', 2),
 ('all', 2),
 ('sources', 2),
 ('in', 2),
 ('applications', 2),
 ('o', 1)]
```

```
1 #Plot of features
```

```
2 fig, ax = plt.subplots(figsize=(10,10))
```

```
3 all_fdist = pd.Series(dict(fdist_top20))
```

```
4 pplot = sns.barplot(x=all_fdist.index, y=all_fdist.values, ax=ax)
```

```
5 plt.xticks(rotation=30);
```

n-gram is a contiguous sequence of n items generated from a given sample of text where the items can be characters or words and n can be any numbers like 1,2,3, etc.

```
1 data_bigrams = list(nltk.bigrams(data_tokens))
2 data_bigrams
  ( 'streaming', 'services' ),
  ('Services', '.'),
  ('.', 'You'),
  ('You', 'can'),
  ('can', 'also'),
  ('also', 'customized'),
  ('customized', 'your'),
  ('your', 'data'),
  ('data', 'connections'),
  ('connections', 'using'),
  ('using', 'Rest'),
  ('Rest', 'API'),
  ('API', 'and'),
  ('and', 'Webhook'),
```

```
( 'Webhook', 'integration'),
( 'integration', 'to'),
( 'to', 'interact'),
( 'interact', 'with'),
( 'with', 'all'),
( 'all', 'kinds'),
( 'kinds', 'of'),
( 'of', 'applications'),
( 'applications', '.'),
( '.', 'The'),
( 'The', 'built-in'),
( 'built-in', 'CDC'),
( 'CDC', 'for'),
( 'for', 'database'),
( 'database', 'sources'),
( 'sources', 'allows'),
( 'allows', 'you'),
( 'you', 'to'),
( 'to', 'get'),
( 'get', 'up-to-date'),
( 'up-to-date', 'data'),
( 'data', 'from'),
( 'from', 'any'),
( 'any', 'database'),
( 'database', 'to'),
( 'to', 'your'),
( 'your', 'destination'),
( 'destination', 'in'),
( 'in', 'real-time'),
( 'real-time', 'without'),
( 'without', 'putting'),
( 'putting', 'additional'),
( 'additional', 'load'),
( 'load', 'on'),
( 'on', 'your'),
( 'your', 'database'),
( 'database', '.'),
( '.', 'The'),
( 'The', 'current'),
( 'current', 'CEO'),
( 'CEO', 'is'),
( 'is', 'Mr.'),
( 'Mr.', 'Pankaj'),
( 'Pankaj', 'Gupta')]
```

```
1 data_trigrams = list(nltk.trigrams(data_tokens ))
2 data_trigrams
```

```
( and , streaming , services ),
( 'Streaming', 'Services', '.'),
( 'Services', '.', 'You'),
( '.', 'You', 'can'),
( 'You', 'can', 'also'),
( 'can', 'also', 'customized'),

( 'also', 'customized', 'your'),
( 'customized', 'your', 'data'),
```



```
('and', 'Streaming', 'Services', '.', 'You'),
('Streaming', 'Services', '.', 'You', 'can'),
('Services', '.', 'You', 'can', 'also'),
('.', 'You', 'can', 'also', 'customized'),
('You', 'can', 'also', 'customized', 'your'),
('can', 'also', 'customized', 'your', 'data'),
('also', 'customized', 'your', 'data', 'connections'),
('customized', 'your', 'data', 'connections', 'using'),
('your', 'data', 'connections', 'using', 'Rest'),
('data', 'connections', 'using', 'Rest', 'API'),
('connections', 'using', 'Rest', 'API', 'and'),
('using', 'Rest', 'API', 'and', 'Webhook'),
('Rest', 'API', 'and', 'Webhook', 'integration'),
('API', 'and', 'Webhook', 'integration', 'to'),
('and', 'Webhook', 'integration', 'to', 'interact'),
('Webhook', 'integration', 'to', 'interact', 'with'),
('integration', 'to', 'interact', 'with', 'all'),
('to', 'interact', 'with', 'all', 'kinds'),
('interact', 'with', 'all', 'kinds', 'of'),
('with', 'all', 'kinds', 'of', 'applications'),
('all', 'kinds', 'of', 'applications', '.'),
('kinds', 'of', 'applications', '.', 'The'),
('of', 'applications', '.', 'The', 'built-in'),
('applications', '.', 'The', 'built-in', 'CDC'),
('.', 'The', 'built-in', 'CDC', 'for'),
('The', 'built-in', 'CDC', 'for', 'database'),
('built-in', 'CDC', 'for', 'database', 'sources'),
('CDC', 'for', 'database', 'sources', 'allows'),
('for', 'database', 'sources', 'allows', 'you'),
('database', 'sources', 'allows', 'you', 'to'),
('sources', 'allows', 'you', 'to', 'get'),
('allows', 'you', 'to', 'get', 'up-to-date'),
('you', 'to', 'get', 'up-to-date', 'data'),
('to', 'get', 'up-to-date', 'data', 'from'),
('get', 'up-to-date', 'data', 'from', 'any'),
('up-to-date', 'data', 'from', 'any', 'database'),
('data', 'from', 'any', 'database', 'to'),
('from', 'any', 'database', 'to', 'your'),
('any', 'database', 'to', 'your', 'destination'),
('database', 'to', 'your', 'destination', 'in'),
('to', 'your', 'destination', 'in', 'real-time'),
('your', 'destination', 'in', 'real-time', 'without'),
('destination', 'in', 'real-time', 'without', 'putting'),
('in', 'real-time', 'without', 'putting', 'additional'),
('real-time', 'without', 'putting', 'additional', 'load'),
('without', 'putting', 'additional', 'load', 'on'),
('putting', 'additional', 'load', 'on', 'your'),
('additional', 'load', 'on', 'your', 'database'),
('load', 'on', 'your', 'database', '.'),
('on', 'your', 'database', '.', 'The'),
('your', 'database', '.', 'The', 'current'),
('database', '.', 'The', 'current', 'CEO'),
('.', 'The', 'current', 'CEO', 'is'),
('The', 'current', 'CEO', 'is', 'Mr.'),
('current', 'CEO', 'is', 'Mr.', 'Pankaj'),
('CEO', 'is', 'Mr.', 'Pankaj', 'Gupta')]
```



Chunking is used to identify parts of speech and short phrases present in a given sentence. The different parts of speech present in the given data are as follows:

```

1 from nltk import ne_chunk
2 data_tokens = word_tokenize(data)
3 data_tags = nltk.pos_tag(data_tokens)
4 data_chunks = ne_chunk(data_tags)
5 print(data_chunks)

((ORGANIZATION SDKS/NNP)
 ,/,
 and/CC
 (ORGANIZATION Streaming/NNP Services/NNPS)
 ./,
 You/PRP
 can/MD
 also/RB
 customized/VB
 your/PRP$
 data/NNS
 connections/NNS
 using/VBG
 (PERSON Rest/NNP API/NNP)
 and/CC
 (GPE Webhook/NNP)
 integration/NN
 to/TO
 interact/VB
 with/IN
 all/DT
 kinds/NNS
 of/IN
 applications/NNS
 ./,
 The/DT
 built-in/JJ
 (ORGANIZATION CDC/NNP)
 for/IN
 database/NN
 sources/NNS
 allows/VBZ
 you/PRP
 to/TO
 get/VB
 up-to-date/JJ
 data/NNS
 from/IN
 any/DT
 database/NN
 to/TO
 your/PRP$
 destination/NN
 in/IN
 real-time/NN

```

```

1 read_line/NN
without/IN
putting/VBG
additional/JJ

load/NN
on/IN
your/PRP$
database/NN
./
The/DT
current/JJ
(ORGANIZATION CEO/NNP)
is/VBZ
(PERSON Mr./NNP Pankaj/NNP Gupta/NNP))

```

## ▼ Method 1: From Scratch

```
1 import regex as re
```

```

1 opt = re.sub(r'[.,]', '', data)           #Removing Punctuations
2 opt = re.sub(r'[-|\n]', ' ', opt)         #Removing new line characters
3 opt = re.split(' ', opt)                  #Splitting the words
4 print(opt)

```

```
['oSr', 'NO', '123', '', '', 'XYZ', 'DataXYZ', 'Data', '', 'is', 'a', 'nocode', 'and',
```

```

1 serial_nos = []
2 for word in opt:
3     if (word.isdigit()) :                 #To check if the any word is numeric or no
4         serial_nos.append(word)
5 print(f"The serial number is: {serial_nos[0]}")

```

```
The serial number is: 123
```

```

1 names = []
2 i = 0
3 for word in opt:
4     if (word == 'Mr'):
5         names.append(opt[i+1] + ' ' + opt[i+2])
6     i = i+1
7 print(names)

```

```
['Somnath Sharma', 'Pankaj Gupta']
```

So, we can extract the names and the serial number from scratch but not the job designation.

## ▼ Approach 1 : Using NLTK

```

1 data_tokens = nltk.word_tokenize(data)           #Spillting the data into tokens
2 data_tags = nltk.pos_tag(data_tokens)           #Part of Speech Tagging
3 data_chunks = nltk.ne_chunk(data_tags, binary = True) #whether NE or not NE

```

```

1 for chunk in data_chunks:
2     print(chunk)

```

```

('o', 'JJ')
('Sr.', 'NNP')
('NO', 'NNP')
('123', 'CD')
('-', ':')
(NE XYZ/NN Data/NNP)
('XYZ', 'NNP')
('Data', 'NNP')
('is', 'VBZ')
('a', 'DT')
('no-code', 'JJ')
('and', 'CC')
('highly', 'RB')
('intuitive', 'JJ')
('data', 'NNS')
('pipeline', 'NN')
('platform', 'NN')
('.', '.')
('was', 'VBD')
('founded', 'VBN')
('by', 'IN')
(NE Mr./NNP Somnath/NNP Sharma/NNP)
('.', '.')
('You', 'PRP')
('can', 'MD')
('reliably', 'VB')
('load', 'NN')
('data', 'NNS')
('from', 'IN')
('all', 'DT')
('your', 'PRP$')
('sources', 'NNS')
('to', 'TO')
('the', 'DT')
('warehouse', 'NN')
('in', 'IN')
('minutes', 'NNS')
('.', '.')
('It', 'PRP')
('provides', 'VBZ')
('100+', 'CD')
('ready-to-use', 'NN')

```

```

('integrations', 'NNS')
('across', 'IN')
(NE Databases/NNP)
(',', ', ', ', ')
(NE SaaS/NNP Applications/NNP)
(',', ', ', ', ')
(NE Cloud/NNP Storage/NNP)
(',', ', ', ', ')
(NE SDKs/NNP)
(',', ', ', ', ')
('and', 'CC')
('Streaming', 'NNP')
('Services', 'NNPS')
('.', '. ', '. ')
('You', 'PRP')
('can', '. ', 'MD')

```

Created a dataframe for only the named entities recognized by nltk

```

1 entities = []
2 labels = []
3 for chunk in data_chunks:
4     if hasattr(chunk, 'label'):                                     #to check if the has the correct giv
5         entities.append(' '.join(c[0] for c in chunk))
6         labels.append(chunk.label())
7
8 entities_labels = list(set(zip(entities, labels)))
9 entities_df = pd.DataFrame(entities_labels)
10 entities_df.columns = ['Entities', 'labels']
11 entities_df

```

Next is another dataframe with entities being organization, person and geopolitical entity. We observe that it has plenty of errors such as 'databases' considered under person and so on.

```
1 data_chunks = nltk.ne_chunk(data_tags, binary = False)      #either NE or not NE
2 entities = []
3 labels = []
4 for chunk in data_chunks:
5     if hasattr(chunk, 'label'):
6         entities.append(' '.join(c[0] for c in chunk))
7         labels.append(chunk.label())
8
9 entities_labels = list(set(zip(entities, labels)))
10 entities_df = pd.DataFrame(entities_labels)
11 entities_df.columns = ['Entities', 'labels']
12 entities_df
```

```
1 entities = []
2 labels = []
3
4 sentence = nltk.sent_tokenize(data)
5 for sent in sentence:
6     for chunk in nltk.ne_chunk(nltk.pos_tag(nltk.word_tokenize(sent)), binary = False):
7         if hasattr(chunk, 'label'):
8             entities.append(' '.join(c[0] for c in chunk))
9             labels.append(chunk.label())
10
```

```
11 entities_labels = list(set(zip(entities, labels)))
12 entities_df = pd.DataFrame(entities_labels)
13 entities_df.columns = ['Entities', 'Labels']
14 entities_df
```

In our last dataframe, we observe the people being classified rightly with their job designations being under 'organizatio' entity. This is moderately good as it still has errors which we will try to remove with spacy.

## ▼ Approach 2: Using Spacy

```
1 spacy.__version__
```

```
1 nlp = spacy.load('en_core_web_sm')
```

```
1 data = nlp(data)
2 entities = []
3 labels = []
4 pos_start = []
5 pos_end = []
```

```
6
7 for ent in data.ents:
8     entities.append(ent)
9     labels.append(ent.label_)
10    pos_start.append(ent.start_char)
11    pos_end.append(ent.end_char)
12
13 df = pd.DataFrame({'Entities':entities, 'Labels':labels, 'Position Start':pos_start
14                    , 'Position End': pos_end})
15 df
```

In spacy, we keep a track of the entities' startig and ending positions aswell. We get both the cardinals as preferred and the people correctly classified aswell along with its designation. XYZ Data beig classified under Organization whu=ich was not the case for nltk.

```
1 df[df["Labels"]=="PERSON"]
```

For visualizing the entities in data:

```
1 displacy.serve(data,style = 'ent')
```

Shutting down server on port 5000.

```
1 spacy.displacy.serve(data, style = 'ent')
```

Using the 'ent' visualizer  
Serving on <http://0.0.0.0:5000> ...

Shutting down server on port 5000.

## ▼ Stanford NER

```
1 pip install stanza
```

```
1 import stanza
```

```
1 data = ""o
2 Sr. NO 123 - XYZ Data
3 XYZ Data is a no-code and highly intuitive data pipeline platform. was founded by Mr. Som
4
5 The built-in CDC for database sources allows you to get up-to-date data from any database
6 The current CEO is Mr. Pankaj Gupta
7 ""
```

```
1 stanza.download('en')
2 nlp = stanza.Pipeline('en')
```



Downloading

142k/?

[https://raw.githubusercontent.com/stanfordnlp/stanza-](https://raw.githubusercontent.com/stanfordnlp/stanza-resources/main/resources_1.3.0.json)

[00:00&lt;00:00,

resources/main/resources\_1.3.0.json:

1.15MB/s]

2022-04-04 12:28:28 INFO: Downloading default packages for language: en (English)...

```
1 doc = nlp(data)
```

```
1 for ent in doc.entities:
```

```
2   print(f"{ent.text} : {ent.type}")
```

```
123 : CARDINAL
```

```
Somnath Sharma : PERSON
```

```
minutes : TIME
```

```
100 : CARDINAL
```

```
Databases : ORG
```

```
SaaS Applications : PRODUCT
```

```
Cloud Storage : PRODUCT
```

```
SDKs : ORG
```

```
Streaming Services : ORG
```

```
Rest API : PRODUCT
```

```
Webhook : PRODUCT
```

```
CDC : ORG
```

```
Pankaj Gupta : PERSON
```

This a better estimation of the entities compared to the rest as we've gotten our cardinals, person and Organization(along with job designation) right on point.

