MIMIC NLP

Tools and Dataset Used:

- Filename: https://github.com/dalwari/mimic-iii-clinical-database-demo-1.4/blob/main/mimic_nlp.ipynb
- Dataset Used: https://github.com/dalwari/mimic-iii-clinical-database-demo-1.4/blob/main/nlp_med_notes.csv
- Package Used: Spacy, Scispacy, Word2Vec, TSNE
- Application Used: Dbeaver, VSCode

Dataset Preparation:

- Write a sql query in Dbeaver
- Use the ‘Export data’ option in Dbeaver to extract the dataset result as CSV file into NLP workspace

```
89 •with cte as (
90  select *  from d_icd_diagnoses did where did.icd9_code between '25000' and '25003'),
91  cte1 as (select di.* from diagnoses_icd di inner join cte on cte.icd9_code =di.icd9_code ),
92  cte2 as (select cte1.icd9_code,n.*  from noteevents n inner join cte1 on n.hadm_id =cte1.hadm_id ),
93  cte3 as (
94  select icd9_code,subject_id, hadm_id,category,text from cte2 where category Like'Discharge%' limit 40)
95  select row_number( ) over (order by hadm_id ) as rn, cte3.* from cte3;
96
```

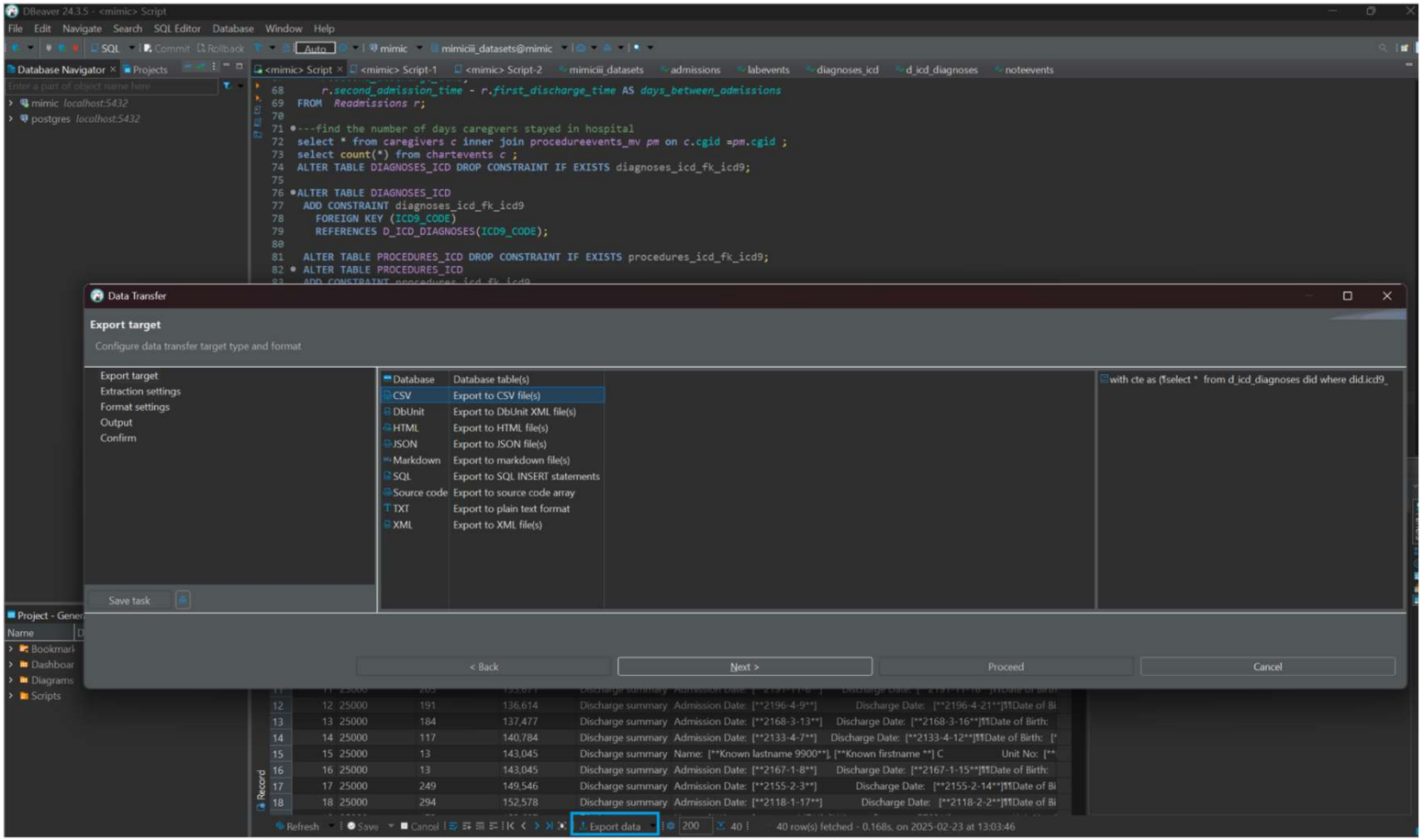
diagnoses_icd(+) 1 ×

with cte as (select * from d_icd_diagnoses did where did.icd9_c | Enter a SQL expression to filter results (use Ctrl+Space)

Grid | Text | Record

	rn	icd9_code	subject_id	hadm_id	category	text
1	1	25000	157	107,880	Discharge summary	Admission Date: [**2106-6-17**] Discharge Date: [**2106-6-24**]¶¶Date of E
2	2	25000	305	108,015	Discharge summary	Admission Date: [**2125-12-31**] Discharge Date: [**2126-1-10**]¶¶Date of
3	3	25000	21	109,451	Discharge summary	Admission Date: [**2134-9-11**] Discharge Date: [**2134-9-24**]¶¶¶Service:
4	4	25000	21	111,970	Discharge summary	Admission Date: [**2135-1-30**] Discharge Date: [**2135-2-8**]¶¶¶Service: M
5	5	25000	75	112,086	Discharge summary	Admission Date: [**2147-4-5**] Discharge Date: [**2147-4-11**]¶¶Date of Bi
6	6	25000	130	113,323	Discharge summary	Unit No: [**Numeric Identifier 56787**]¶Admission Date: [**2119-11-14**]¶Discharge I
7	7	25000	249	116,935	Discharge summary	Admission Date: [**2149-12-17**] Discharge Date: [**2149-12-31**]¶¶Date c
8	8	25000	188	132,401	Discharge summary	Admission Date: [**2161-11-1**] Discharge Date: [**2162-1-17**]¶¶Date of E
9	9	25000	305	133,059	Discharge summary	Admission Date: [**2125-4-26**] Discharge Date: [**2125-5-3**]¶¶Date of Bi
10	10	25000	205	135,671	Discharge summary	Admission Date: [**2191-11-6**] Discharge Date: [**2191-11-14**]¶¶Date of Birth
11	11	25000	205	135,671	Discharge summary	Admission Date: [**2191-11-6**] Discharge Date: [**2191-11-16**]¶¶Date of Birth
12	12	25000	191	136,614	Discharge summary	Admission Date: [**2196-4-9**] Discharge Date: [**2196-4-21**]¶¶Date of Bi
13	13	25000	184	137,477	Discharge summary	Admission Date: [**2168-3-13**] Discharge Date: [**2168-3-16**]¶¶Date of Birth:
14	14	25000	117	140,784	Discharge summary	Admission Date: [**2133-4-7**] Discharge Date: [**2133-4-12**]¶¶Date of Birth: [*
15	15	25000	13	143,045	Discharge summary	Name: [**Known lastname 9900**], [**Known firstname **] C Unit No: [**
16	16	25000	13	143,045	Discharge summary	Admission Date: [**2167-1-8**] Discharge Date: [**2167-1-15**]¶¶Date of Birth:
17	17	25000	249	149,546	Discharge summary	Admission Date: [**2155-2-3**] Discharge Date: [**2155-2-14**]¶¶Date of Bi
18	18	25000	294	152,578	Discharge summary	Admission Date: [**2118-1-17**] Discharge Date: [**2118-2-2**]¶¶Date of Bi

- The disease of interest here is diabetes
- The icd9_code for diabetes is between ‘25000' and '25003’
- First we need to join d_icd_diagnoses and diagnoses_icd
- The result of previous step subquery is joined with noteevents on hospital admission id.
- The notes of ‘Discharge Summary’ needs to be filtered out limited to 40 records in order to process effectively.



Named Entity Recognition:

- Import the Spacy, Scispacy, Word2Vec, TSNE libraries

```python
import pandas as pd
pd.options.mode.chained_assignment = None
import numpy as np
import re
from gensim.models import Word2Vec
import gensim.downloader as api
from sklearn.manifold import TSNE
import matplotlib.pyplot as plt
from spacy import displacy
import spacy
spacy.require_gpu()
```

Python

- Load the spacy model

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

- Prepare utility functions like,

- clean_and_split_paragraph()
- extract_entities()
- fetch_entities()
- visualize_entities()
- extract_corpus()
- fetch_corpus()
- extract_passage_by_label()
- do_data_process()

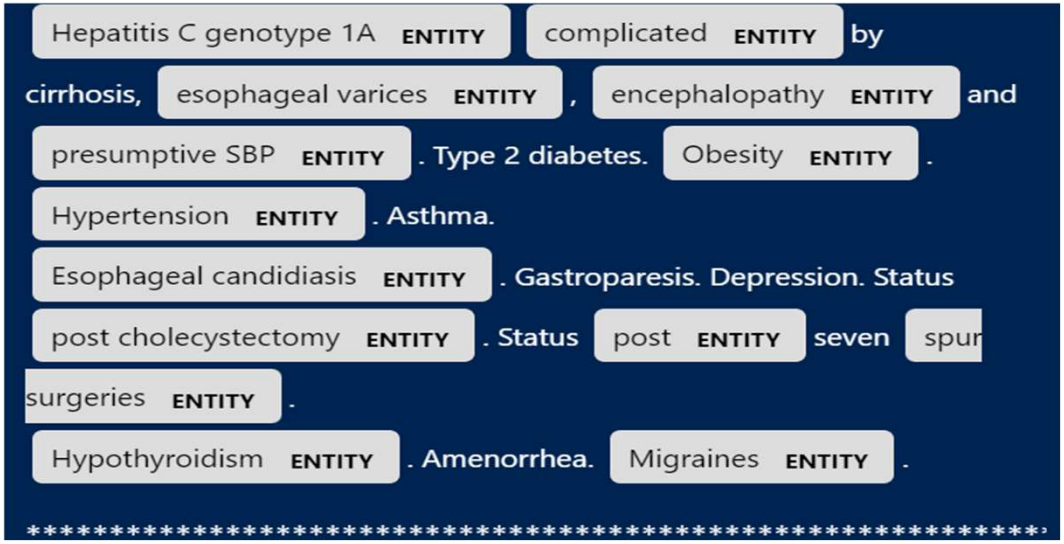
Utility Helper Functions:

- **clean_and_split_paragraph()** is used to perform data cleaning by removing the extra spaces and redundant lines.
- **fetch_entities()** and **extract_entities()** are used to print entities fetched using NER models.
- **fetch_corpus()** and **extract_corpus()** are used to print corpus recognized from the dataset.
- **do_data_process()** and **extract_passage_by_label()** are used to extract subparagraph from medical notes passage.
- **visualize_entities()** is used to identitfy the different types of entities provided by different *Spacy* and *SciSpacy* models
- For instance , *History of Present Illness, Past Medical History, Brief Hospital Course, REVIEW OF SYSTEMS* are the subparagraph labels

- Load and clean the ‘nlp_med_notes.csv’ dataset

```
notes_df = pd.read_csv("nlp_med_notes.csv")['text']
notes=[]
for notes_data in notes_df:
    notes.append(clean_and_split_paragraph(notes_data))
```

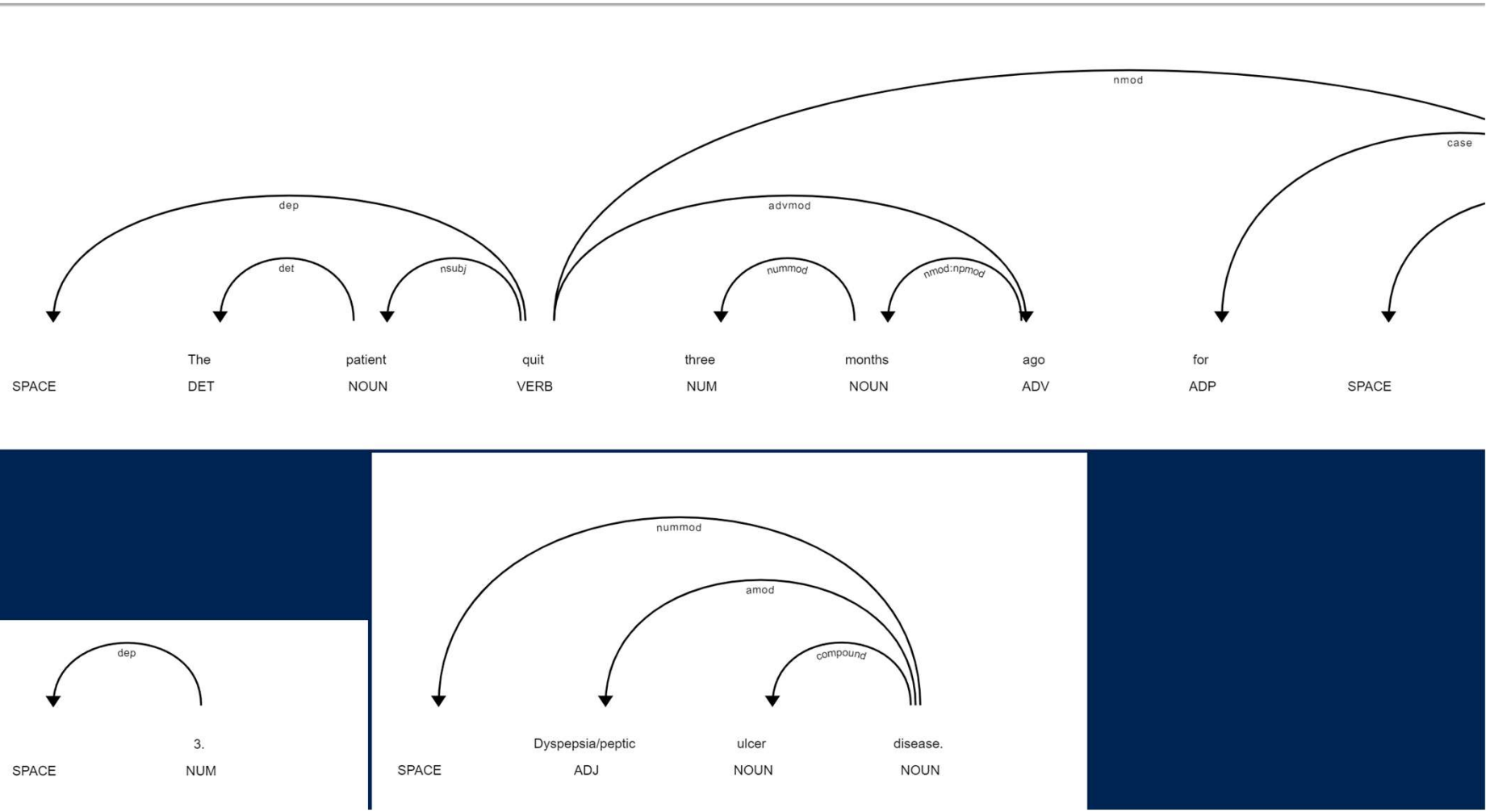
- For each record extract the ‘PAST MEDICAL HISTORY’ subparagraph from text (i.e medical notes) field.
- Visualize the extracted label_df to view the entities of spacy model.

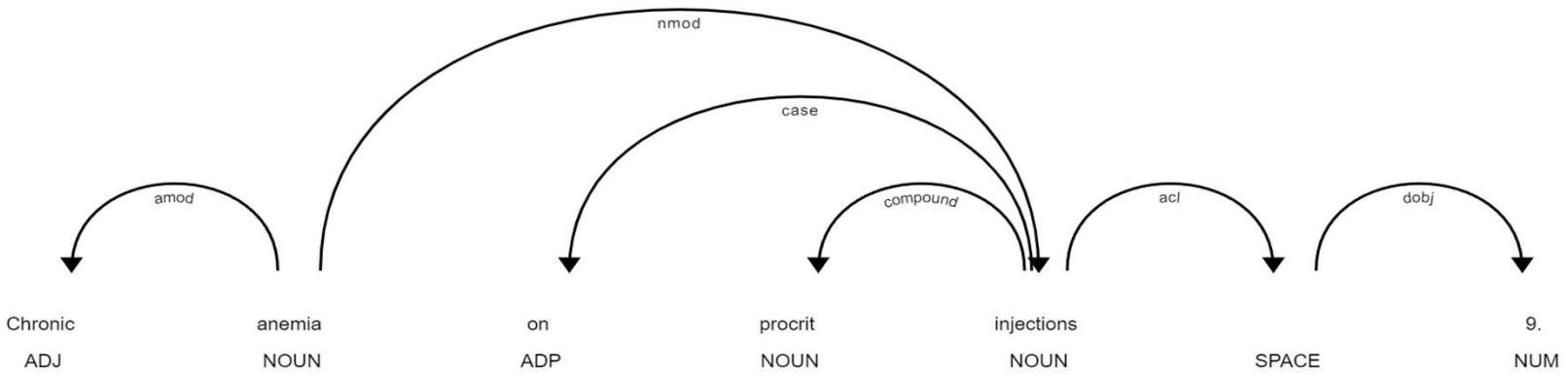


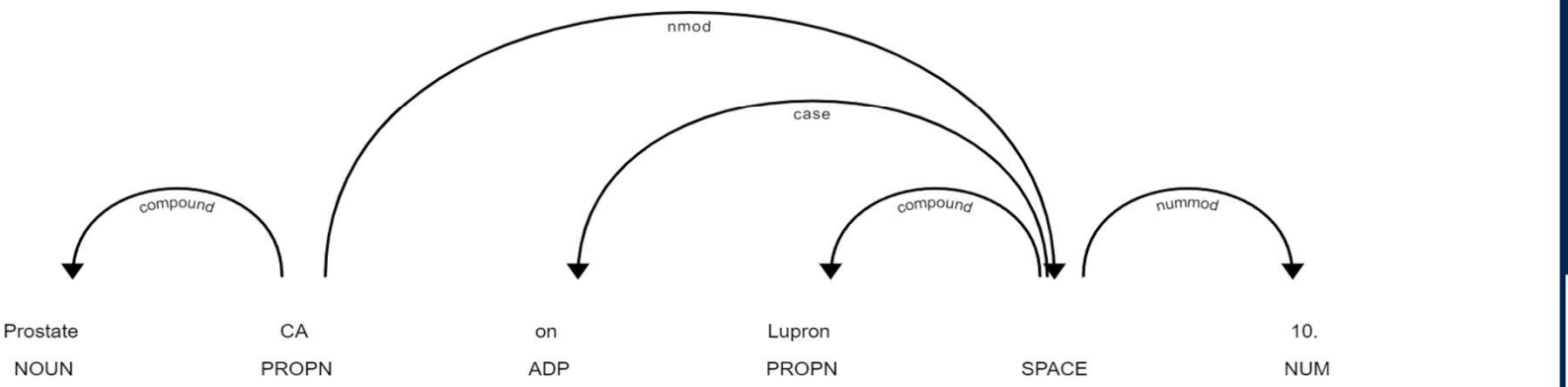

```
df =pd.read_csv("nlp_med_notes.csv")['text']
label_df=do_data_process(df,'REVIEW OF SYSTEMS')

for data in label_df:
    doc = nlp(data)
    sentence_spans = list(doc.sents)
    displacy.render(sentence_spans, style="dep", jupyter=True)
print('**********
```

- Create a **dependency tree** using displacy from spacy model to show the relationship between words in a sentence.
- A meaningful sentence can be extracted from ‘REVIEW OF SYSTEMS’ sub paragraph.


nmod
case
dep
advmod
det
nsubj
nummod
nmod:npmod
The
patient
quit
three
months
ago
for
SPACE
DET
NOUN
VERB
NUM
NOUN
ADV
ADP
SPACE
dep
3.
SPACE
NUM
nummod
amod
compound
Dyspepsia/peptic
ulcer
disease.
SPACE
ADJ
NOUN
NOUN


nmod
case
amod
compound
acl
dobj
Chronic
ADJ
anemia
NOUN
on
ADP
procrit
NOUN
injections
NOUN
SPACE
9.
NUM

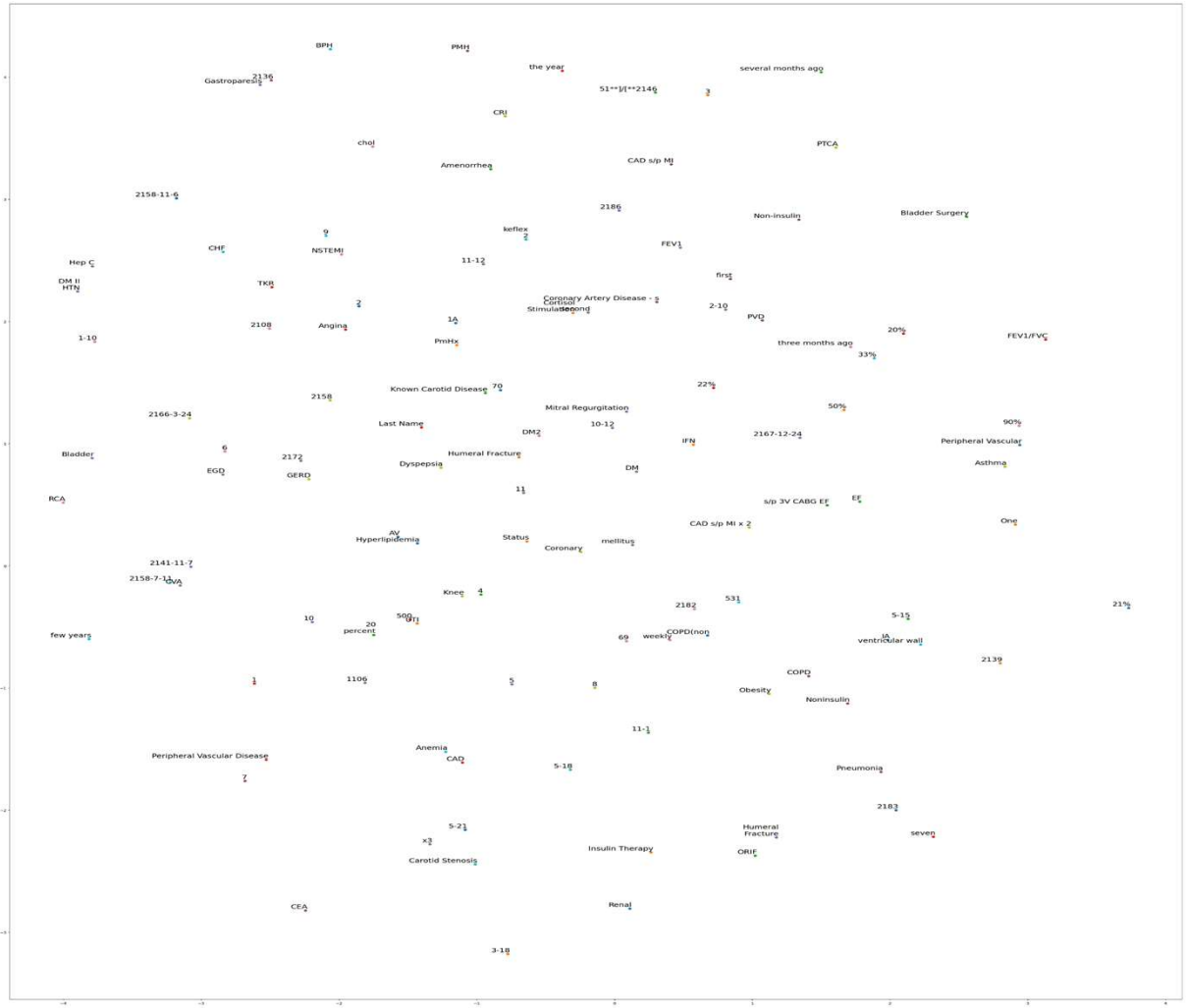

nmod
case
compound
compound
nummod
Prostate
NOUN
CA
PROPN
on
ADP
Lupron
PROPN
SPACE
10.
NUM

Gout
NOUN

Spacy TSNE visualization: For Medication in medical notes

https://github.com/dalwari/mimic-iii-clinical-database-demo-1.4/blob/main/spacy_tsne_plot_diag.png

```
df =pd.read_csv("nlp_med_notes.csv")['text']
label_df=do_data_process(df,'MEDICATIONS')
```



SciSpacy Usage:

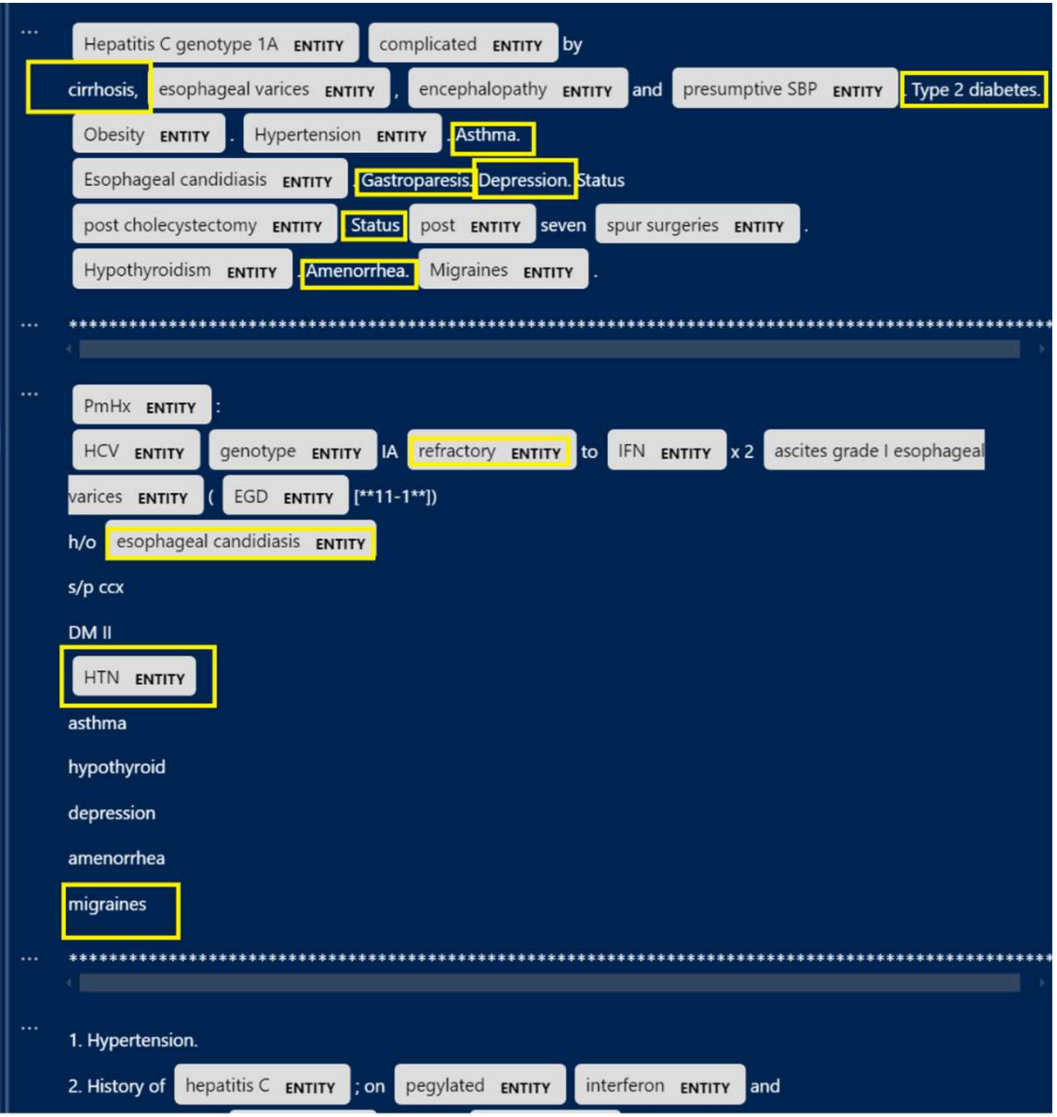
- SciSpacy models:
- en_core_sci_md
- en_core_sci_lg
- en_ner_craft_md
- en_ner_jnlpba_md
- en_ner_bionlp13cg_md
- en_ner_bc5cdr_md

- Load the model and visualize the corpus.

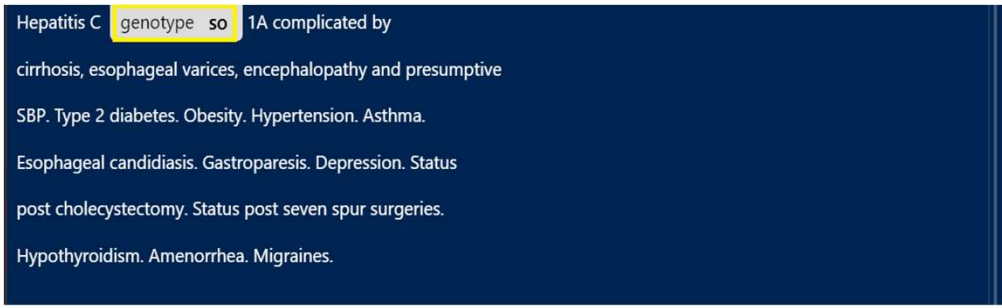
```
import en_ner_craft_md
nlp = en_ner_craft_md.load()
visualize_entities(label_df)
```

Sciscpacy: ***en_core_sci_lg***

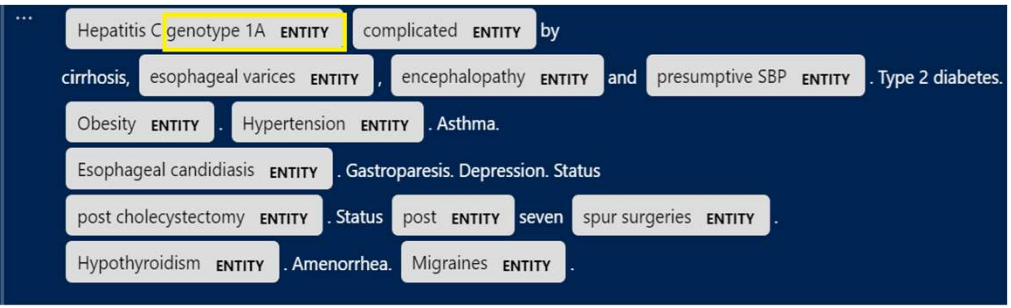
Spacy:

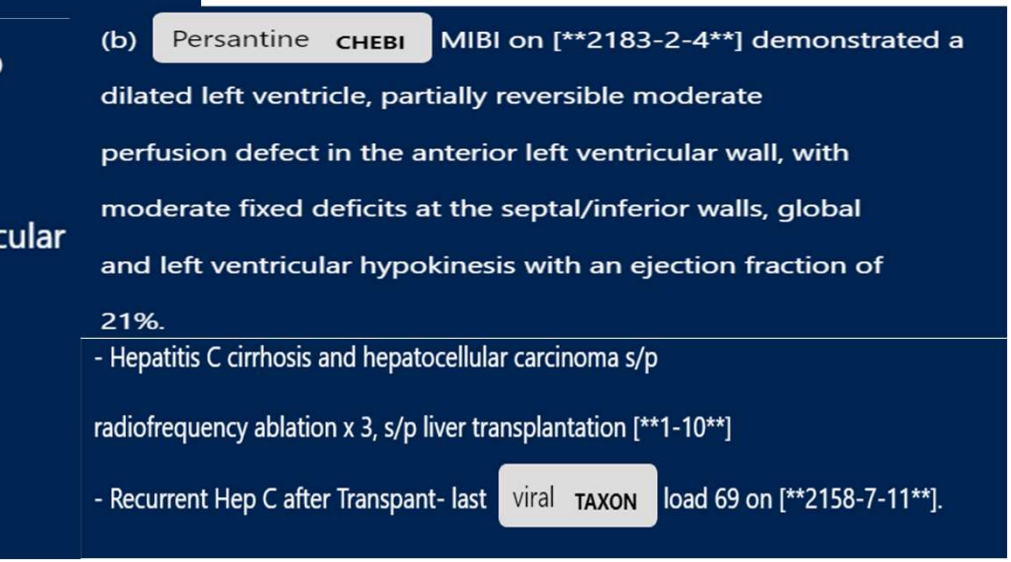


Scipacy: ***en_ner_craft_md***



Spacy:

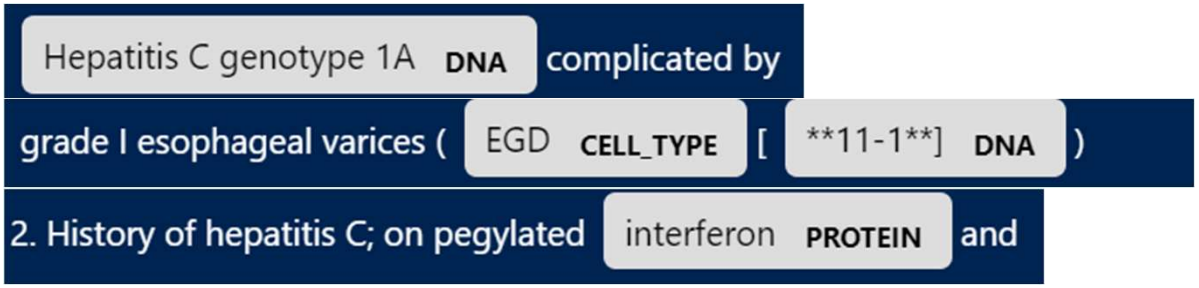




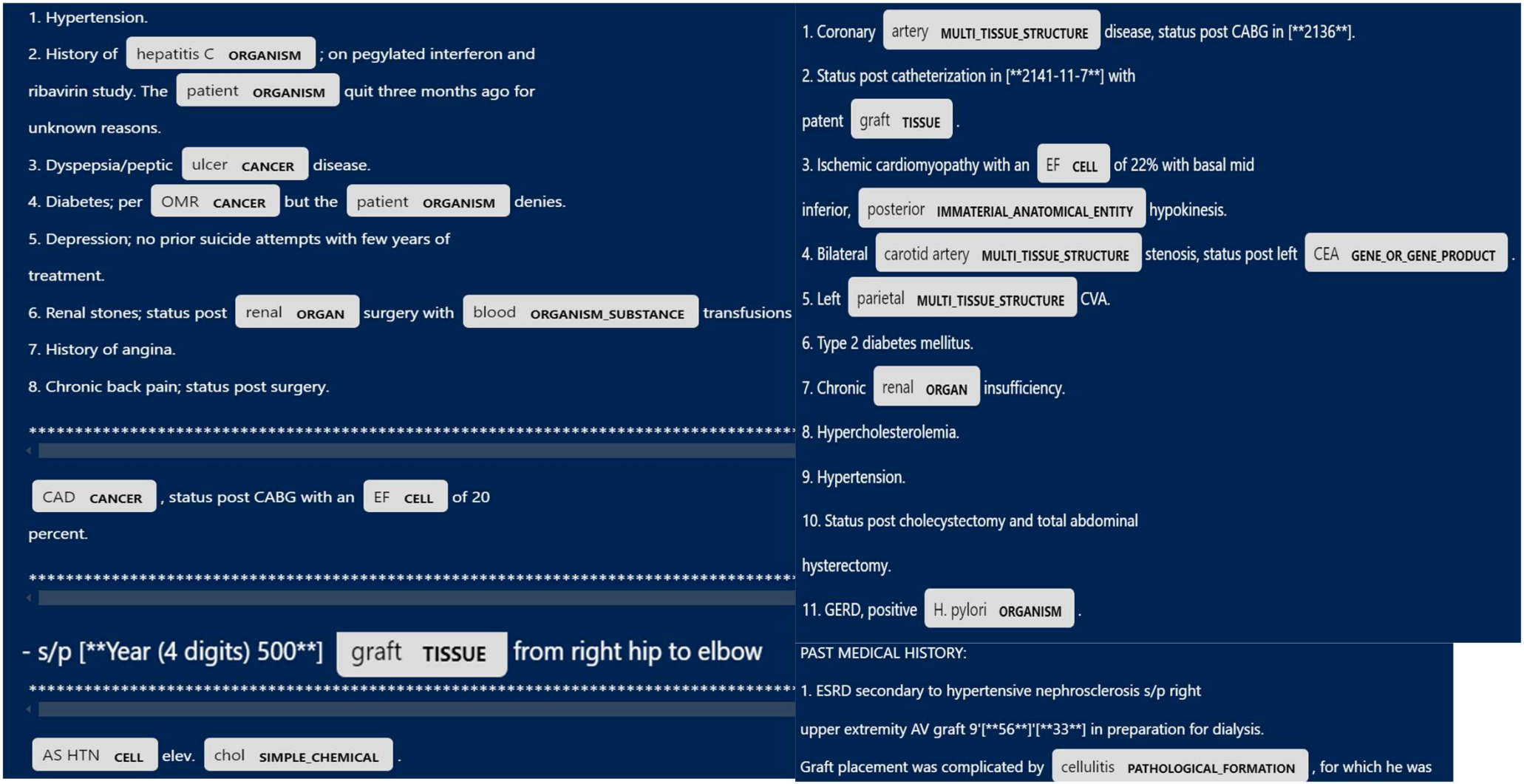
Scipacy: ***en_ner_craft_md***

- **SO** stands for Sequence Ontology
- **CL** stands for Cell Ontology
- **CHEBI** stands for Chemical Entities of Biological Interest
- **TAXON** stands for Taxonomy
- **GCP** stands for Gene Ontology Cellular Component

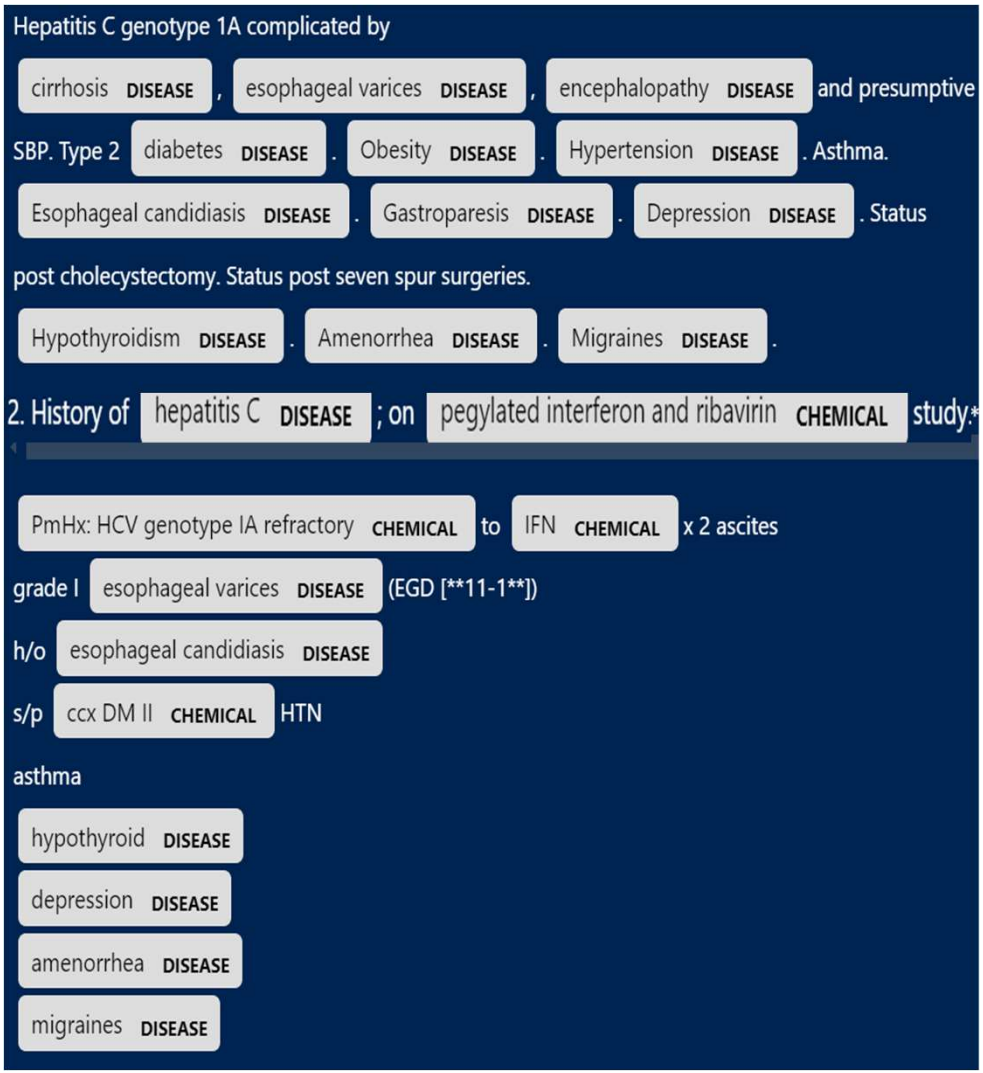
Scipacy: ***en_ner_jnlpba_md***



Scipacy: ***en_ner_bionlp13cg_md***



Scipacy: ***en_ner_bc5cdr_md***

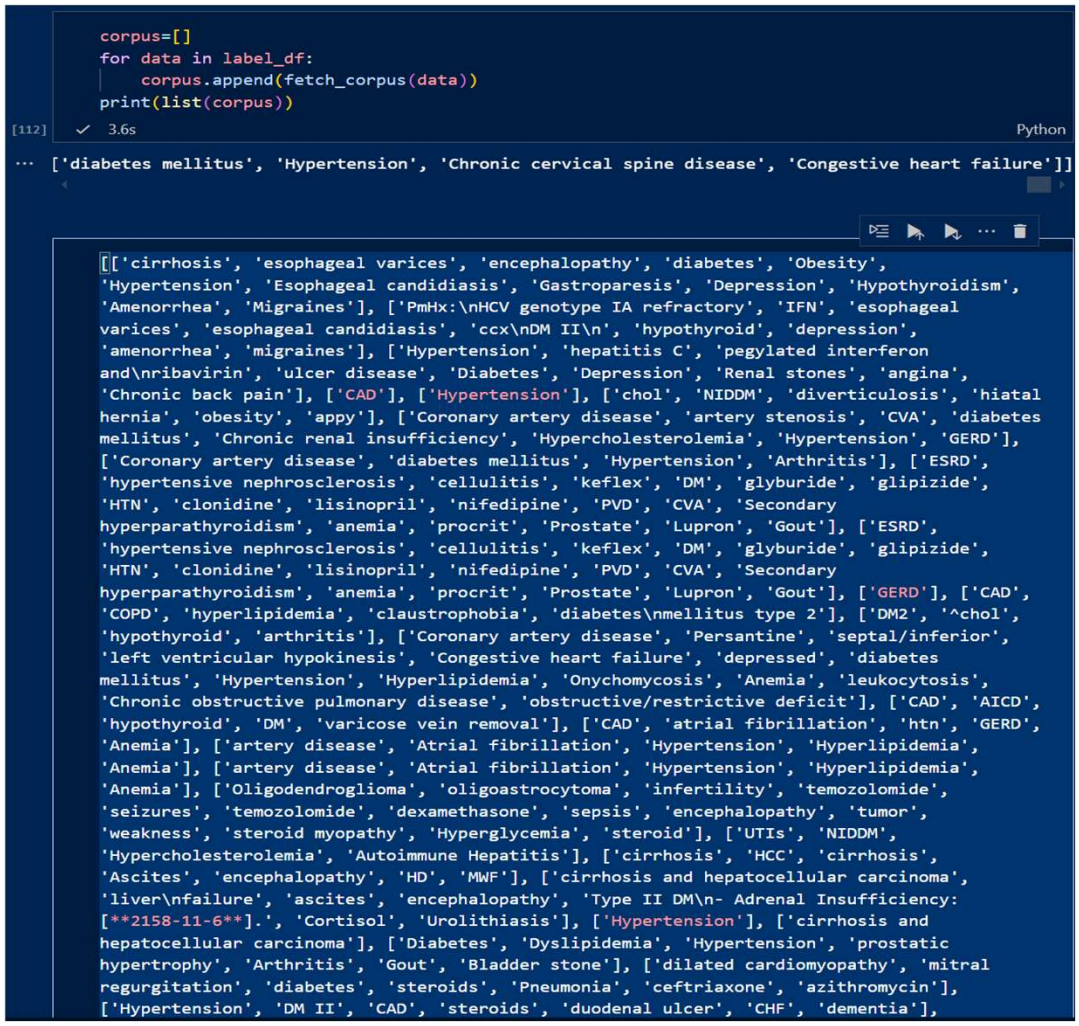


Conclusion:

With the help of different spacy models we are able to extract different kinds of NERs.

Word2Vec and TSNE plot usage:

Apply **fetch_corpus()** to retrieve the list of corpus from the medical note.



Code Snippet:

The resultant presents the similarity logits for the word ‘encephalopathy’

```
model1 = Word2Vec(corpus, min_count=1)
model1.wv['encephalopathy']
```

✓ 0.0s

```
array([-8.7531786e-03,  2.1741530e-03, -8.6094369e-04, -9.3106795e-03,
       -9.4064260e-03, -1.4538610e-03,  4.4581434e-03,  3.7536507e-03,
       -6.5508662e-03, -6.8758638e-03, -5.0241956e-03, -2.3389754e-03,
       -7.2221956e-03, -9.5775630e-03, -2.7493779e-03, -8.3579253e-03,
       -6.0137236e-03, -5.7307631e-03, -2.3647691e-03, -1.7745970e-03,
       -8.9362776e-03, -6.9540367e-04,  8.1498129e-03,  7.6987552e-03,
       -7.2270953e-03, -3.6619261e-03,  3.0702241e-03, -9.5547633e-03,
        1.4801961e-03,  6.5093059e-03,  5.7971054e-03, -8.7778289e-03,
       -4.5126704e-03, -8.1743700e-03,  3.6444104e-05,  9.3236137e-03,
        5.9737498e-03,  5.0418158e-03,  5.0477851e-03, -3.3005176e-03,
        9.5378207e-03, -7.3622023e-03, -7.3122410e-03, -2.2796686e-03,
       -7.5115956e-04, -3.1877523e-03, -6.4017362e-04,  7.4983523e-03,
       -6.7837693e-04, -1.5929459e-03,  2.7603914e-03, -8.3850855e-03,
        7.8556603e-03,  8.5417535e-03, -9.6132429e-03,  2.4651806e-03,
        9.9031590e-03, -7.6433863e-03, -6.9885631e-03, -7.6803914e-03,
        8.3996654e-03, -6.9426361e-04,  9.1576520e-03, -8.1540635e-03,
        3.7199876e-03,  2.6663735e-03,  7.5992203e-04,  2.3442844e-03,
       -7.5090886e-03, -9.2971604e-03,  2.3168572e-03,  6.1675226e-03,
        8.0000097e-03,  5.6976336e-03, -7.6059706e-04,  8.2836589e-03,
       -9.3513643e-03,  3.3959236e-03,  2.5762038e-04,  3.8506044e-03,
        7.3216450e-03, -6.7115389e-03,  5.5358177e-03, -9.4783595e-03,
       -8.4873615e-04, -8.6890254e-03, -5.0572841e-03,  9.3041677e-03,
       -1.8036201e-03,  2.8908700e-03,  9.0945661e-03,  8.9400755e-03,
       -8.2035558e-03, -3.0187166e-03,  9.9292845e-03,  5.0835693e-03,
       -1.5810047e-03, -8.7010888e-03,  2.9685348e-03, -6.6635790e-03],
      dtype=float32)
```

The code snippet shows the similar words that the model contains within it.

```
model1.wv.similar_by_word('encephalopathy')
```

✓ 0.0s

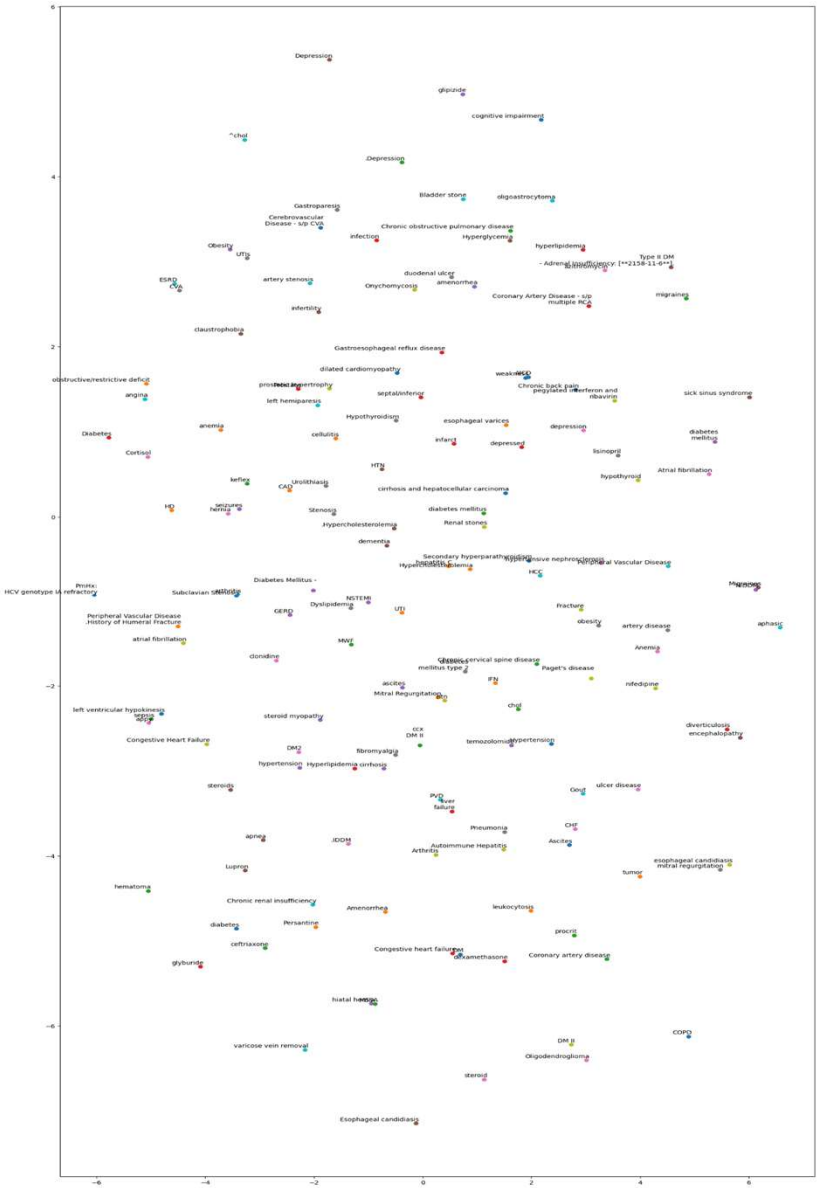
```
[('diverticulosis', 0.30835863947868347),
 ('Renal stones', 0.2804599106311798),
 ('esophageal candidiasis', 0.2360624074935913),
 ('Hyperglycemia', 0.20393291115760803),
 ('Fracture', 0.2022656798362732),
 ('Persantine', 0.19107292592525482),
 ('Gastroparesis', 0.17959770560264587),
 ('Hypercholesterolemia', 0.168697327375412),
 ('Secondary hyperparathyroidism', 0.16342854499816895),
 ('Peripheral Vascular Disease', 0.14657379686832428)]
```

TSNE plot:

```
vocabs = model1.wv.key_to_index.keys()
new_v = list(vocabs)
tsne_plot(model1, new_v)
```

The corpus are plotted in tsne word embedding as per their logits.

SciSpacy TSNE plot:
https://github.com/dalwari/mimic-iii-clinical-database-demo-1.4/blob/main/tsne_plot_diag.png


Depression
glipizide
cognitive impairment
^chol
,Depression
Bladder stone
oligoastrocytoma
Gastroparesis
Cerebrovascular Disease - s/p CVA
Chronic obstructive pulmonary disease
infection
Hyperglycemia
hyperlipidemia
Obesity
UTIs
Type II DM
- Adrenal insufficiency: [**2138-11-6**]
hemodialysis
ESRD
CVA
artery stenosis
duodenal ulcer
Onychomycosis
amenorrhea
Coronary Artery Disease - s/p multiple RCA
migraines
infertility
claustrophobia
Gastroesophageal reflux disease
dilated cardiomyopathy
weakness
obstructive/restrictive defect
Chronic back pain
pegylated interferon and ribavirin
sick sinus syndrome
angina
septal/inferior
left hemiparesis
Hypothyroidism
esophageal varices
anemia
depression
Diabetes
cellulitis
infarct
depressed
diabetes mellitus
lisinopril
Cortisol
HTN
hypothyroid
Atrial fibrillation
reflux
CAD
cirrhosis and hepatocellular carcinoma
HD
seizures
hernia
Stenosis
diabetes mellitus
Renal stones
,Hypercholesterolemia
dementia
HCC
Peripheral Vascular Disease
Diabetes Mellitus -
NSTEMI
Dyslipidemia
UTI
Fracture
obesity
artery disease
aphasic
atrial fibrillation
MWF
Anemia
clonidine
Paget's disease
nifedipine
ascites
IFN
Mitral Regurgitation
left ventricular hypokinesis
steroid myopathy
CHF
diverticulosis
encephalopathy
Congestive Heart Failure
DM2
temozolomide
hypertension
fibromyalgia
hypertension
Hyperlipidemia
cirrhosis
steroids
Gout
ulcer disease
PVD
liver failure
Pneumonia
CHF
apnea
IDDM
Ascites
Autoimmune Hepatitis
Arthritis
esophageal candidiasis
mitral regurgitation III
tumor
Lupron
hematoma
Chronic renal insufficiency
Amenorrhea
leukocytosis
diabetes
Persantine
procrit
ceftriaxone
Congestive heart failure
dexamethasone
Coronary artery disease
glyburide
hiatal hernia
COPD
varicose vein removal
Oligodendroglioma
steroid
Esophageal candidiasis

Clinical BERT:

- Import transformers and classification package.

```
from transformers import AutoTokenizer,
AutoModelForTokenClassification
from transformers import pipeline
```

- Initialize model.

```
tokenizer = AutoTokenizer.from_pretrained("emilyalsentzer/Bio_ClinicalBERT")
model = AutoModelForTokenClassification.from_pretrained("emilyalsentzer/Bio_ClinicalBERT")
nlp = pipeline("ner", model=model, tokenizer=tokenizer)
```

- Load dataset and examine 'PHYSICAL EXAMINATION' sub passage.

- The output displays label and score for each word in a sentence.
- The label is categorized by ClinicalBERT into LABEL_0 and LABEL_1
- The score is given to each word in that label category.

```
result=[]
for data in label_df:
    value=nlp(data)
    print(value)
```

✓ 0.4s

```
[{'entity': 'LABEL_0', 'score': 0.6019393, 'index': 1, 'word': 'the', 'start': 0, 'end': 3}, {
[{'entity': 'LABEL_0', 'score': 0.5833172, 'index': 1, 'word': 'the', 'start': 0, 'end': 3}, {
[{'entity': 'LABEL_1', 'score': 0.524752, 'index': 1, 'word': 'vital', 'start': 0, 'end': 5},
[{'entity': 'LABEL_1', 'score': 0.59145355, 'index': 1, 'word': 'vs', 'start': 0, 'end': 2}, {
[{'entity': 'LABEL_1', 'score': 0.60578763, 'index': 1, 'word': 'height', 'start': 0, 'end': 6
[{'entity': 'LABEL_0', 'score': 0.5948016, 'index': 1, 'word': '-', 'start': 0, 'end': 1}, {'e
[{'entity': 'LABEL_1', 'score': 0.52503663, 'index': 1, 'word': 'vital', 'start': 0, 'end': 5}
[{'entity': 'LABEL_1', 'score': 0.5799533, 'index': 1, 'word': 'vital', 'start': 0, 'end': 5},
```

ClinicalBERT TSNE visualization: For Medication in medical notes

https://github.com/dalwari/mimic-iii-clinical-database-demo-1.4/blob/main/clinical_bert_tsne_plot.png