

Name_entity_classifier

May 10, 2022

0.1 my approach is to build a custom NER model for store number

0.2 Custom NER using Spacy

- **Features Extraction:** From every transaction_descriptor I extract the first and last index the store number appear in. So, the input to the model is (start_index,end_index,label). in our case the label is "StoreNumber"
- **Data preparation:** As I use spacy model(Rule based model) to be trained in our data. The model accept only *.spacy* files. So, I convert to .spacy

```
[ ]: !pip install spacy==3.0.6
```

```
[4]: from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
[6]: import pandas as pd
import re
import spacy
from spacy.tokens import DocBin
from tqdm import tqdm
from spacy.util import filter_spans
import numpy as np

df=pd.read_csv("/content/drive/MyDrive/Summer Internship - Homework Exercise.
→csv")
df
```

```
[6]: transaction_descriptor store_number dataset
0    DOLRTREE 2257 00022574 ROSWELL      2257  train
1              AUTOZONE #3547      3547  train
2          TGI FRIDAYS 1485 0000      1485  train
3    BUFFALO WILD WINGS 003          3    train
4              J. CREW #568 0        568  train
..              ...            ...    ...
295          MCDONALD'S F2151      F2151  test
```

296	NST BEST BUY #1403 332411	1403	test
297	CVS/PHARMACY #06689	6689	test
298	BANANA REPUBLIC #8109	8109	test
299	BOSTON MARKET 0443	443	test

[300 rows x 3 columns]

0.3 Data Splinging

```
[7]: df_train=df[df["dataset"]=="train"]
df_valid=df[df["dataset"]=="validation"]
df_test=df[df["dataset"]=="test"]

df_train
```

```
[7]:      transaction_descriptor store_number dataset
0  DOLRTREE 2257 00022574 ROSWELL      2257  train
1      AUTOZONE #3547      3547  train
2      TGI FRIDAYS 1485 0000      1485  train
3  BUFFALO WILD WINGS 003          3  train
4      J. CREW #568 0          568  train
..      ...      ...      ...
95      DUNKIN #355514      355514  train
96  THE HOME DEPOT #6828      6828  train
97      RUE21 #1129 BLUE      1129  train
98      WM SUPERCENTER #34          34  train
99  RACETRAC485 00004853      485  train
```

[100 rows x 3 columns]

```
[28]: print(df_train["store_number"].apply(lambda x : len(x)).max())
print(df_valid["store_number"].apply(lambda x : len(x)).max())
print(df_test["store_number"].apply(lambda x : len(x)).max())
```

9
9
9

```
[ ]: #downlad it for utils GPU during training
!pip install spacy[transformers]
```

0.3.1 Feature Extraction

```
[19]: def extract_text_spans(df):
      L = []
      for i in range(len(df)):
          S = df['transaction_descriptor'].iloc[i]
          w = df['store_number'].iloc[i]
          match = re.search( w, S)
          start = match.start()
          end = match.end()
          l = [S ,w ,start , end]
          L.append(l)

      data = pd.DataFrame( L , columns=['transaction_descriptor', 'store_number' ,
      ↪ 'start', 'end'])
      return data
```

```
[35]: train=extract_text_spans(df_train)
      valid=extract_text_spans(df_valid)
```

```
[21]: train
```

```
[21]:
```

	transaction_descriptor	store_number	start	end
0	DOLRTREE 2257 00022574 ROSWELL	2257	9	13
1	AUTOZONE #3547	3547	10	14
2	TGI FRIDAYS 1485 0000	1485	12	16
3	BUFFALO WILD WINGS 003	3	21	22
4	J. CREW #568 0	568	9	12
..
95	DUNKIN #355514	355514	8	14
96	THE HOME DEPOT #6828	6828	16	20
97	RUE21 #1129 BLUE	1129	7	11
98	WM SUPERCENTER #34	34	16	18
99	RACETRAC485 00004853	485	8	11

[100 rows x 4 columns]

```
[22]: nlp = spacy.blank("en")
      doc_bin = DocBin()
```

0.3.2 Data preparation

```
[23]: def create_spacy_file(df, file_name):
        skipped=0
        for i in tqdm(range(len(df))):
            text = df["transaction_descriptor"].iloc[i]
            doc = nlp.make_doc(text)
            ents = []
            span = doc.char_span(df['start'].iloc[i], df['end'].iloc[i], label='␣
↪ 'StoreNumber', alignment_mode="contract")
            if span is None:
                skipped+=1
            else:
                ents.append(span)
            filtered_ents = filter_spans(ents)
            doc.ents = filtered_ents
            doc_bin.add(doc)

        doc_bin.to_disk(f"{file_name}.spacy")
        print(f"\nskipped {file_name} entities={skipped}\n")
```

```
[24]: create_spacy_file(train, "train")
        create_spacy_file(valid, "validation")
```

```
100%|      | 100/100 [00:00<00:00, 1176.38it/s]
```

```
skipped train entities=34
```

```
100%|      | 100/100 [00:00<00:00, 1100.89it/s]
```

```
skipped validation entities=39
```

0.3.3 Training the Model

```
[25]: !python -m spacy init config --lang en --pipeline ner config.cfg --force
```

Generated config template specific for your use case

- Language: en
- Pipeline: ner
- Optimize for: efficiency
- Hardware: CPU
- Transformer: None

```

Auto-filled config with all values
Saved config
config.cfg
You can now add your data and train your pipeline:
python -m spacy train config.cfg --paths.train ./train.spacy --paths.dev
./dev.spacy

```

```
[26]: !python -m spacy train config.cfg --output ./ --paths.train ./train.spacy
      ↪--paths.dev ./train.spacy
```

```

Using CPU
To switch to GPU 0, use the option: --gpu-id 0

```

```

===== Initializing pipeline
=====
[2022-05-10 03:46:07,361] [INFO] Set up nlp object from config
[2022-05-10 03:46:07,373] [INFO] Pipeline: ['tok2vec', 'ner']
[2022-05-10 03:46:07,378] [INFO] Created vocabulary
[2022-05-10 03:46:07,378] [INFO] Finished initializing nlp object
[2022-05-10 03:46:07,585] [INFO] Initialized pipeline components: ['tok2vec',
'ner']
    Initialized pipeline

```

```

===== Training pipeline
=====
    Pipeline: ['tok2vec', 'ner']
    Initial learn rate: 0.001

```

E	#	LOSS TOK2VEC	LOSS NER	ENTS_F	ENTS_P	ENTS_R	SCORE
0	0	0.00	64.00	0.00	0.00	0.00	0.00
50	200	3.31	727.12	100.00	100.00	100.00	1.00
113	400	0.00	0.00	100.00	100.00	100.00	1.00
180	600	0.00	0.00	100.00	100.00	100.00	1.00
277	800	0.00	0.00	100.00	100.00	100.00	1.00
377	1000	0.00	0.00	100.00	100.00	100.00	1.00
477	1200	0.00	0.00	100.00	100.00	100.00	1.00
670	1400	0.00	0.00	100.00	100.00	100.00	1.00
870	1600	50.08	11.66	100.00	100.00	100.00	1.00
1070	1800	19.81	3.98	100.00	100.00	100.00	1.00

```

    Saved pipeline to output directory
model-last

```

```
[27]: !python -m spacy train config.cfg --output ./ --paths.train ./validation.spacy
      ↪--paths.dev ./train.spacy
```

```
Using CPU
```

To switch to GPU 0, use the option: `--gpu-id 0`

===== Initializing pipeline

=====

```
[2022-05-10 03:50:46,379] [INFO] Set up nlp object from config
[2022-05-10 03:50:46,391] [INFO] Pipeline: ['tok2vec', 'ner']
[2022-05-10 03:50:46,396] [INFO] Created vocabulary
[2022-05-10 03:50:46,396] [INFO] Finished initializing nlp object
[2022-05-10 03:50:46,691] [INFO] Initialized pipeline components: ['tok2vec',
'ner']
```

Initialized pipeline

===== Training pipeline

=====

Pipeline: ['tok2vec', 'ner']

Initial learn rate: 0.001

E	#	LOSS TOK2VEC	LOSS NER	ENTS_F	ENTS_P	ENTS_R	SCORE
0	0	0.00	62.83	0.00	0.00	0.00	0.00
25	200	9.21	726.38	100.00	100.00	100.00	1.00
55	400	0.00	0.00	100.00	100.00	100.00	1.00
92	600	0.00	0.00	100.00	100.00	100.00	1.00
137	800	0.00	0.00	100.00	100.00	100.00	1.00
189	1000	0.00	0.00	100.00	100.00	100.00	1.00
256	1200	88.97	24.24	100.00	100.00	100.00	1.00
332	1400	41.58	11.53	100.00	100.00	100.00	1.00
432	1600	0.00	0.00	100.00	100.00	100.00	1.00
532	1800	0.00	0.00	100.00	100.00	100.00	1.00

Saved pipeline to output directory

model-last

0.3.4 Testing model and visualize the results

```
[37]: nlp_ner = spacy.load("model-best")
preds=[]
matched=0
for i in range(len(df_test)):
    doc = nlp_ner(df_test["transaction_descriptor"].iloc[i])
    colors = {"StoreNumber": "#7DF6D9"}
    options = {"colors": colors}
    if len(doc.ents) >= 1 and doc.ents[0].label_ == "StoreNumber":
        preds.append(str(doc.ents[0]))
        doc.ents[0].label_ == "StoreNumber"
    spacy.displacy.render(doc, style="ent", options= options, jupyter=True)
```

```
        matched+=1
    else:
        preds.append('Not matched')
```

<IPython.core.display.HTML object>

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```
[30]: print(f"extracted True sample = {matched}\nNot Extracted sample =   
      ↳{100-matched}")
```

extracted True sample = 72

Not Extracted sample = 28

0.3.5 the model extract 72 store number out of 100 entries

```
[38]: np.sum(np.equal(df_test['store_number'], preds))
```

```
[38]: 71
```

0.3.6 71 of 72 extracted number are correctly identified as Store number

```
[33]: test_preds = pd.DataFrame({'test' : df_test['store_number'] , 'preds': preds})
test_preds
```

```
[33]:
```

	test	preds
200	242	242
201	9442088	Not matched
202	1419	1419
203	1019	1019
204	38	38
..
295	F2151	F2151
296	1403	1403
297	6689	Not matched
298	8109	8109
299	443	Not matched

```
[100 rows x 2 columns]
```

0.4 model accuracy

```
[34]: import numpy as np
acc = np.sum(np.equal(df_test['store_number'], preds)) /
↳ len(df_test['store_number'])
acc
```

```
[34]: 0.71
```

```
[ ]:
```

```
[ ]: !wget -nc https://raw.githubusercontent.com/brpy/colab-pdf/master/colab_pdf.py
from colab_pdf import colab_pdf
colab_pdf('Name_entity_classifier.ipynb')
```

File 'colab_pdf.py' already there; not retrieving.

WARNING: apt does not have a stable CLI interface. Use with caution in scripts.

WARNING: apt does not have a stable CLI interface. Use with caution in scripts.

Extracting templates from packages: 100%