

# Baseline\_model

September 17, 2020

```
[1]: import numpy as np
import pandas as pd

from baseline.torch.mydataset import CustomDataset
from baseline.torch.model import CNN_Text
from baseline.torch.train import train, predict_prob, run

import torch
from torchtext import data
import time
from tqdm import tqdm
```

## 1 load data

```
[2]: data_root = './data/'
```

```
[3]: drugs = ['trametinib',
              'fulvestrant',
              'lovastatin',
              'abiraterone',
              'thalidomide',
              'sirolimus',
              'simvastatin',
              'methotrexate',
              'bortezomib',
              'gemcitabine',
              'tamoxifen',
              'dexamethasone',
              'doxorubicin']

len(drugs)
```

```
[3]: 13
```

```
[4]: alllab_df = pd.read_csv(data_root+'lab_finfin.csv')
alllab_df.head()
```

```
[4]:
```

	drug	file	lab	comment
0	tamoxifen	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3000794.xml	0	none
1	tamoxifen	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3005955.xml	0	none
2	tamoxifen	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3010527.xml	0	none
3	tamoxifen	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3011858.xml	0	none
4	tamoxifen	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3014261.xml	0	none

```
[11]: allfea_df = pd.read_csv(data_root+'fea_finfin.csv')
allfea_df.head()
```

```
[11]:
```

	file	title	abstract
0	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3000794.xml		
1	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3001231.xml		
2	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3003872.xml		
3	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3004744.xml		
4	PMC003XXXXXX.xml\PMC0030XXXXXX\PMC3005850.xml		

  

	title	abstract
0	erk1 2 dependent vascular endothelial growth f...	background amp aims severe polycystic liver di...
1	peg functionalized magnetic nanoparticles for ...	purpose polyethylene glycol ( peg ) functional...
2	combination testing \ ( stage 2 \ ) of rapamycin...	purpose rapamycin demonstrated broad spectrum ...
3	durable responses with the metronomic regimen ...	background targeting the tumor microenvironmen...
4	ph sensitive ionomeric particles obtained via ...	silk fibroin based biomaterials have been wide...

```
[12]: train_lab_df = alllab_df[alllab_df['drug'].isin(drugs[:8])].
      ↪reset_index(drop=True)
test_lab_df = alllab_df[alllab_df['drug'].isin(drugs[8:])].
      ↪reset_index(drop=True)
```

```
[13]: len_0 = len(train_lab_df[train_lab_df['lab']==0])
len_1 = len(train_lab_df[train_lab_df['lab']==1])
ratio = (len_0 - len_1)/len_1
```

```
[14]: train_lab_df_tmp = train_lab_df
for _ in range(int(ratio)):
    train_lab_df_tmp = pd.concat([train_lab_df_tmp,
    ↪train_lab_df[train_lab_df['lab']==1]], ignore_index=True)
train_lab_df = train_lab_df_tmp
```

```
[15]: len(train_lab_df[train_lab_df['lab']==0]),  
      ↪ len(train_lab_df[train_lab_df['lab']==1])
```

```
[15]: (847, 828)
```

## 2 Build Vocabulary

```
[16]: start_t = time.time()  
  
text_field = data.Field(lower=True) # Text field  
label_field = data.Field(sequential=False) # Label field  
  
train_data, dev_data = CustomDataset.splits(text_field, label_field,  
      ↪ train_lab_df, allfea_df, shuffle=True)  
test_data = CustomDataset(text_field, label_field, test_lab_df, allfea_df)  
  
end_t = time.time()-start_t  
print("Time elapse (min): ", end_t/60)
```

```
4%|          | 69/1675 [00:00<00:02, 688.49it/s]  
preparing examples...  
100%|         | 1675/1675 [00:02<00:00, 567.31it/s]  
6%|          | 148/2681 [00:00<00:03, 741.43it/s]  
  
dev_index: -167  
preparing examples...  
100%|         | 2681/2681 [00:03<00:00, 729.41it/s]  
Time elapse (min): 0.11108090082804362
```

```
[21]: batch_size = 32  
text_field.build_vocab(train_data, dev_data, test_data)  
label_field.build_vocab(train_data, dev_data, test_data)  
train_iter, dev_iter = data.Iterator.splits((train_data, dev_data),  
      batch_sizes=(batch_size,  
      ↪ len(dev_data)))
```

## 3 Run the Baseline Model

```
[22]: fields = [('text', text_field), ('label', label_field)]
```

```
[23]: model_dir_root = './trained_models/'
```

```
[24]: run(CNN_Text, model_dir_root+'model_baseline_fin.pkl', drugs[8:], alllab_df,
        ↪allfea_df, fields)
```

```
6%|          | 24/371 [00:00<00:01, 234.32it/s]
```

```
#####
```

```
#### drug bortezomib
```

```
100%|         | 371/371 [00:01<00:00, 242.17it/s]
```

```
7%|          | 28/415 [00:00<00:01, 272.90it/s]
```

```
The first paper is PMC004XXXXXX.xml\PMC0042XXXXX\PMC4266584.xml
```

```
Number of papers be read of drug [bortezomib]: 119
```

```
#####
```

```
#### drug gemcitabine
```

```
100%|         | 415/415 [00:01<00:00, 254.81it/s]
```

```
5%|          | 28/526 [00:00<00:01, 279.68it/s]
```

```
The first paper is PMC004XXXXXX.xml\PMC0048XXXXX\PMC4873426.xml
```

```
Number of papers be read of drug [gemcitabine]: 5
```

```
#####
```

```
#### drug tamoxifen
```

```
100%|         | 526/526 [00:01<00:00, 263.32it/s]
```

```
5%|          | 28/565 [00:00<00:01, 278.02it/s]
```

```
The first paper is PMC003XXXXXX.xml\PMC0037XXXXX\PMC3711713.xml
```

```
Number of papers be read of drug [tamoxifen]: 46
```

```
#####
```

```
#### drug dexamethasone
```

```
100%|         | 565/565 [00:02<00:00, 267.28it/s]
```

```
4%|          | 29/804 [00:00<00:02, 282.54it/s]
```

```
The first paper is PMC004XXXXXX.xml\PMC0044XXXXX\PMC4422178.xml
```

```
Number of papers be read of drug [dexamethasone]: 141
```

```
#####
```

```
#### drug doxorubicin
```

```
100%|         | 804/804 [00:02<00:00, 271.49it/s]
```

```
The first paper is PMC003XXXXXX.xml\PMC0032XXXXX\PMC3298037.xml
```

```
Number of papers be read of drug [doxorubicin]: 384
```

```
[24]: ({'bortezomib': 119,
        'gemcitabine': 5,
        'tamoxifen': 46,
        'dexamethasone': 141,
        'doxorubicin': 384},
        {'bortezomib': 'PMC004XXXXXX.xml\\PMC0042XXXXX\\PMC4266584.xml',
```

```
'gemcitabine': 'PMC004XXXXXX.xml\\PMC0048XXXXX\\PMC4873426.xml',  
'tamoxifen': 'PMC003XXXXXX.xml\\PMC0037XXXXX\\PMC3711713.xml',  
'dexamethasone': 'PMC004XXXXXX.xml\\PMC0044XXXXX\\PMC4422178.xml',  
'doxorubicin': 'PMC003XXXXXX.xml\\PMC0032XXXXX\\PMC3298037.xml'})
```