Matheus_Schmitz_hw7_task_2

optimizer='adam',

loss='multiclass_nll',
regularizer='LP',

optimizer params={'lr':1e-3},

regularizer params={'p':3, 'lambda':1e-5},

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```
In [7]:
# File manipulation imports for Google Colab
from google.colab import drive
drive.mount('/content/drive')
import os
os.chdir("/content/drive/My Drive/Colab Notebooks/DSCI 558")
Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount
("/content/drive", force remount=True).
In [8]:
!pip install -q ampligraph
!pip install -q tensorflow==1.15
In [9]:
import requests
from ampligraph.datasets import load from csv
import numpy as np
import pandas as pd
Train-Test Split
In [10]:
from ampligraph.evaluation import train test split no unseen
X = load from csv('', 'freebase-237-merged-and-remapped.csv', sep=',')
print(f'X.shape: {X.shape}')
X train, X test = train test split no unseen(X, test size=10000)
print('Train set size: ', X train.shape)
print('Test set size: ', X test.shape)
X.shape: (308722, 3)
Train set size: (298722, 3)
Test set size: (10000, 3)
In [11]:
# ComplEx was the best performing model in task 1
from ampligraph.latent features import ComplEx
from ampligraph.evaluation import evaluate performance
import tensorflow as tf
tf.logging.set verbosity(tf.logging.ERROR)
from ampligraph.evaluation import mr score, mrr score, hits at n score
model = ComplEx(batches count=100,
                seed=0,
                epochs=200,
                k = 300,
                eta=5,
```

```
verbose=True)
positives filter = X
tf.logging.set verbosity(tf.logging.ERROR)
model.fit(X train, early stopping = False)
ranks = evaluate performance(X test,
                             model=model,
                             filter triples=positives filter, # Corruption strategy fi
Iter defined above
                             use default protocol=True, # corrupt subj and obj separatel
y while evaluating
                             verbose=True)
mrr = mrr score(ranks)
print("MRR: %.2f" % (mrr))
hits_10 = hits_at_n_score(ranks, n=10)
print("Hits@10: %.2f" % (hits 10))
hits_3 = hits_at_n_score(ranks, n=3)
print("Hits@3: %.2f" % (hits 3))
hits 1 = hits at n score(ranks, n=1)
print("Hits@1: %.2f" % (hits 1))
Average Loss: 0.027143: 100% | 200/200 [07:10<00:00, 2.15s/epoch]
WARNING - DeprecationWarning: use default protocol will be removed in future. Please use
corrupt side argument instead.
       | 10000/10000 [01:52<00:00, 88.62it/s]
100%|
MRR: 0.21
Hits@10: 0.37
Hits@3: 0.24
Hits@1: 0.13
In [12]:
# Randomly select 10 different triples to run the prediction and show the predicted resul
t in your notebook using the format below
random_test_indexes = np.random.choice(X test.shape[0], 10)
X unseen = X test[random test indexes]
unseen filter = np.array(list({tuple(i) for i in np.vstack((positives filter, X unseen))
}))
ranks unseen = evaluate performance(
   X unseen,
   model=model,
   filter triples=unseen filter, # Corruption strategy filter defined above
    corrupt side = 's+o',
    use default protocol=False, # corrupt subj and obj separately while evaluating
    verbose=True
scores = model.predict(X unseen)
from scipy.special import expit
probs = expit(scores)
import pandas as pd
pd.DataFrame(list(zip([' '.join(x) for x in X_unseen],
                      ranks unseen,
                      np.squeeze(scores),
                      np.squeeze(probs))),
             columns=['statement', 'rank', 'score', 'prob']).sort_values("score")
100%|
          | 10/10 [00:00<00:00, 55.70it/s]
```

Out[12]:

	statement	rank	score	prob
9	bafta award for best film /award/award_categor	1067	4.594365	0.989993
7	europe /location/location/contains czech republic	1213	5.777078	0.996912
4	don cheadle /award/award_winner/awards_won./aw	267	8.492310	0.999795
5	kathleen quinlan /people/person/place_of_birth	2015	12.807221	0.999997
1	harmonica /music/performance_role/regular_perf	39	14.166807	0.999999
8	crunk /music/genre/artists lil jon	16	15.645445	1.000000
6	soprano saxophone /music/performance_role/trac	30	15.965117	1.000000
3	university of delaware /education/educational	551	16.134884	1.000000
0	jeffrey jones /award/award_nominee/award_nomin	6	17.202389	1.000000
2	ratatouille /award/award_winning_work/awards_w	38	17.536339	1.000000