

PLAGIARISM SCAN REPORT

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#Installing networkx Library

!pip install network

#Importing Required Libraries

import networkx #For handling Network graphs

import random

import csv

import numpy as np

import pandas as pd

from tensorflow.keras.layers import Dense,Input,Activation

from tensorflow.keras.models import Sequential

from sklearn.preprocessing import Normalizer

from tensorflow.keras.losses import categorical crossentropy

from sklearn.model selection import train test split

import matplotlib.pyplot as plt

from tensorflow.keras.optimizers import Adam

from keras.utils import np_utils

#stroing dataset path

path='/content/drive/MyDrive/SENA/facebookdata.txt'

#Loading dataset into dataset as graph in variable

with open(path,'rb') as file:

graph=networkx.read edgelist(file)

networkx.draw_spectral(graph,with_labels = True)

def Generate_False_Graph(Graph,neg_no):

Input: a graph as Networkx class

number of negative samples

Output :neg no number of false edges (edges which does not exist in the graph)

and build a new graph using negative edges

.....

Negative graph= networkx.Graph()

count=0

while (count [source,target]=random.sample(Graph.nodes(),2) # pick random two nodes in the graph

try:

short path=networkx.shortest path length(Graph,source,target)

if(short_path>=2):# there does not exist a direct edges between them

Negative_graph.add_edge(source,target, positive="False") # adding the false edges in the new graph created

count += 1 except:

```
pass
return Negative graph
def Extract Sample(Graph,pos no,neg no):
Input: a graph as Networkx class
Output: List of Positive and Negative samples
no edges=Graph.number of edges()
# Poistive samples
pos graph = networkx.Graph()
positves=random.sample(Graph.edges(),pos no)
pos graph.add edges from(positves, positive="True")
networkx.write edgelist(pos graph, "positive graph.txt", data=['positive'])
# Negative Samples
neg graph=Generate False Graph(Graph,neg no)
networkx.write edgelist(neg graph, "negatvie graph.txt", data=['positive'])
return pos graph.edges(),neg graph.edges()
def Extract feature(Graph,positives,negatives):
Input: Graph as network class
Positive dataset(Nodes having edges among them)
Negative Dataset(Nodes having shortest path greater than 2)
Output: Features of the two nodes in the given dataset
Features: Common Neighbours, Jacaard Coefficient, Resource allocation index, Adamic Adar Coefficient, Preferential
Attachment
data = []
feature name = [common neighbors,networkx.resource allocation_index,networkx.jaccard_coefficient,
networkx.adamic adar index, networkx.preferential attachment]
label = ["label"] + ["1" for i in range(len(positives))] + ["0" for i in range(len(negatives))]
for func in feature name:
preds = func(Graph, positives)
feature = [func.name] + [i[2] for i in preds]
preds = func(Graph, negatives)
feature = feature + [i[2] for i in preds]
data.append(feature)
data.append(label)
col names = []
records = []
for col in data:
col names.append(col[0])
records.append(col[1:])
records = np.array((records)).T
csvfile = pd.DataFrame(records, columns = col names).to csv('features.csv')
#Extracting 5000 samples for each dataset
x,y=Extract Sample(graph,5000,5000)
Extract feature(graph, x, y)
dataset=pd.read csv("features.csv")
x=dataset.iloc[:,:-1].values
y=dataset.iloc[:,-1].values
#Normalizing the dataset
t = Normalizer().fit(x)
x = t.transform(x)
def rand hexcolor():
Input:None
Output: Random color code in hexadecimal format
Process:pick random number and convert ot hexadecimal
```

rgb = ""

```
i = random.randrange(0, 2**8)
rgb += i.to bytes(1, "big").hex()
return '#'+rgb
k = [0, 1]
numbers=set(k.copy())
numbercol={i:rand hexcolor() for i in numbers } # giving each number a color
from sklearn.manifold import TSNE # tsne converts the image to dimensional feature vector which is used for plotting
# dimensionality reduction for viewing 400 dimesion hyperspace in 2d space
tsne = TSNE(metric='cosine') # we use cosine metric which finds similarty between two feature vectors(pixels of the
embed tsne = tsne.fit transform(np.array((x))) # converting 400 feature
plt.figure(figsize=(35, 35)) # assigning size for each scatter plot
for idx in range(len(x)):# iterating through all training data
plt.scatter(*embed_tsne[idx, :],c=numbercol[v[idx]]) # each point is marked in the plot with this feature considering as
coordinate with color chosen for that label corresponding to the record
X train, X test, y train, y test = train test split(x, y, test size=0.33, random state=42)
y train = np utils.to categorical(y train).astype('float32')
y_test = np_utils.to_categorical(y_test).astype('float32')
y train.shape
#Building the base model
model = Sequential()
model.add(Input(len(x[0])))
model.add(Dense(32, activation='relu'))
model.add(Dense(2, activation='sigmoid'))
#Compiling the model
model.compile(optimizer='adam', loss='categorical crossentropy', metrics=['accuracy'])
#Model training
model.fit(X train, y train, batch size=200, epochs=1000)
model.evaluate(X_test,y_test)
model.save('/content/drive/MyDrive/SENA/model.h5')
from tensorflow.keras.models import load model
model = load model('/content/drive/MyDrive/SENA/model.h5')
X = X \text{ test}[3000]
X = np.expand dims(X, axis=0)
np.argmax(model.predict(X))
```

for in "RGB": # creating color for each channel

y test[3000]

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