# LINEAR ALGEBRA ASSIGNMENT-5

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### PES1UG20CS150

### SEC C

#### IMPLEMENTATION OF PAGE RANK ALGORITHM IN PYTHON:

#### 1)GENERATING PAGE RANK PROBABILITIES IN THE FORM OF A MATRIX

```
def pagerank(G, alpha=0.85, personalization=None,
      max iter=100, tol=1.0e-6, nstart=None, weight='weight',
      dangling=None):
 if len(G) == 0:
   return {}
  if not G.is_directed():
    D = G.to directed()
  else:
   D = G
 # Create a copy in (right) stochastic form
 W = nx.stochastic graph(D, weight=weight)
 N = W.number_of_nodes()
  # Choose fixed starting vector if not given
 if nstart is None:
   x = dict.fromkeys(W, 1.0 / N)
   # Normalized nstart vector
    s = float(sum(nstart.values()))
    x = dict((k, v / s) for k, v in nstart.items())
 if personalization is None:
    # Assign uniform personalization vector if not given
    p = dict.fromkeys(W, 1.0 / N)
    missing = set(G) - set(personalization)
      raise NetworkXError('Personalization dictionary '
               'must have a value for every node. '
```

```
s = float(sum(personalization.values()))
   p = dict((k, v / s) for k, v in personalization.items())
if dangling is None:
  # Use personalization vector if dangling vector not specified
  dangling_weights = p
else:
  missing = set(G) - set(dangling)
  if missing:
    raise NetworkXError('Dangling node dictionary '
               'must have a value for every node. '
               'Missing nodes %s' % missing)
  s = float(sum(dangling.values()))
  dangling_weights = dict((k, v/s) for k, v in dangling.items())
dangling nodes = [n for n in W if W.out degree(n, weight=weight) == 0.0]
# power iteration: make up to max iter iterations
for _ in range(max_iter):
  xlast = x
  x = dict.fromkeys(xlast.keys(), 0)
  danglesum = alpha * sum(xlast[n] for n in dangling_nodes)
   for n in x:
    # this matrix multiply looks odd because it is
    # doing a left multiply x^T=xlast^T*W
    for nbr in W[n]:
       x[nbr] += alpha * xlast[n] * W[n][nbr][weight]
    x[n] += danglesum * dangling_weights[n] + (1.0 - alpha) * p[n]
  # check convergence, l1 norm
  err = sum([abs(x[n] - xlast[n]) for n in x])
  if err < N*tol:</pre>
      raise NetworkXError('pagerank: power iteration failed to converge in %d iterations.' % max iter)
[ ] import networkx as nx
    G=nx.barabasi_albert_graph(60,41)
    pr=nx.pagerank(G,0.4)
```

'Missing nodes %s' % missing)

pr

- {0: 0.027974270717457006, 1: 0.0127628368839034, 2: 0.013176723444143786, 3: 0.013177747527072622, 4: 0.013187466898160255, 5: 0.013161209959699736, 6: 0.0127628368839034, 7: 0.012562160948942153, 8: 0.013370175461666607, 9: 0.013367423527823493, 10: 0.012966440442363685, 11: 0.012572083959984303, 12: 0.012981177194839124, 13: 0.013357664225546649, 14: 0.013162784091391478, 15: 0.012794116114652159, 16: 0.012746551076996878, 17: 0.012966525434565043,
  - 18: 0.01299082083909732, 19: 0.013358460968529162,
  - 20: 0.012559926628535813, 21: 0.012385097979873872,
  - 22: 0.012981843309988613,
  - 23: 0.012985402118596601, 24: 0.012961209517747601,
  - 25: 0.012972463934502566, 26: 0.013162787965996983,
  - 27: 0.013166775312276354,
  - 28: 0.011946861285440464,
  - 29: 0.01297171164976596, 30: 0.012767283970356874,
  - 31: 0.012742473235933041,
  - 32: 0.012774581194681032,
  - 33: 0.013165728970937722, 34: 0.012559423699148328,
  - 34: 0.012559423699148328, 35: 0.012574366399099425,
  - 36: 0.013178994660872476,
  - 37: 0.013162447872856397.

```
38: 0.012546618649433343,
39: 0.0131727390026888,
```

40: 0.013565732355562396,

41: 0.013150392838513414,

42: 0.02758678002625911,

43: 0.02727992233732545,

44: 0.026720909580798917,

45: 0.026542949274437247,

46: 0.026140850136958396,

47: 0.02585214873990138,

48: 0.02543710445857681,

49: 0.02492353658663134,

50: 0.02471539936160183,

51: 0.024186424530845164,

52: 0.023684717633201,

53: 0.02352070217524788,

54: 0.02306055382593085,

55: 0.022830596519100375,

56: 0.02288679193854571,

57: 0.0222811536758942,

58: 0.02177521090987354,

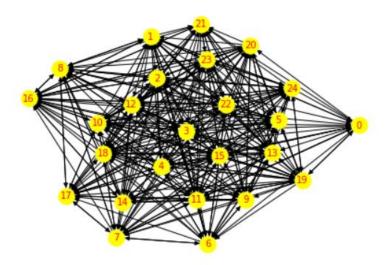
59: 0.021719909135324508}

## 2.IMPLEMENTATION BY USING A DIRECTED GRAPH TO VERIFY THE ALGORITHM USING NETWORKX LIBRARY AND THE WRITTEN FUNCTION

```
import networkx as nx
    import numpy as np
    import pandas as pd
    import matplotlib.pyplot as plt
    import operator
    import random as rd
    # created a directed graph
    graph=nx.gnp_random_graph(25,0.6,directed=True)
    #draw a graph
    nx.draw(graph,with labels=True,font color='red',font size=10,node color='yellow')
    #plot a graph
    plt.show()
    #number of nodes for graph
    count=graph.number of nodes()
    #graph neighbours of a node 1
    print(list(graph.neighbors(1)))
    #Page Rank Algorithm-Calculating random walk score
    rank dict={}
    x=rd.randint(0,25)
    for j in range(0,25):
      rank dict[j]=0
    rank_dict[x]=rank_dict[x]+1
    for i in range(600000):
      list n=list(graph.neighbors(x))
      if(len(list_n)==0):
        x=rd.randint(0,25)
        rank_dict[x]=rank_dict[x]+1
```

```
x=rd.choice(list_n)
    rank_dict[x]=rank_dict[x]+1
print("Random Walk Score Updated")
#normalising values
for j in range(0,25):
 rank dict[j]=rank dict[j]/600000
#Page rank by networkx library
pagerank=nx.pagerank(graph)
#sorting both dictionaries based on items
pagerank_sorted=sorted(pagerank.items(),key=lambda v:(v[1],v[0]),reverse=True)
pagerank_sorted
#sorting the rank_dict based on values
rank dict sorted=sorted(rank dict.items(),key=lambda v:(v[1],v[0]),reverse=True)
rank_dict_sorted
print("The order generated by our implementation algorithm is\n")
for i in rank_dict_sorted:
 print(i[0],end=" ")
print("\n\nThe order generated by networkx library is\n")
for i in pagerank_sorted:
 print(i[0],end=" ")
```





[0, 2, 3, 5, 6, 8, 11, 12, 13, 16, 18, 21, 22, 23, 24]
Random Walk Score Updated
The order generated by our implementation algorithm is

18 16 11 23 5 1 12 8 10 24 20 19 7 15 9 21 17 3 14 13 2 22 6 4 0
The order generated by networkx library is

18 16 11 23 1 12 5 10 8 24 7 20 19 15 9 17 21 3 14 13 2 22 6 4 0