Networks Exercise 2 In 11.18 Version 3.2

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1 复杂网络: 小世界模型

```
#!/usr/bin/env python
# -*- coding: utf-8 -*-
import numpy as np # Import modules
import networkx as nx
import matplotlib.pyplot as plt

G = nx.watts_strogatz_graph(100, 10, 0.1) # Make graph
plt.figure(3, figsize=(12, 12), dpi=720)
nx.draw_circular(G, with_labels=False, node_size=30)
plt.savefig("G.png")
plt.show()

l_average = nx.average_shortest_path_length(G) # Cal parame
c_average = nx.average_clustering(G)

print "average_shortest_path_length = %s" % l_average # Proprint "average_clustering = %s" % c_average
```

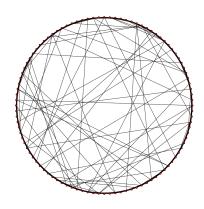


Figure 1: 小世界模型

2 画图: 平均最短路径和群聚系数 随概率的变化趋势

```
#!/usr/bin/env python
# -*- coding: utf-8 -*-
import networkx as nx
import numpy as np
import matplotlib.pyplot as plt
l_average_list = []
c_average_list = []
p = np.linspace(0.01, 0.1, 20)
for i in range(len(p)): # 平均化处理
     l_avg_list = []
     c_avg_list = []
     for j in range(20):
          1_average = nx.average_shortest_path_length(
          nx.watts_strogatz_graph(1000, 10, p[i]))
c_average = nx.average_clustering(
    nx.watts_strogatz_graph(1000, 10, p[i]))
          1_avg_list.append(1_average)
          c_avg_list.append(c_average)
    l_average_list.append(float(sum(l_avg_list) / len(l_avg_list)))
c_average_list.append(float(sum(c_avg_list) / len(c_avg_list)))
plt.title(r"$p-\bar 1$")
plt.xlabel(r"$p$")
plt.ylabel(r"$\bar 1$")
plt.legend()
plt.plot(p, l_average_list, label=r"$\bar 1$",
           color="blue", linewidth=0, marker='o')
plt.legend()
plt.savefig("l_average.png")
plt.show()
plt.title(r"$p-\bar c$")
plt.xlabel(r"$p$")
plt.ylabel(r"$\bar c$")
plt.plot(p, c_average_list, label=r"$\bar 1$",
color="red", linewidth=0, marker='o')
plt.legend()
plt.savefig("c_average.png")
plt.show()
print "p = %s" % p
print "average_shortest_path_length = %s " % 1_average_list
print "average_clustering = %s" % c_average_list
```

3 求出介数并作散点图

```
#!/usr/bin/env python
# -*- coding: utf-8 -*-
# Version 1.2
import itertools
import numpy as np
```

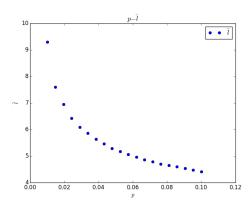


Figure 2: 平均最短路径随概率的变化

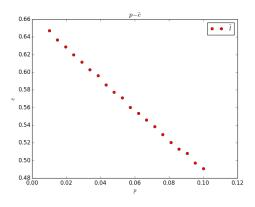


Figure 3: 群聚系数随概率的变化

```
import networkx as nx
import matplotlib.pyplot as plt

G = nx.random_graphs.erdos_renyi_graph(100, 0.5)
degree = nx.degree_histogram(G)
k = range(len(degree))
sigma_m1 = nx.betweenness_centrality(G)
print "betweenness_centrality = %s" % sigma_m1
plt.title(r"$k- \sigma (m)$")
plt.xlabel(r"$\sigma (m)$")
plt.ylabel(r"$\sigma (m)$")
plt.ylabel(r"$\sigma (m)$")
plt.plot(*zip(*sorted(sigma_m1.items())), color="red", line)
plt.savefig("betweenness_centrality.png")
plt.show()
```

4 画图:对平均最短路径和群聚系数作归一化处理

```
#!/usr/bin/env python
# -*- coding: utf-8 -*-
```

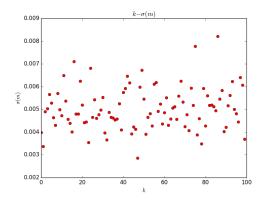


Figure 4: 介数随节点的散点图

```
import networkx as nx
import numpy as np
import matplotlib.pyplot as plt
l_average_list = []
c_average_list = []
l_average_max = nx.average_shortest_path_length(
    nx.watts_strogatz_graph(1000, 10, 0))
c_average_max = nx.average_clustering(nx.watts_strogatz_graph(1000, 10, 0))
p = [1, 0.00016.
     0.00028900000000000003,
     0.0004913000000000001,
     0.0008352100000000001,
     0.001419857,
     0.0024137569.
     0.004103386729999999
     0.006975757440999999,
     0.011858787649699998,
     0.020159939004489997,
     0.034271896307633,
     0.0582622237229761,
     0.09904578032905938,
     0.16837782655940095,
     0.2862423051509816,
     0.48661191875666865,
     0.8272402618863367,
     1]
for i in range(len(p)):
    l_avg_list = []
    c_avg_list = []
    for j in range(20): # 平均化处理
         1_average = nx.average_shortest_path_length(
             nx.watts_strogatz_graph(1000, 10, p[i])) / l_average_max
         c_average = nx average_clustering(
             nx.watts_strogatz_graph(1000, 10, p[i])) / c_average_max
         1_avg_list.append(l_average)
    c_avg_list.append(c_average)
l_average_list.append(float(sum(l_avg_list) / len(l_avg_list)))
c_average_list.append(float(sum(c_avg_list) / len(c_avg_list)))
print l_average_list
print c_average_list
plt.xlabel(r"$p$")
{\tt plt.semilogx(p, l_average\_list, label=r"$L(p) / L(0)$",}
```

```
color="blue", linewidth=0, marker='o') # 对横坐标作对数处理 plt.semilogx(p, c_average_list, label=r"$C(p) / C(0)$", color="red", linewidth=0, marker='o') plt.legend() # make legend plt.savefig("lc.png") plt.show()
```

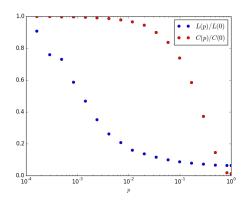


Figure 5: 平均最短路径和群聚系数随概率变化的对比图

5 小结:

通过这次作业感受到了处理数据中的几种方式: 1. 平均化处理 2. 归一化处理 3. 对数坐标; 以及在程序设计中的几种思维方式: 1. 先数据结构的观点去看待问题 2. 从循环或者函数的角度去看待问题——循环单值循环和定义函数——可以复用 3. 算法设计及优化 ... 以及小世界模型在小概率时的一些特征.