

Problem_8_1

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```
[33]: import pandas as pd
import networkx as nx
import matplotlib as mpl
import matplotlib.pyplot as plt
import numpy as np
from collections import Counter
import random
```

1 Assignment 8

1.1 Problem 8-1 European Power Grid

```
[3]: !wget https://zenodo.org/record/47317/files/gridkit_euorpe.zip --no-clobber

--2022-01-16 19:17:44--
https://zenodo.org/record/47317/files/gridkit_euorpe.zip
Resolving zenodo.org (zenodo.org)... 137.138.76.77
Connecting to zenodo.org (zenodo.org)|137.138.76.77|:443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 1805506 (1.7M) [application/octet-stream]
Saving to: 'gridkit_euorpe.zip'

gridkit_euorpe.zip 100%[=====>] 1.72M 2.48MB/s in 0.7s

2022-01-16 19:17:46 (2.48 MB/s) - 'gridkit_euorpe.zip' saved [1805506/1805506]
```

```
[6]: !unzip *.zip
```

```
Archive:  gridkit_euorpe.zip
  inflating: gridkit_europe-highvoltage-links.csv
```

```

    inflating: gridkit_europe-highvoltage-vertices.csv
gridkit_europe.zip          gridkit_europe-highvoltage-vertices.csv
'gridkit_europe.zip?download=1' sample_data
gridkit_europe-highvoltage-links.csv

```

```
[7]: !ls
```

```

gridkit_europe.zip          gridkit_europe-highvoltage-vertices.csv
gridkit_europe-highvoltage-links.csv sample_data

```

```
[11]: data = pd.read_csv('./gridkit_europe-highvoltage-links.csv', sep=',')

display(data)
```

```

      l_id  ...                               wkt_srid_4326
0    22139  ...  SRID=4326;LINESTRING(7.61602141237887 47.23259...
1    65908  ...  SRID=4326;LINESTRING(26.6449735029238 50.29628...
2    67370  ...  SRID=4326;LINESTRING(18.1844471738857 40.01608...
3     3868  ...  SRID=4326;LINESTRING(10.2724738664462 49.88060...
4     4982  ...  SRID=4326;LINESTRING(3.80033661099728 45.04510...
...  ...  ...
18799 33852  ...  SRID=4326;LINESTRING(34.891839086949 48.400953...
18800 12859  ...  SRID=4326;LINESTRING(13.0725880043994 46.33610...
18801 17974  ...  SRID=4326;LINESTRING(22.9977678819448 63.71031...
18802 69707  ...  SRID=4326;LINESTRING(5.34304391214955 43.44674...
18803 28512  ...  SRID=4326;LINESTRING(10.3699302949897 45.62925...

```

[18804 rows x 17 columns]

create network

```
[12]: G = nx.from_pandas_edgelist(data, source='v_id_1',target='v_id_2')
print(nx.info(G))
# convert to undirected graph
G = G.to_undirected()
```

Graph with 13844 nodes and 17277 edges

```
[13]: G = nx.Graph(G) # multigraph to graph
print(nx.info(G))
```

Graph with 13844 nodes and 17277 edges

1. Molloy-Reed criterion: Calculate κ

```
[17]: # average degree <k>
degrees = [val for (node, val) in G.degree()]
avg_k = sum(degrees)/len(degrees)
print(f"<k> = {avg_k}")
```

$\langle k \rangle = 2.4959549263218723$

```
[23]: # second moment of degree <k^2>
avg_k2 = sum([k**2 for k in degrees])/len(degrees)
print(f"<k^2> = {avg_k2}")
```

$\langle k^2 \rangle = 7.888182606183184$

```
[22]: # Molloy-Reed criterion
kappa = avg_k2/avg_k
print(f"Molloy-Reed: kappa = {kappa}")
```

Molloy-Reed: kappa = 3.1603866411992825

$\kappa > 2$, i.e. the Molloy-Reed criterion is satisfied and a giant component is expected in the network.

2. Absolute and relative size of the largest component:

```
[26]: largest_cc = max(nx.connected_components(G), key=len)
print(f"absolute size of largest component {len(largest_cc)}")
print(f"relative size of largest component {len(largest_cc)/G.
↪number_of_nodes()}")
```

absolute size of largest component 13478

relative size of largest component 0.9735625541750939

Yes, the largest component is almost of the same size as the network and can therefore be regarded as the giant component.

3. Compute f_c and f_c^{ER} for the network's degree distribution.

```
[28]: fc = 1-1/(kappa-1)
fc_ER = 1-1/avg_k
print(f"fc = {fc} < fc_ER = {fc_ER}")
```

$fc = 0.5371198928332217 < fc_{ER} = 0.5993517393065926$

The critical threshold of the network is slightly smaller than what would be expected by an Erdős-Rényi model with the same $\langle k \rangle$, the network is therefore slightly less robust against random failure.

4.

(i) maximum component size and the (ii) average component size as a function of $f \in [0,1]$.

```
[57]: def get_component_sizes(G, n_samples=10, f_samples=21):
    fractions = np.linspace(0,1,f_samples)
    G = G.copy()
    N = G.number_of_nodes()
    Gc_0 = len(max(nx.connected_components(G), key=len))
    Gc_max = np.zeros((len(fractions),n_samples))
    Gc_avg = np.zeros((len(fractions),n_samples))
    for i in range(n_samples):
```

```

for j, f in enumerate(fractions):
    G_ = G.copy()
    nodes = [n for n in G_.nodes]
    # remove f*N nodes
    if f > 0.0:
        rnd_nodes = random.sample(nodes, round(f*N))
        for n in rnd_nodes:
            G_.remove_node(n)
    cc = [len(c) for c in sorted(nx.connected_components(G_), key=len,
↪reverse=True)]
    largest_cc = max(cc) if len(cc)>0 else 0
    Gc_max[j,i] = largest_cc
    Gc_avg[j,i] = sum(cc)/len(cc) if len(cc)>0 else 0
Gc_max = np.average(Gc_max,axis=1)
Gc_avg = np.average(Gc_avg,axis=1)
return Gc_max, Gc_avg, fractions

```

```
[58]: Gc_max, Gc_avg, fractions = get_component_sizes(G, n_samples=10)
```

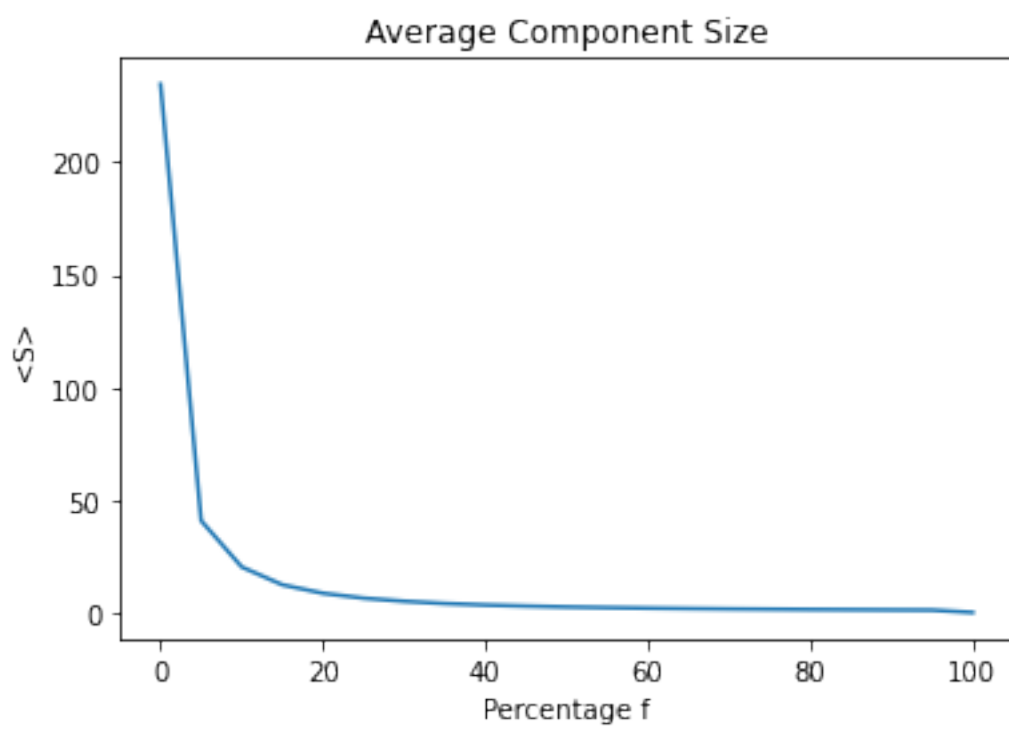
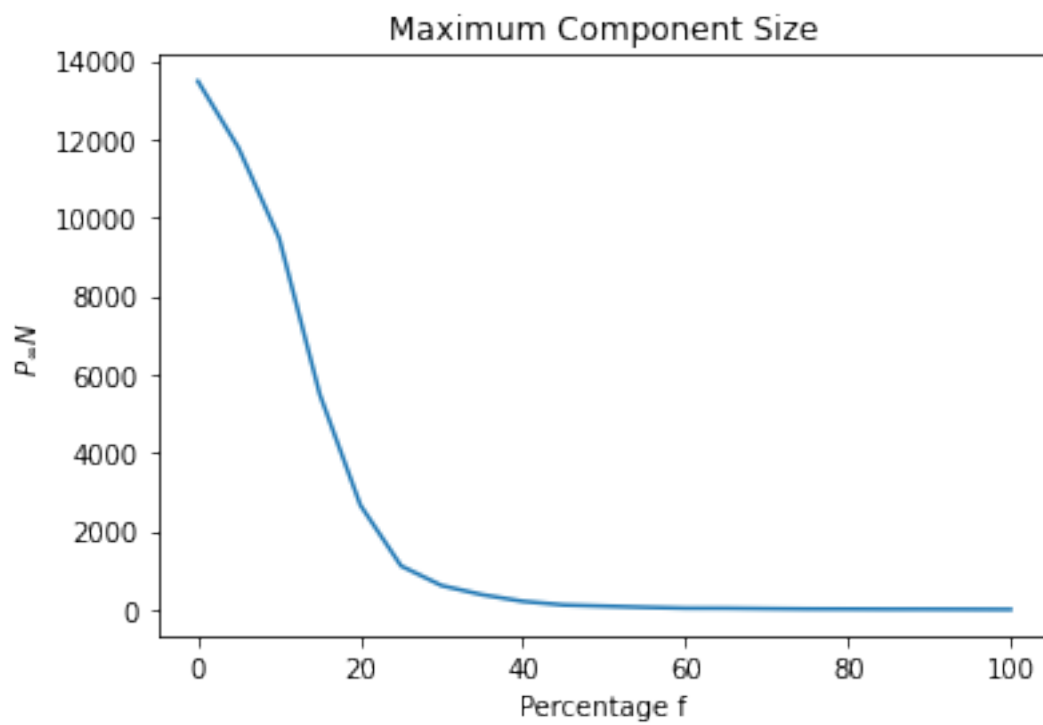
```

[60]: plt.figure()
plt.plot(fractions*100, Gc_max)
plt.ylabel("$P_{\infty}N$")
plt.xlabel("Percentage f")
plt.title("Maximum Component Size")

plt.figure()
plt.plot(fractions*100, Gc_avg)
plt.ylabel("<S>")
plt.xlabel("Percentage f")
plt.title("Average Component Size")

```

```
[60]: Text(0.5, 1.0, 'Average Component Size')
```



Both maximum and average component size are rapidly decreasing, even before a fraction f_c or f_{c_ER} is removed, indicating that the network is very prone to random attacks. (Note: When calculating the average component size we have also included the largest (giant) component)

[]: