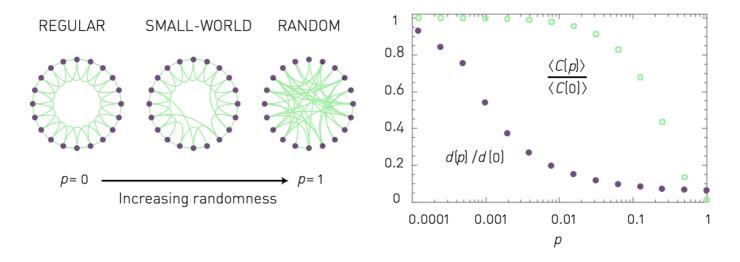
Small-world and scale-free models, graphs vs networks

You are given six networks in Pajek format.

- Zachary karate club network (<u>karate_club.net</u>)
- Map of Darknet from Tor network (<u>darknet.net</u>)
- iMDB actors collaboration network (collaboration imdb.net)
- WikiLeaks cable reference network (<u>wikileaks.net</u>)
- Enron e-mail communication network (enron.net)
- A small part of Google web graph (www_google.net)

I. Watts-Strogatz small-world graphs

1. **(discuss)** Study the algorithm for generating Watts-Strogatz small-world graphs G(n, k, p) introduced in lectures. Does the algorithm generate networks with realistic structure? What is the time complexity of the algorithm?



1. **(code)** Implement the algorithm and generate Watts-Strogatz small-world graphs that best match the networks above. (Set k to $\langle k \rangle$ rounded to the nearest even number and try to find the value of p that best reproduces $\langle C \rangle$, $p \approx 1 - \sqrt[3]{\frac{4C(k-1)}{3(k-2)}}$.) Compute their average node clustering coefficient $\langle C \rangle$ and approximate average distance between the nodes $\approx \langle d \rangle$. Are the results expected?

II. Barabási-Albert and Price scale-free graphs

1. **(discuss)** Study the following two algorithms for generating Barabási-Albert scale-free graphs G(n,c)

and Price scale-free graphs G(n,c,a) using the relation $\frac{q+a}{n(c+a)} = \frac{c}{c+a} \frac{q}{nc} + \frac{a}{c+a} \frac{1}{n}$. What is the main difference between the algorithms? What is the time complexity of the algorithms?

```
input nodes n, out-degree c, free a
                                                        output directed scale-free G
input nodes n, degree c
                                                            1: Q \leftarrow \text{empty queue}
output undirected scale-free G
                                                            2: G \leftarrow c isolated nodes
   1: Q \leftarrow \text{empty queue}
   2: G \leftarrow \text{empty graph}
                                                            3: while not G has n nodes do
   3: while not G has n nodes do
                                                                    i \leftarrow \text{add node to } G
                                                                    for c times do
                                                            5:
           i \leftarrow \text{add node to } G
                                                                        if [0,1).random() < c/(c+a) then
   5:
           for c times do
                                                            7:
                                                                             Q.add(j \leftarrow Q.random())
                Q.add(i)
                Q.add(j \leftarrow Q.random())
                                                                             Q.add(j \leftarrow \{0, \ldots, i\}.random())
                                                            9:
               add link between i and j
                                                           10:
                                                                        add link from i to i
   9: return G
                                                           11: return G
```

2. **(code)** Implement both algorithms and generate Barabási-Albert and Price scale-free graphs corresponding to larger networks above. Plot their degree distribution p_k and compute power-law exponents γ of seemingly scale-free distributions using the maximum likelihood formula below. Are the results expected?

$$\gamma = 1 + \bar{n} \left[\sum_{i=1}^{n} \ln \frac{k_i}{k_{min} - \frac{1}{2}} \delta(k_i \ge k_{min}) \right]^{-1}$$

III. Synthetic random graphs vs real networks

Consider different large-scale properties of real networks. Namely, low average node degree $\langle k \rangle \ll n$, one giant connected component $S \approx 1$, short distances between the nodes $\langle d \rangle \approx \frac{\ln n}{\ln \langle k \rangle}$, high average node clustering coefficient $\langle C \rangle \gg 0$, power-law degree distribution $p_k \sim k^{-\gamma}$, pronounced community structure etc.

- 1. **(discuss)** Design synthetic graph model that generates undirected graphs that are *most different* from real networks.
- 2. **(code)** Implement generative graph model that *well reproduces* the structure of real undirected networks.
- 3. **(discuss)** Does your model have reasonable interpretation or explanation? Does it also reproduce the structure of real directed networks?