1. (20 pts) Consider the following simple and rather unrealistic model of a network: each of \( n \) vertices belongs to one of \( g \) groups. The \( m \)th group has \( n_m \) vertices and each vertex in that group is connected to others in the group with independent probability \( p_m = A(n_m - 1)^{-\beta} \), where \( A \) and \( \beta \) are constants, but not to any vertices in other groups. Thus, this network takes the form of a set of disjoint groups of communities.

(a) Calculate the expected degree \( \langle k \rangle \) of a vertex in group \( m \).

(b) Calculate the expected value \( \langle C_m \rangle \) of the local clustering coefficient for vertices in group \( m \).

(c) Hence show that \( \langle C_m \rangle \propto \langle k \rangle^{-\beta/(1-\beta)} \). What value would \( \beta \) have to assume for the expected value of the local clustering coefficient to fall off as \( \langle k \rangle^{-0.75} \), as has been conjectured by some researchers?

2. (20 pts) Consider the random graph \( G(n, p) \) with average degree \( c \).

(a) Show that in the limit of large \( n \) the expected number of triangles in the network is \( \frac{1}{6} c^3 \). In other words, show that the number of triangles is constant, neither growing nor vanishing in the limit of large \( n \).

(b) Show that the expected number of connected triples in the network, as in Eq. (7.41) in Networks, is \( \frac{1}{2} n c^2 \).

(c) Hence, calculate the clustering coefficient \( C \), as defined in Eq. (7.41) in Networks, and confirm that it agrees for large \( n \) with the value given in Eq. (12.11) in Networks.

3. (15 pts) Consider an undirected, unweighted network of \( n \) vertices that contains exactly two subnetworks of size \( n_A \) and \( n_B \), which are connected by a single edge \((A, B)\), as sketched here:

![Diagram](image)

Show that the closeness centralities \( C_A \) and \( C_B \) of vertices \( A \) and \( B \), as defined by Eq. (7.29) in Networks, are related by

\[
\frac{1}{C_A} + \frac{n_A}{n} = \frac{1}{C_B} + \frac{n_B}{n}.
\]
4. (15 pts) Consider an undirected (connected) tree of \( n \) vertices. Suppose that a particular vertex in the tree has degree \( k \), so that its removal would divide the tree into \( k \) disjoint regions, and suppose that the sizes of those regions are \( n_1, \ldots, n_k \). Show that the betweenness centrality \( b \) of the vertex is
\[
b = n^2 - \sum_{i=1}^{k} n_i^2.
\]

5. (30 pts) The Medici family was a powerful political dynasty and banking family in 15th century Florence. The classic network-based explanation of their power, offered by Padgett and Ansell in 1993, claims that they established themselves as the most central players within the network of prominent Florentine families (shown in Fig. 1 below).

Visit the Index of Complex Networks (ICON) at icon.colorado.edu and obtain a copy of the Medici network data file, under the “Padgett Florentine families” ICON entry.

Conduct the following two tests of their structural importance hypothesis:

(a) Construct a table with four columns, each containing a list of (family, score) pairs for the following centrality score functions:
  - degree centrality,
  - harmonic centrality (Eq. 7.30 in Networks),
  - eigenvector centrality (Eq. 7.6 in Networks),
  - betweenness centrality\(^1\) (Eq. 7.38 in Networks).

Within each centrality, sort the pairs in decreasing order of importance; three decimal places is sufficient detail.

Discuss (i) the degree to which these scores agree with Padgett and Ansell’s claim that the Medicis occupied a structurally important position in this network, and (ii) what the scores say about the second most important family.

(b) Determine whether the relative structural importance of the Medicis can be explained solely by the degree sequence \( \{k_i\} \) of the network \( G \). To do this, use the configuration model to produce an ensemble of 1000 random graphs with the same degree sequence \( \{k_i\} \) as \( G \), and extract a distribution of harmonic centrality scores for each node.

Make a figure showing the difference between a node’s harmonic centrality on \( G \) and its average harmonic centrality in the ensemble. On this figure, include lines showing the 25% and 75% quantiles of the distribution around the mean for all vertices (as in the figure below, for the Karate club).

Finally, (i) discuss your results here in terms of Padgett and Ansell’s story of the Medici family, and (ii) comment on whether your results from part (5a) have changed.

6. (25 pts extra credit) Consider the task of designing a network in which distinct vertices hold the status of highest centrality for some set \( S \) of centrality measures. (Ties are prohibited.) If the network is relatively small compared to the number of centrality measures, this can be a very difficult task.

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\(^1\)To properly compute betweenness, you must handle multiple geodesics with equal length between a pair \( i \) and \( j \). See Lectures 2 nodes, footnote 16 for one way to do this.
• As a warm up, design by hand a small network ($n \leq 7$) that has this property with respect to $\mathcal{S} = \{\text{degree, betweenness}\}$ centralities. Include a visualization of the network (label all the vertices, and indicate which ones win at which centrality measure) and produce a table in which the columns show the ranked vertices and scores for these centrality measures.

• Now, write a program to search the space of all possible connected, non-isomorphic networks of a given size $n$ to find a network with this property for $\mathcal{S} = \{\text{degree, harmonic, betweenness, eigenvector}\}$ centralities, or report that no such network exists for that size. Run this program for $n = 3, 4, 5, \ldots$ and report the smallest value of $n$ for which at least one such network exists. Visualize that network, label all the vertices, and indicate which ones win at which centrality measure; and produce a table in which the columns show the ranked vertices and scores for these centrality measures. Comment briefly about what insights you gained about how to build networks that have distinct winners for different measures of centrality.

Figure 1: (left) The Medici family alliance network, from Padgett and Ansell (1993). (right) The difference between the observed centrality and the mean centrality under the configuration model, for the karate club network.