## Lab3: Significance of Network Metrics

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## 4.1 Analytical estimation of the p-value

## 4.1.1 Erdös-Rényi Graph

Given our original graph with N vertices and E edges, the Erdös-Rényi graph has the same size and order but with its edges randomized. Of the two possible Erdös-Renyi models, we will use G(N,p) where p will be such that the expected number of edges is E (which is  $p = \frac{E}{\binom{N}{2}}$ ). This has the advantage that  $X_j$  the random variables that for every possible vertex indicate wether it exists  $(X_j=1)$  or not  $(X_j=0)$  are independent, which will be very useful later on. When working with large values of N and E (such as those we study in this lab session), the models G(N,E) and G(N,p) should give very similar random graphs. For  $X_j$  where j is the index of any vertex, we calculate its expectation and variance:

$$E[X_j] = 0 \cdot (1 - p) + 1 \cdot p = p$$
  
 $Var(X_j) = E[X^2] - (E[X])^2 = p - p^2$ 

Given a vertex v of the graph, it can have N-1 neighbours (each with probability p), which results in  $\binom{N-1}{2}$  possible pairs of neighbours. The expected number of pairs. This gives an expected number of pairs of neighbours  $n_v = p^2 \binom{N-1}{2}$ . Then, we can estimate the local clustering of v by

$$C_v pprox rac{\sum_{j=1}^{\lfloor n_v \rfloor} X_j}{n}.$$

Since  $X_j$  are independent and equally distributed, we can apply the central limit theorem, which states that for a large enough sample size, the distribution of  $C_v$  is the normal  $N(\mathrm{E}(C_V), \frac{\mathrm{Var}(X_j)}{n}) \approx N(C_V^S, \frac{p-p^2}{p\binom{n-1}{2}}) = N(C_v^S, \frac{1-p}{p\binom{N-1}{2}}).$ 

We can apply the central limit theorem again to  $C_{WS} = \frac{1}{N} \sum_{i=1}^{N} C_i$ , wich gives us the distribution of  $C_{WS}$ : the normal distribution

$$N(C_v^S, \frac{1-p}{pN\binom{N-1}{2}}) = N(C_v^S, \frac{1-p}{E(N-2)}).$$

Note that in the last step we used the equality  $pN\binom{N-1}{2} = \frac{EN\binom{N-1}{2}}{\binom{N}{2}} = E(N-2)$ .