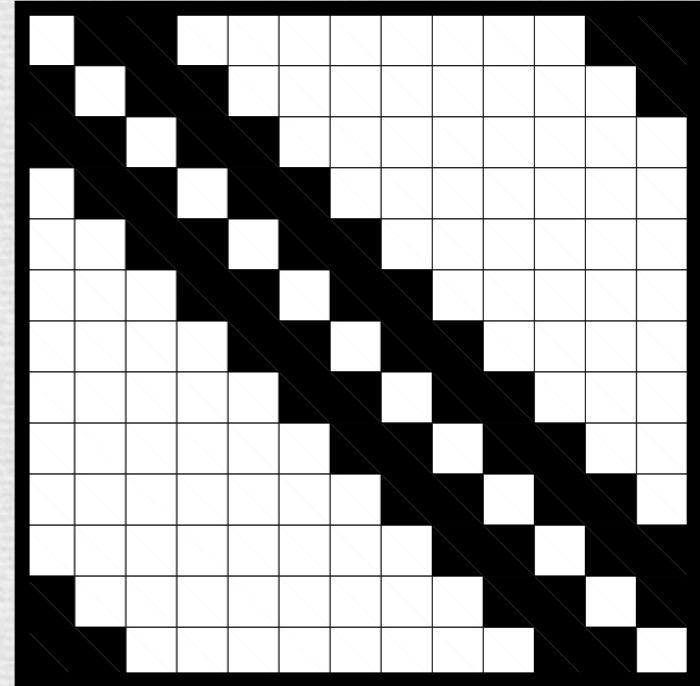
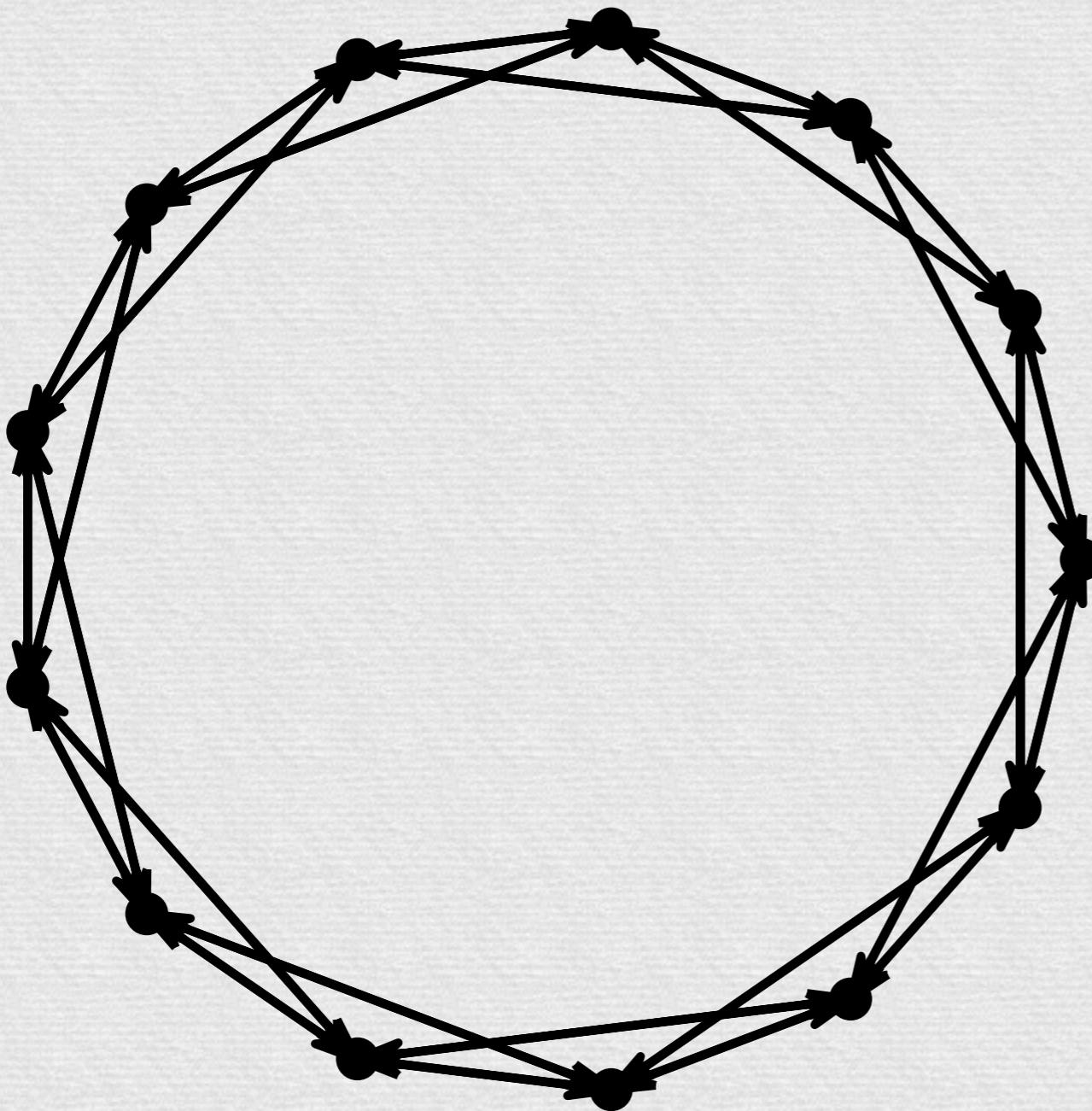


michelle rudolph-lilith  
\*lyle e muller

**an algebraic approach to  
small-world network models**

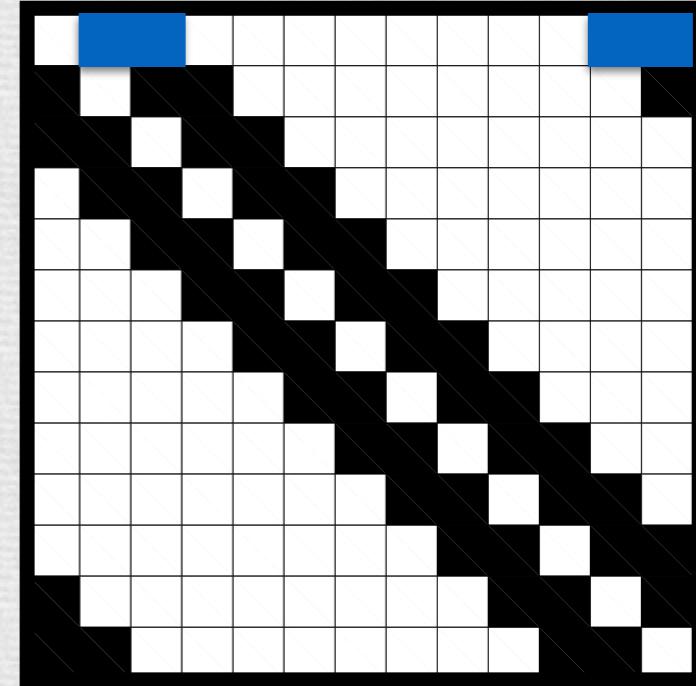
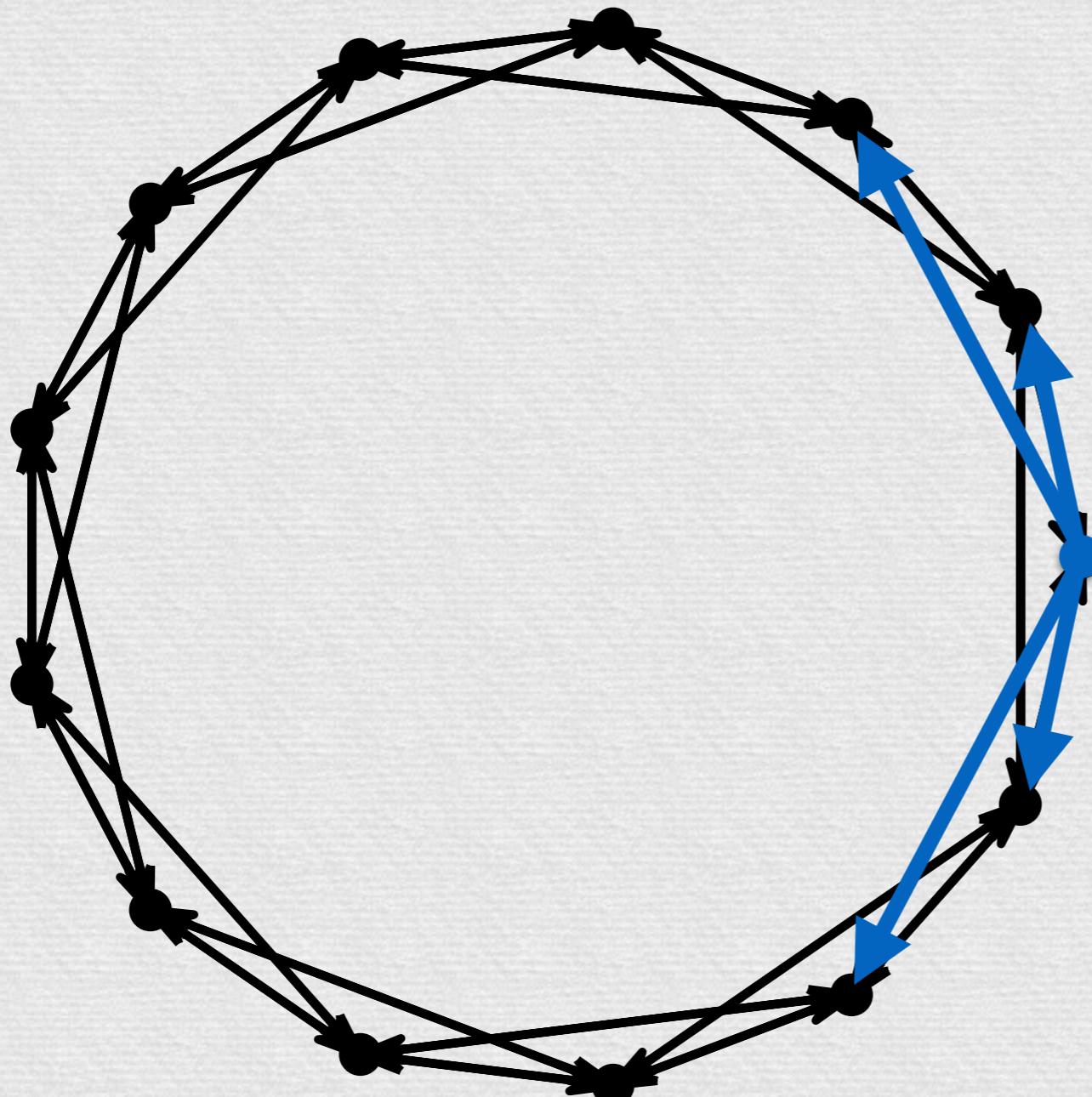
**brain networks: small-worlds, after all?**

# Ring Graph



$G_{RG}$

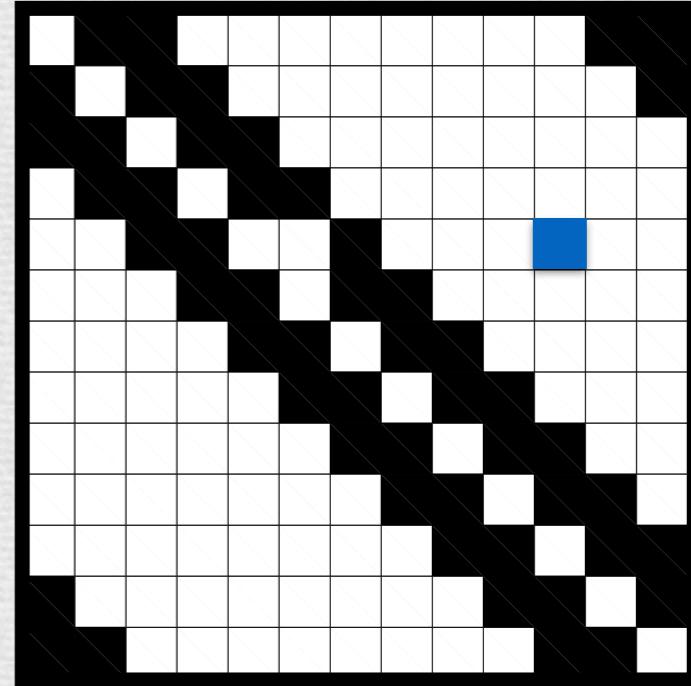
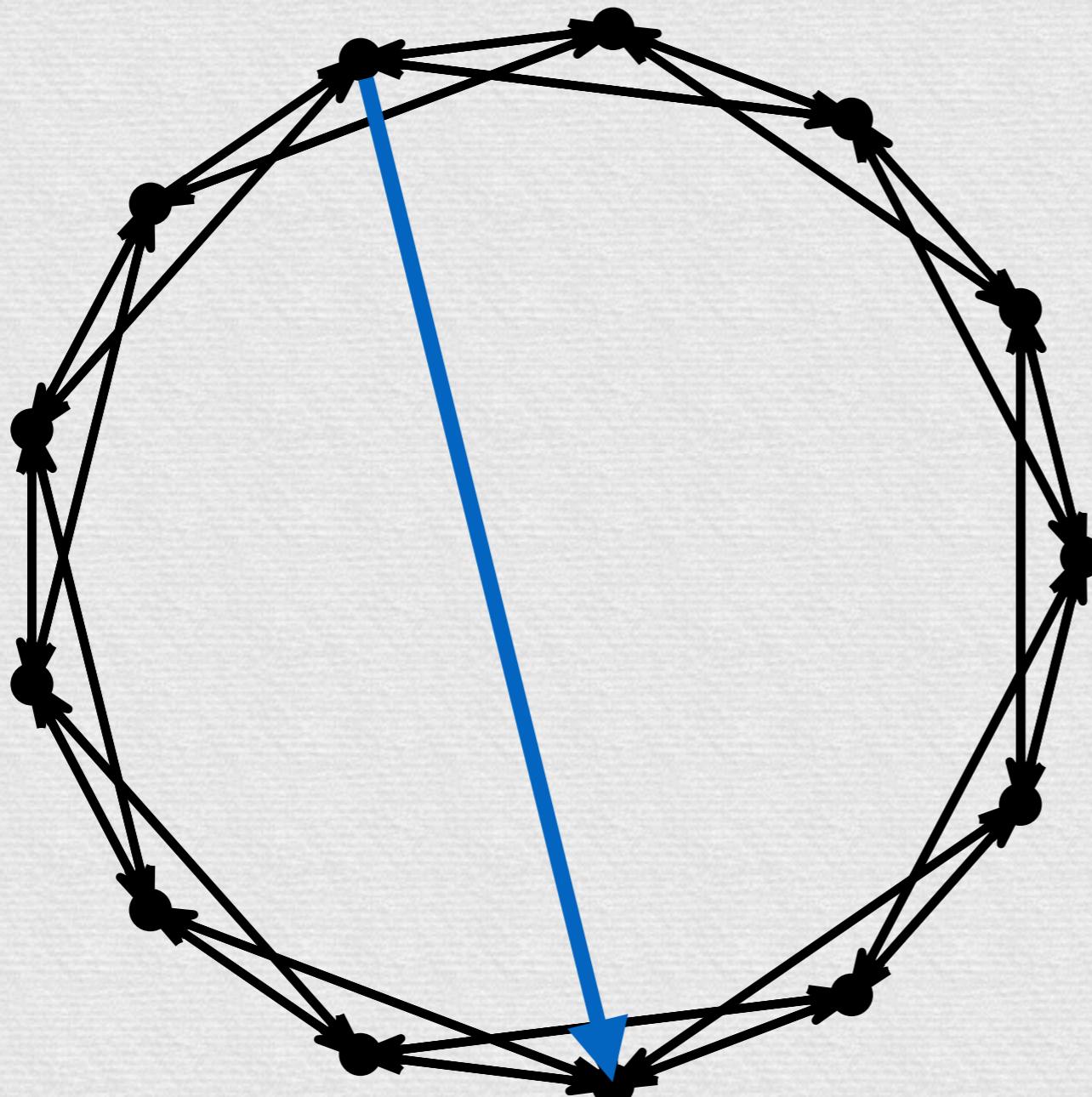
# Ring Graph



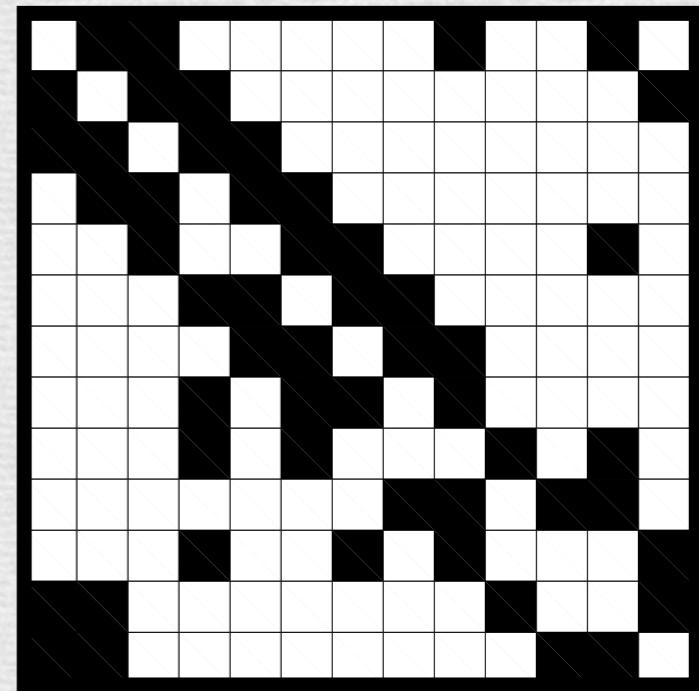
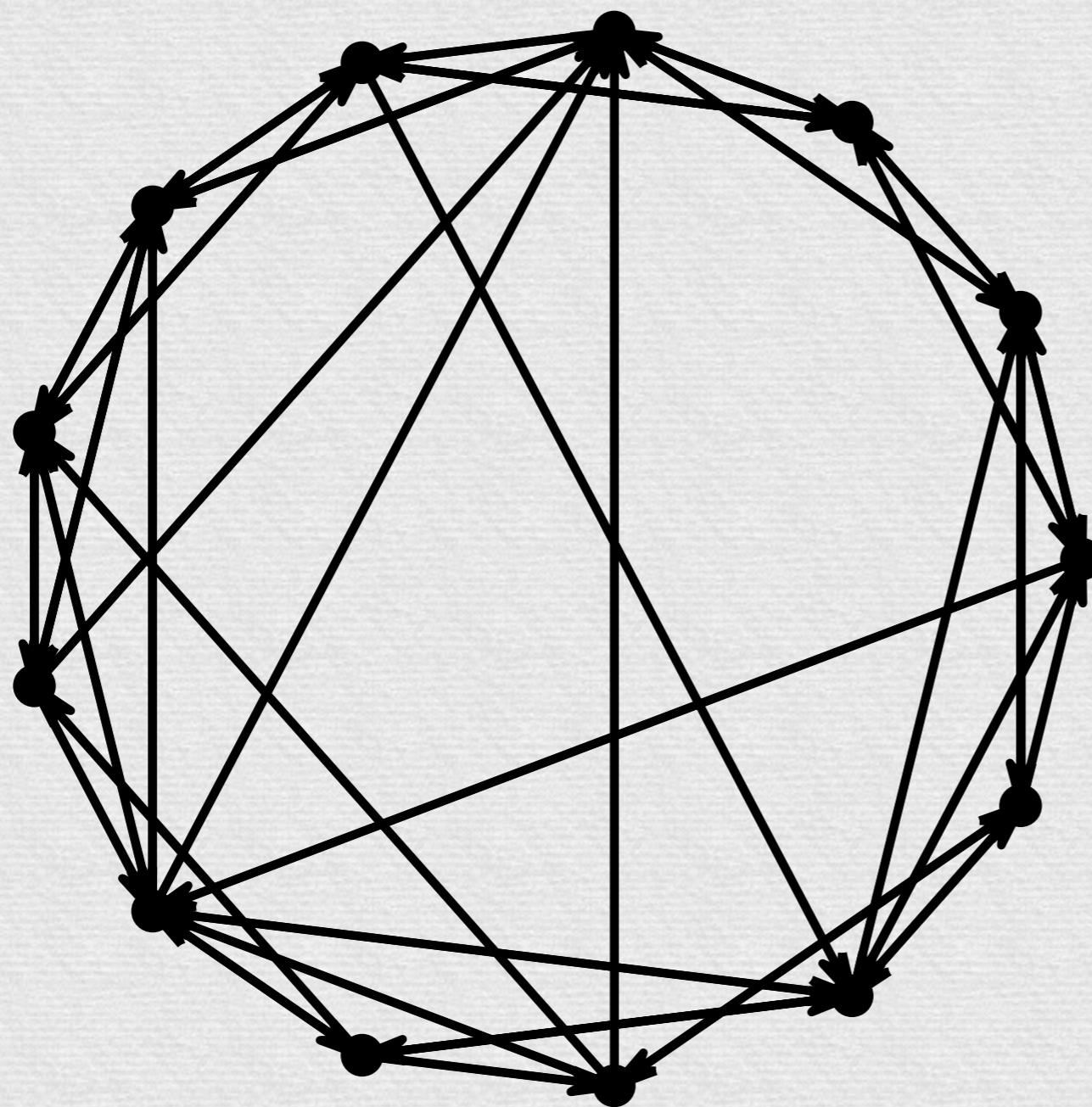
$k$

$$N_E^{RG} = 2kN_N$$

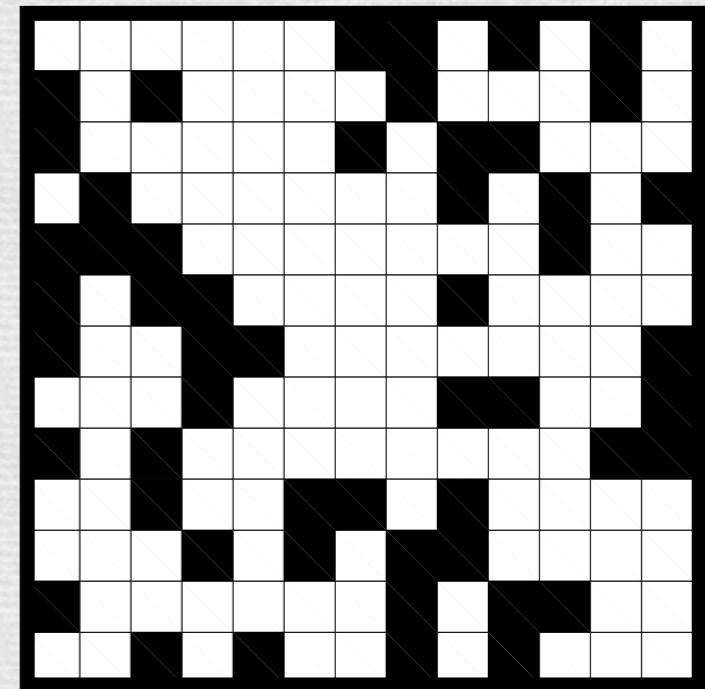
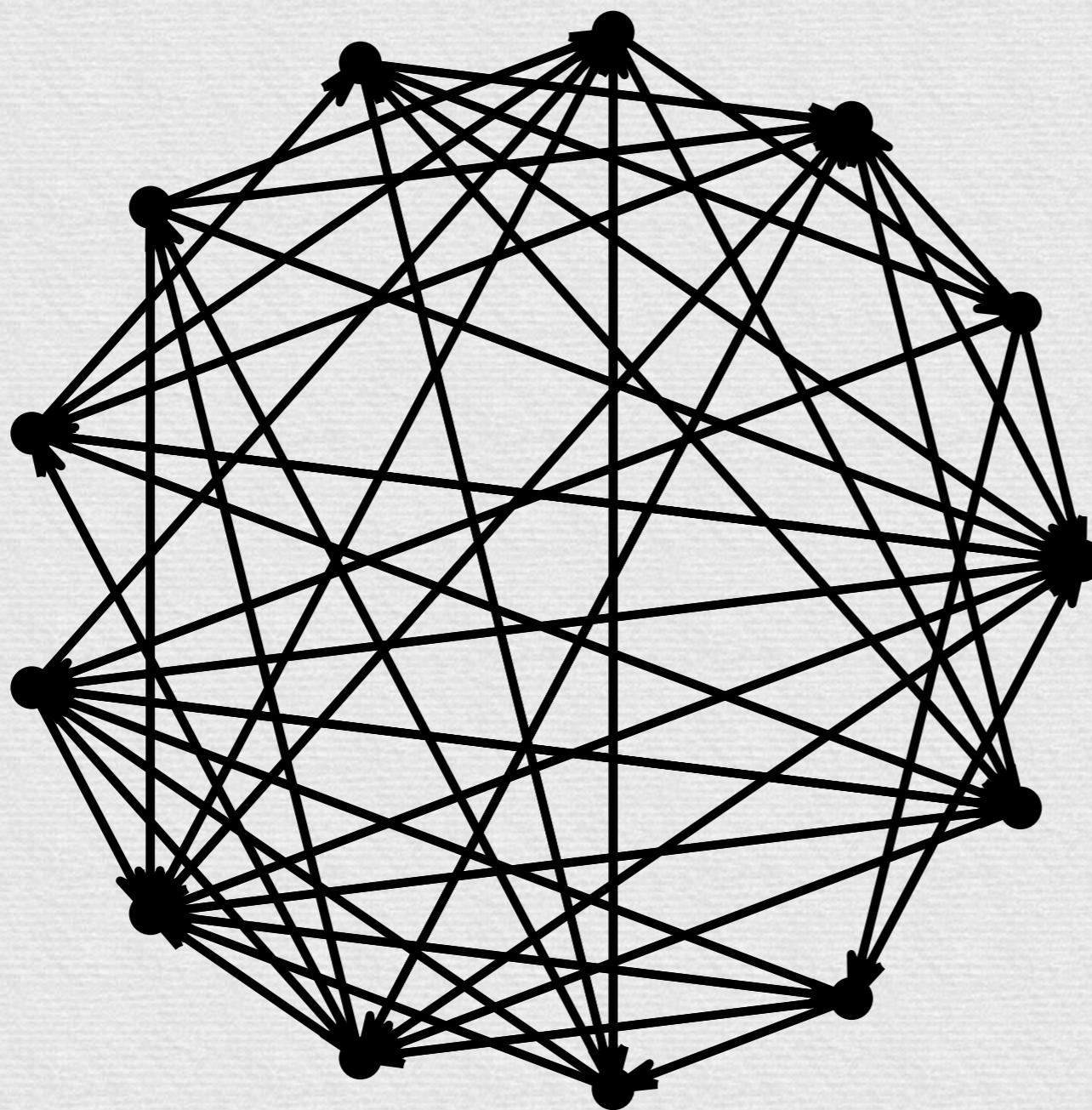
# Rewiring

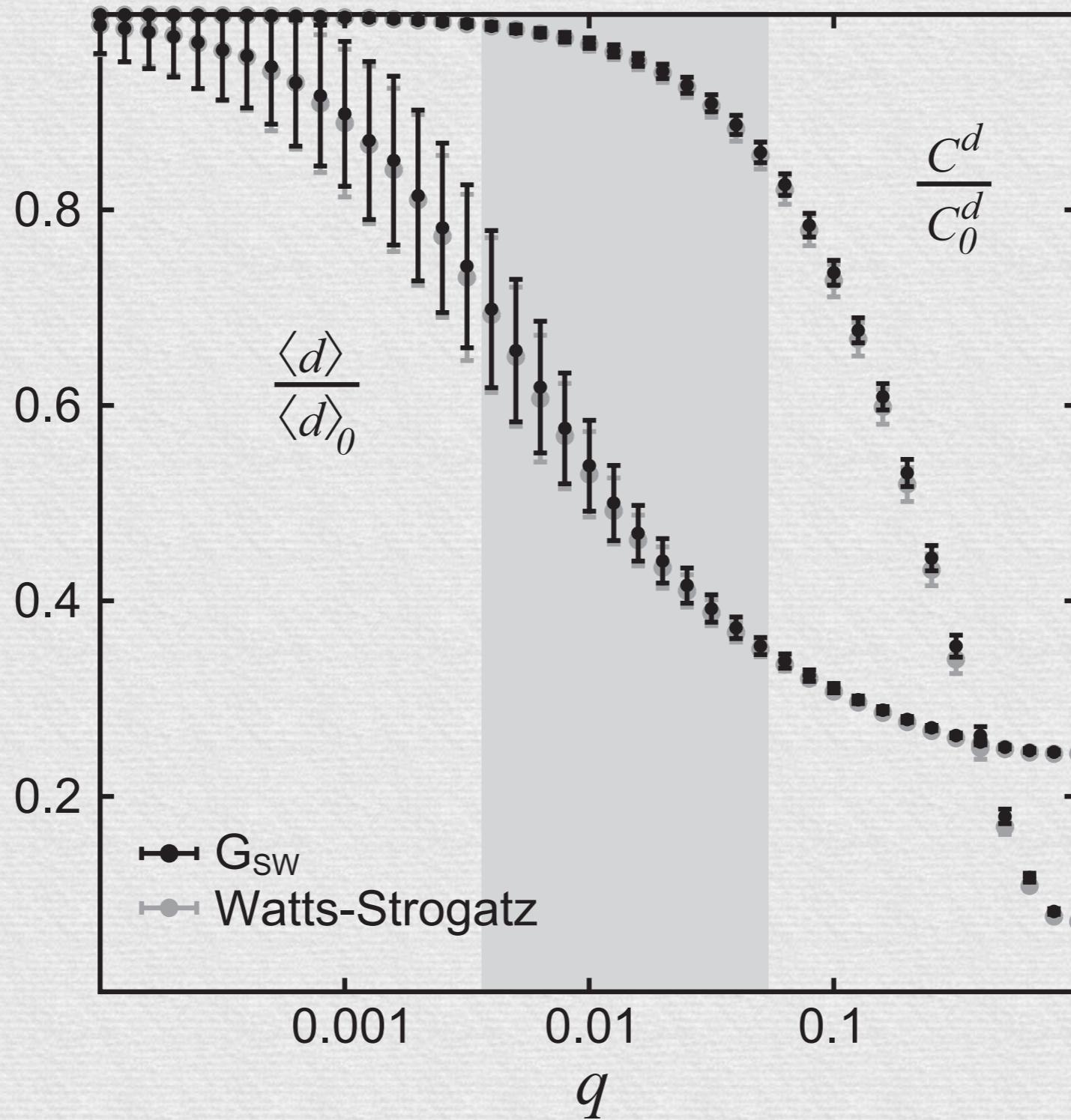


# Rewiring



# Random Graph





- Consider random, uniform rewiring on a ring graph.
- Note that this is equivalent to starting with a “reduced” ring graph with

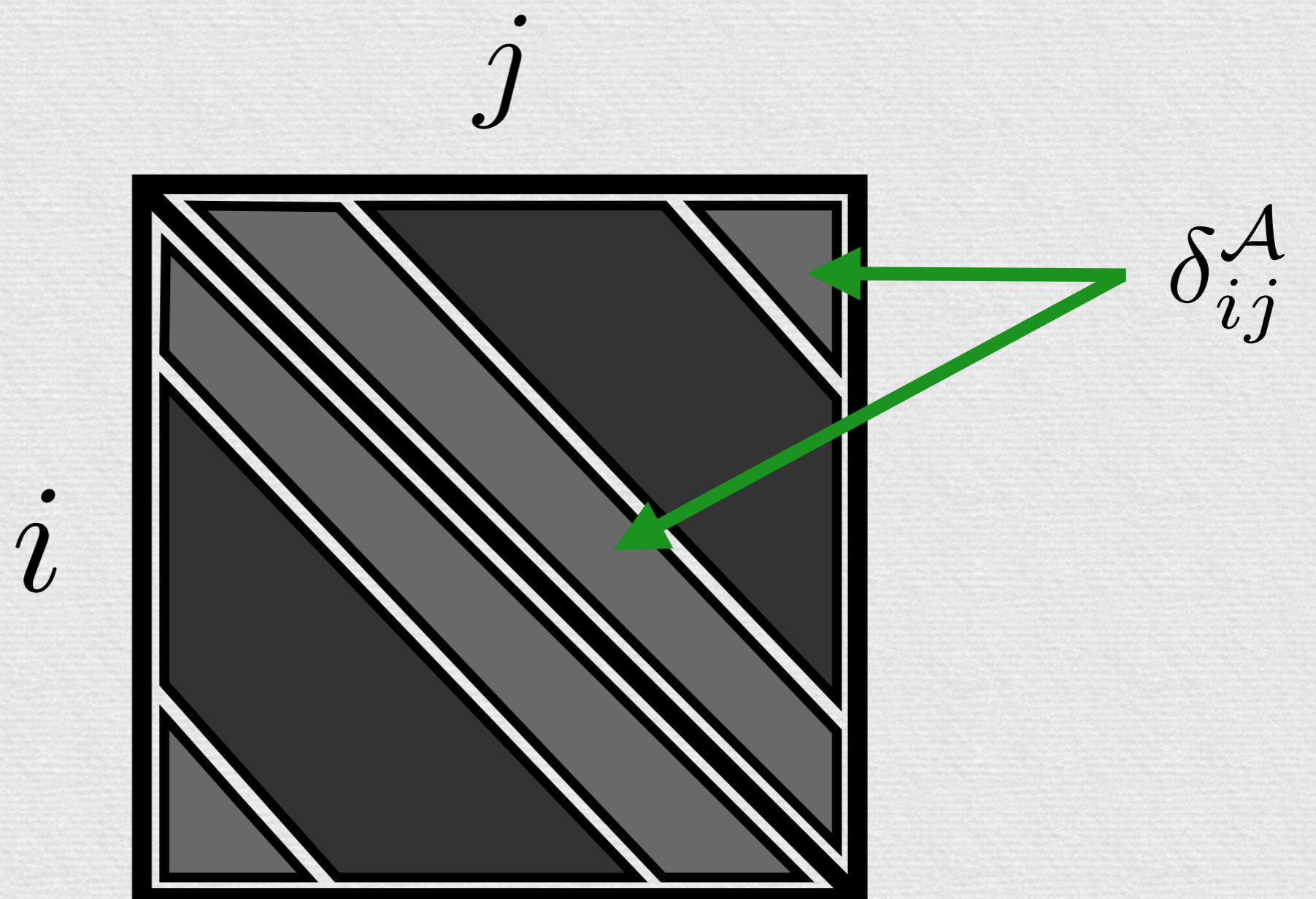
$$2kN_N - q2kN_N(N_N - 2k - 1)/(N_N - 1) \text{ edges}$$

uniform randomly distributed across the  $2kN_N$  edges of  $\mathfrak{G}_{RG}$ , and distributing the remaining

$$q2kN_N(N_N - 2k - 1)/(N_N - 1) \text{ edges}$$

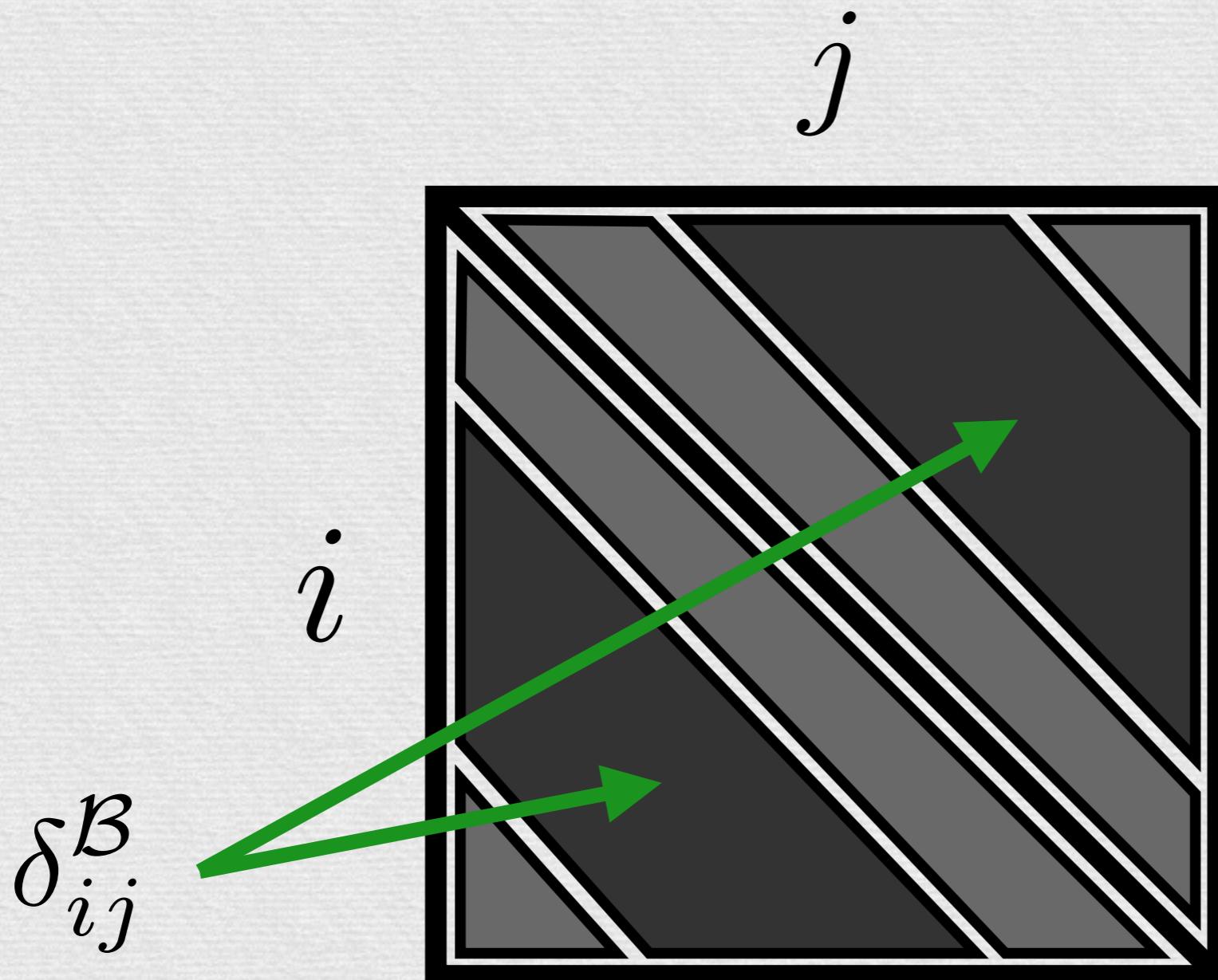
outside  $\mathfrak{G}_{RG}$ .

$$\delta_{ij}^{\mathcal{A}}$$



$$\begin{aligned}\mathcal{A} = & \left( \{i, j\}, i, j \in [1, N_N] : ((i - k \leq j \leq i + k) \wedge (i \neq j)) \right. \\ & \left. \vee (i + N_N - k \leq j) \vee (j + N_N - k \leq i) \right)\end{aligned}$$

$$\delta_{ij}^{\mathcal{B}}$$



$$\begin{aligned}\mathcal{B} = & \left( \{i, j\}, i, j \in [1, N_N] : (i + k < j < i + N_N - k) \right. \\ & \left. \vee (j + k < i < j + N_N - k) \right)\end{aligned}$$

# Random Annihilation Operator

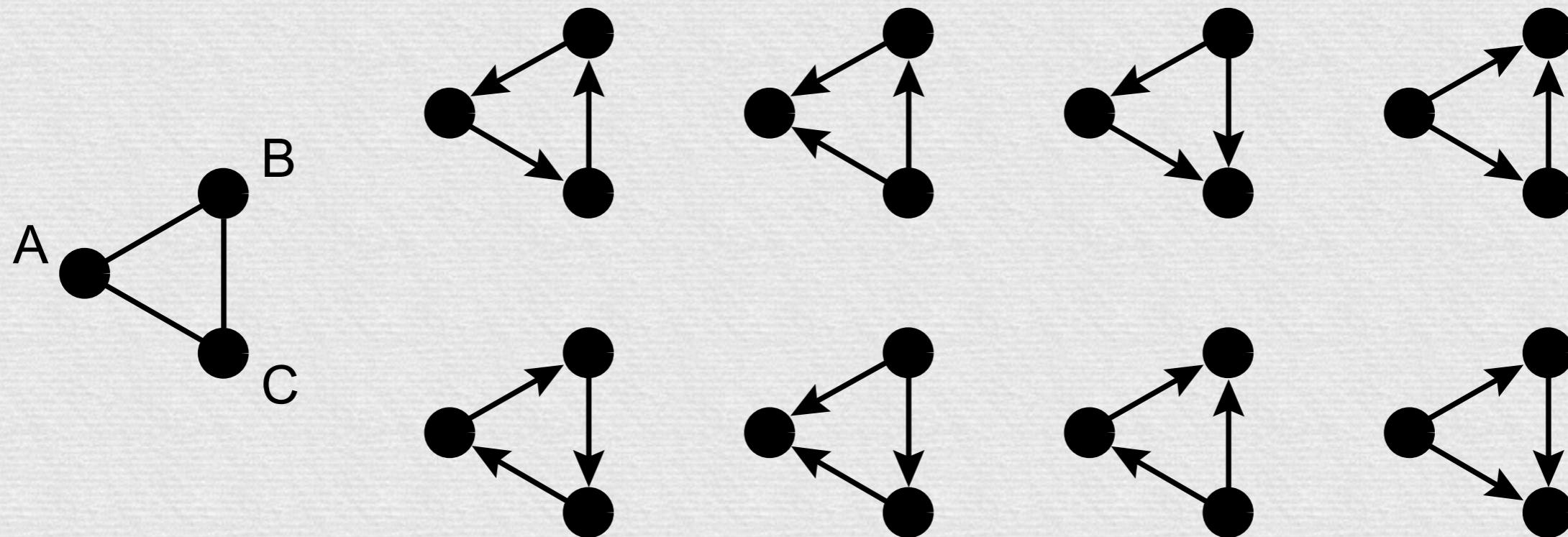
$$\hat{r}^p(x) = \begin{cases} x & \text{with probability } p \\ 0 & \text{with probability } (1 - p) \end{cases}$$

$$\sum_{x=1}^n \hat{r}^p(x) = np$$

# Algebraic Adjacency Matrix

$$a_{ij}^{SW} = (1 - \hat{r}_{ij}^{p_{\mathcal{A}}})\delta_{ij}^{\mathcal{A}} + \hat{r}_{ij}^{p_{\mathcal{B}}}\delta_{ij}^{\mathcal{B}}$$

# Clustering Coefficient

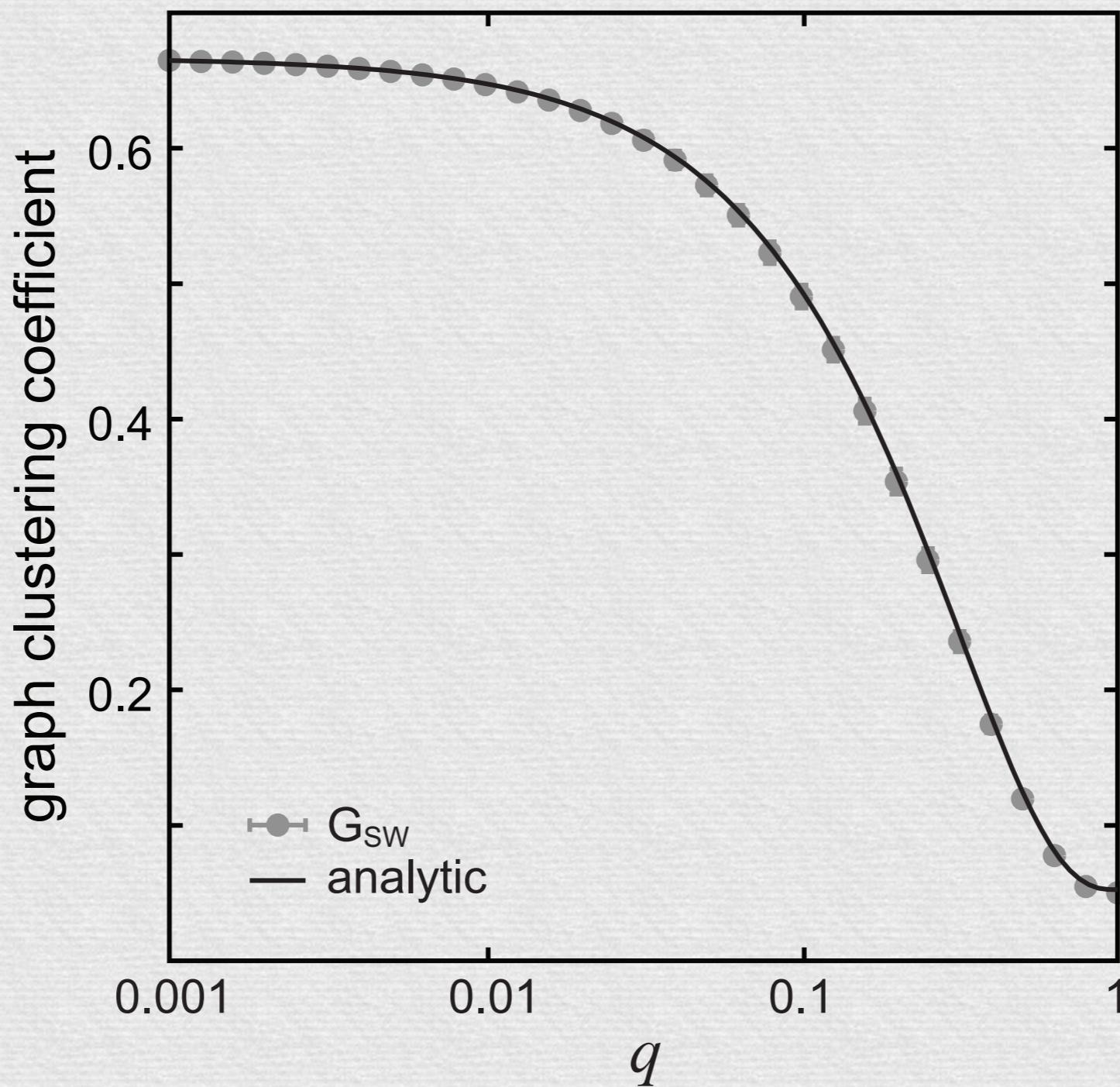


$$C^d = \frac{1}{N} \sum_{i,j,h=1}^{N_N} (a_{ij} + a_{ji})(a_{ih} + a_{hi})(a_{jh} + a_{hj})$$

# Clustering Coefficient

$$\begin{aligned} C^{d SW} = & \frac{64}{\mathcal{N}^{SW}(N_N - 1)^3} \\ & \times \left\{ \left( N_N - 1 - q(N_N - 2k - 1) \right)^3 \left( k^3 + \sum_{m=1}^{N_N-1} a_m^3(N_N, k) \right) \right. \\ & + 3kq \left( N_N - 1 - q(N_N - 2k - 1) \right)^2 \left( k^2(N_N - 2k - 1) - \sum_{m=1}^{N_N-1} a_m^2(N_N, k)b_m(N_N, k) \right) \\ & + 3k^2q^2 \left( N_N - 1 - q(N_N - 2k - 1) \right) \left( k(N_N - 2k - 1)^2 + \sum_{m=1}^{N_N-1} a_m(N_N, k)b_m^2(N_N, k) \right) \\ & \left. + q^3k^3 \left( (N_N - 2k - 1)^3 - \sum_{m=1}^{N_N-1} b_m^3(N_N, k) \right) \right\} \end{aligned}$$

# Clustering Coefficient

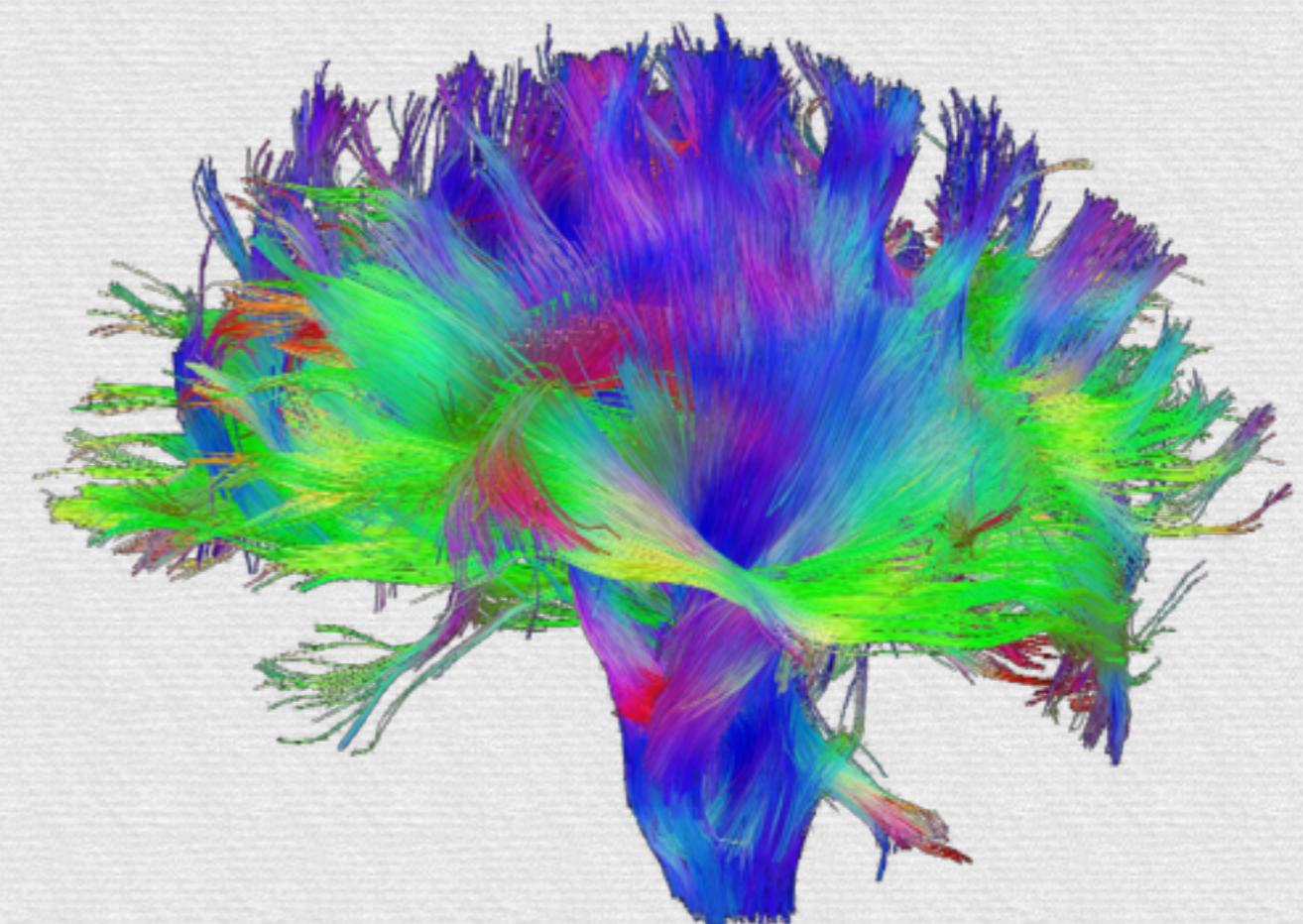
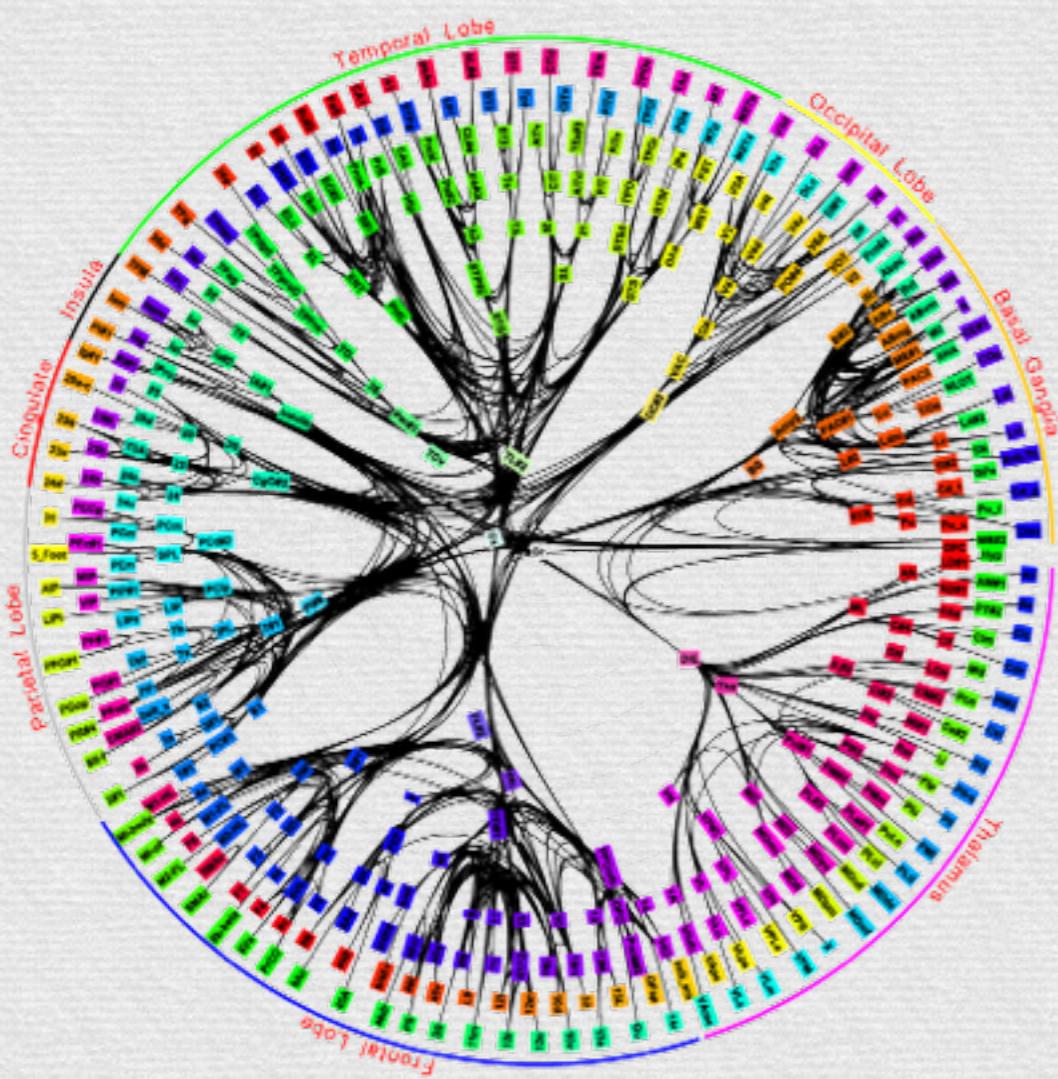


# Clustering Coefficient: Scaling Limit

$$C^{d SW} \Big|_{N_N \rightarrow \infty} = \frac{3(k-1)}{2(2k-1) + q(2-q)} (q-1)^3$$

[ cf. Barrat & Weigt, *European Physics Journal B*. **13** (2000): 547-560 ]

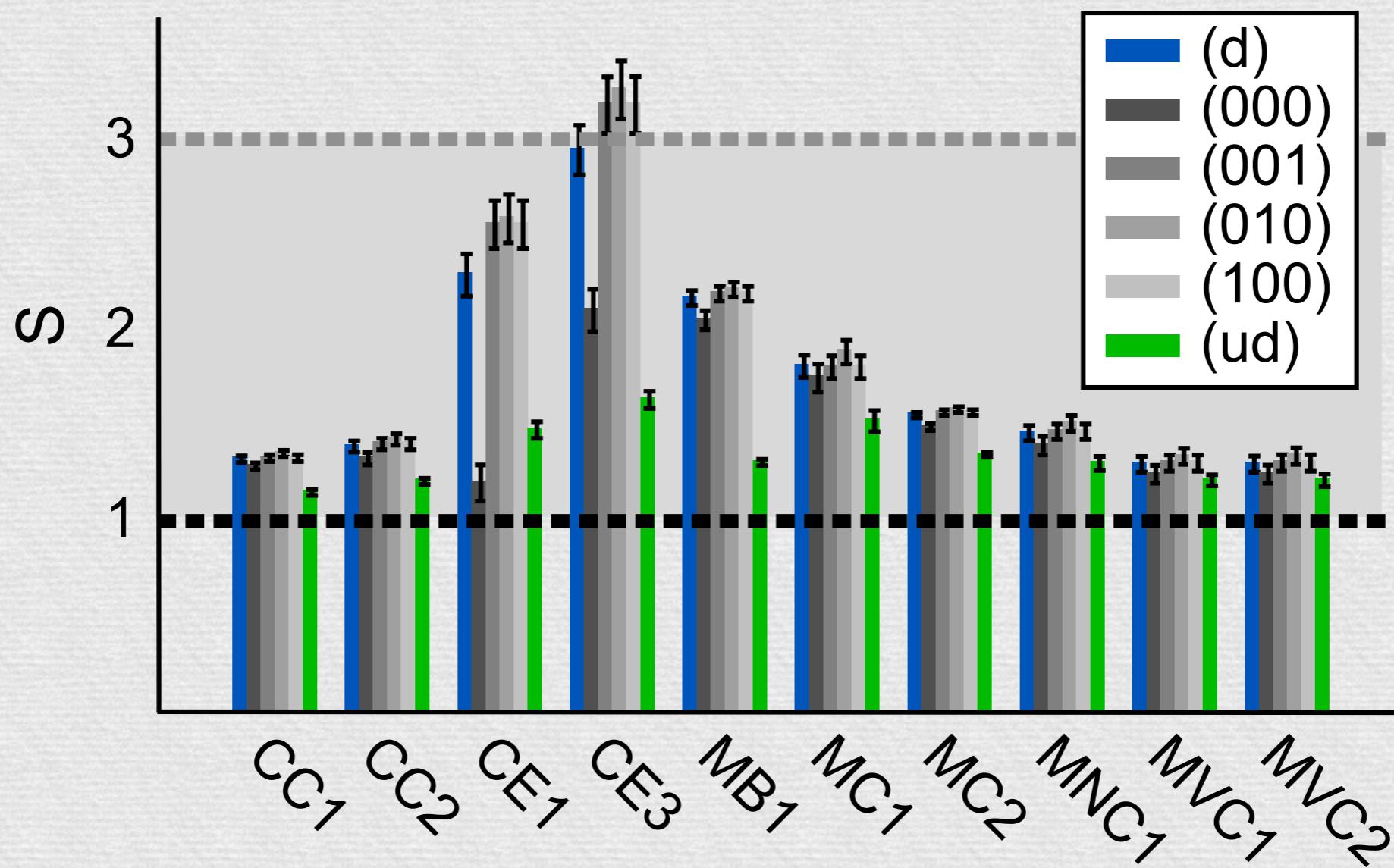
# Application to Neural Graphs



[ Muller et al., *New Journal of Physics*. (2014): basically submitted ]

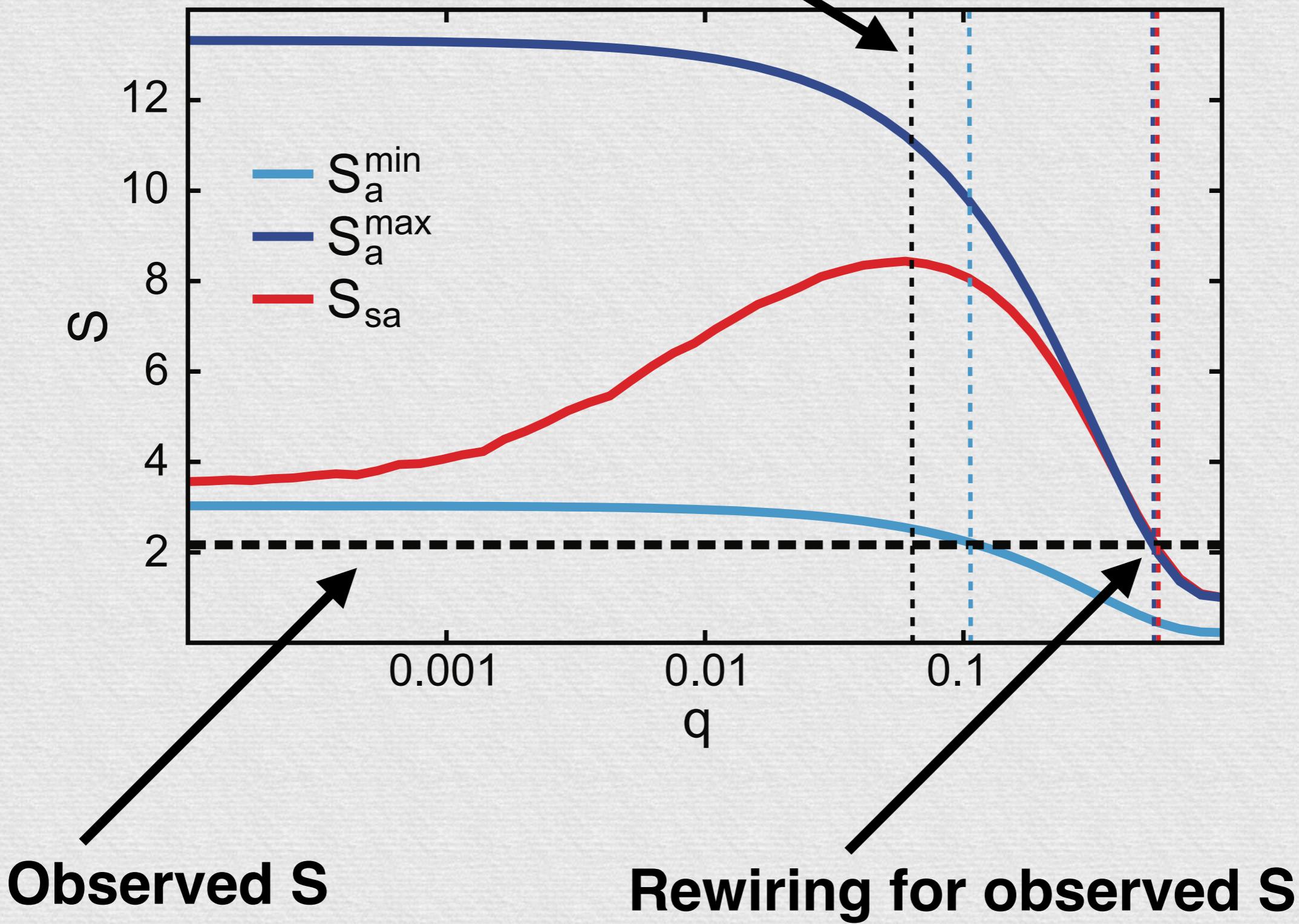
# Small-Worldness Index

$$S^d = \frac{C_{SW}^d}{C_{ER}^d} \frac{\langle d \rangle_{ER}}{\langle d \rangle_{SW}}$$

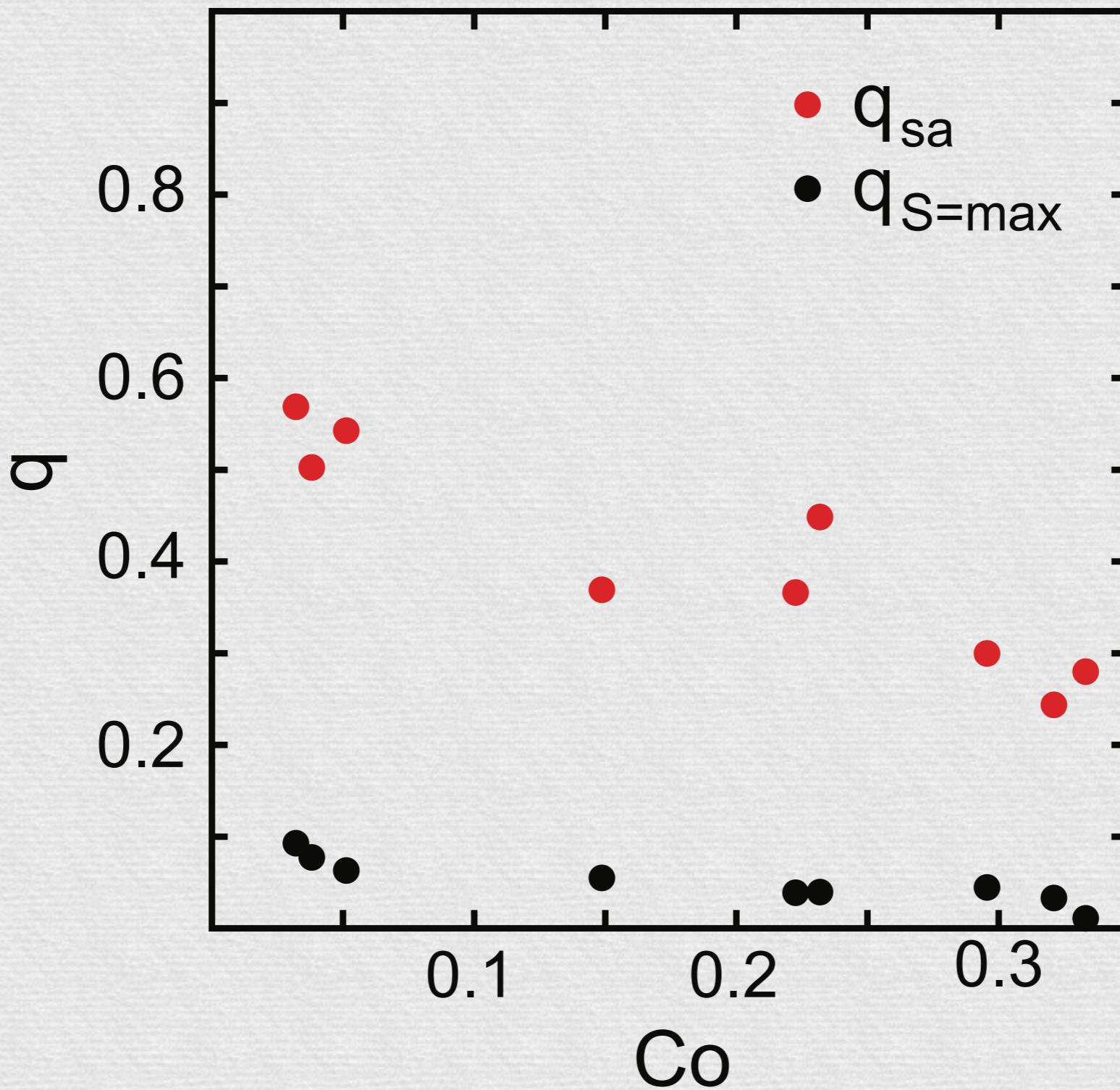


# Network Small-Worldness

Rewiring for maximal S



# Required Rewiring



# Conclusions

