## 1 4-component graph

An experiment with synthetic data generated by a 4-component graph was conducted. We consider the following model:

$$\mathbf{L}_{\mathsf{noisy}} = (1 - \kappa) \mathbf{L}_{\mathsf{true}} + \kappa \mathbf{L}_{\mathsf{ER}},\tag{1}$$

where  $\mathbf{L}_{\mathsf{true}}$  represents the Laplacian matrix of a K-component graph (for this example, K=4) denoted as  $\mathcal{G}_K^{(p_1,p_2)}$ , in which  $p_1$  and  $p_2$  represent the probabilities of node connections across components and within components, respectively;  $\mathbf{L}_{\mathsf{ER}}$  represents the Laplacian of an Erdos-Renyi graph  $\mathcal{G}_{\mathsf{ER}}^{(p)}$ , in which p is the probability of a node connecting to any other node; and  $\kappa \in (0,1)$  controls how much noise is added into  $\mathbf{L}_{\mathsf{true}}$  by the Erdos-Renyi model. Additionally, the weighted edges of both  $\mathcal{G}_K^{(p_1,p_2)}$  and  $\mathcal{G}_{\mathsf{ER}}^{(p)}$  were drawn from Uniform(0,1). Finally, we set  $p_1=0$ ,  $p_2=1$ , p=0.35, and  $\kappa=0.2$ .

Then, data were sampled in the form of  $\mathbf{Y} \sim \mathcal{N}(\mathbf{0}, \mathbf{L}_{\mathsf{noisy}}^{\dagger})$ , where  $\mathbf{A}^{\dagger}$  denotes the generalized inverse of the matrix  $\mathbf{A}$ . The total number of nodes N and the number of drawn samples T were set to N=64 (16 nodes per component) and T/N=30.

Figure ?? illustrates the ground truth model, its noisy version, and the model learned by our spectral topology algorithm with  $\beta=10$ . We compute the performance of the learning process by means of the relative error (RE) and the F-score (FS). For this example, our algorithm achives (RE, FS) = (0.131, 0.993) which means almost perfect clustering accuracy even in a noisy model that heavily supress the ground truth weights when  $\kappa$  is much larger than zero.

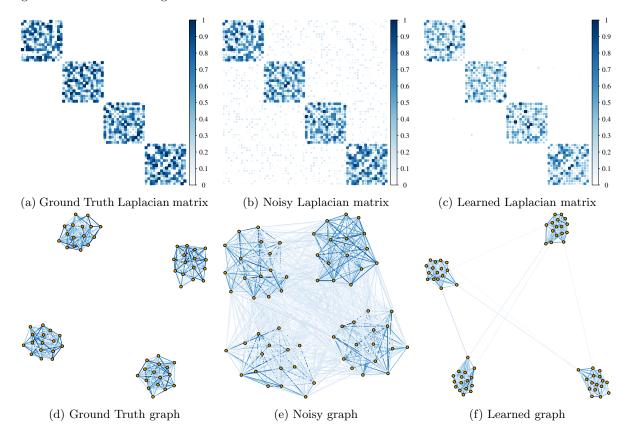


Figure 1: An example of estimating a 4-component graph. (a) the ground truth graph Laplacian matrix ( $\mathbf{L}_{\mathsf{true}}$ ), (b)  $\mathbf{L}_{\mathsf{true}}$  after being corrupted by noise, (c) the learned graph Laplacian with a performance of (RE, FS) = (0.131, 0.993).. The panels (d), (e), and (f) correspond to the graphs represented by the Laplacian matrices in (a), (b), and (c), respectively.