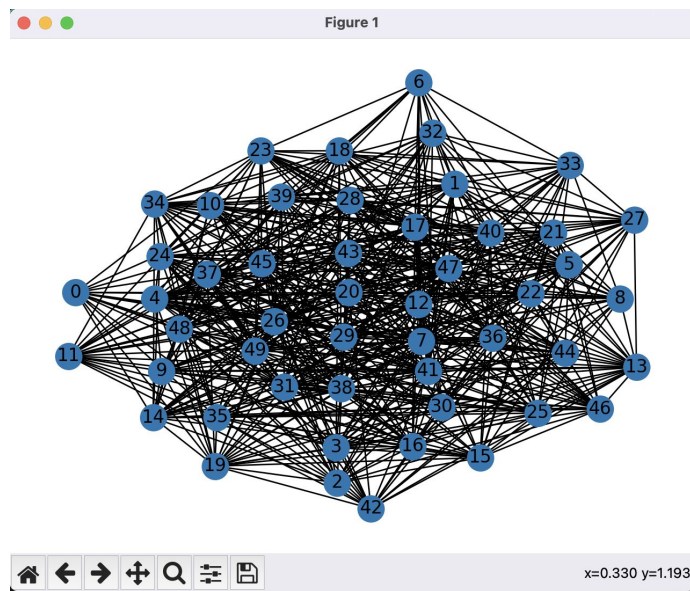




# Spectral Clustering and Cheeger's Inequality

Kaicheng Guo & Zhengyu Zou

How do you find “good” clusters in an extremely large and messy graph ?

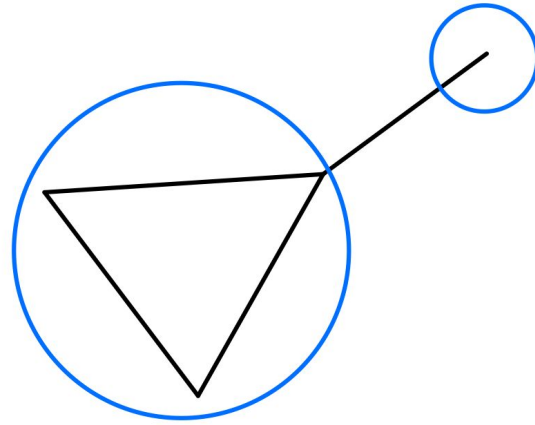
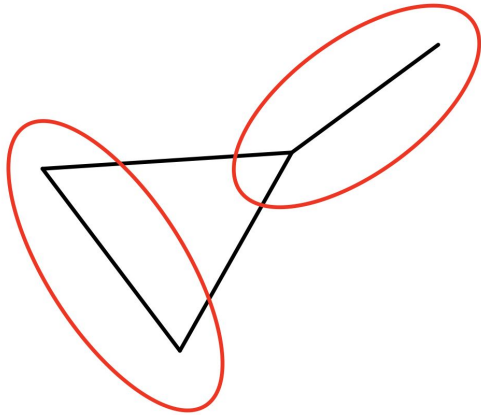




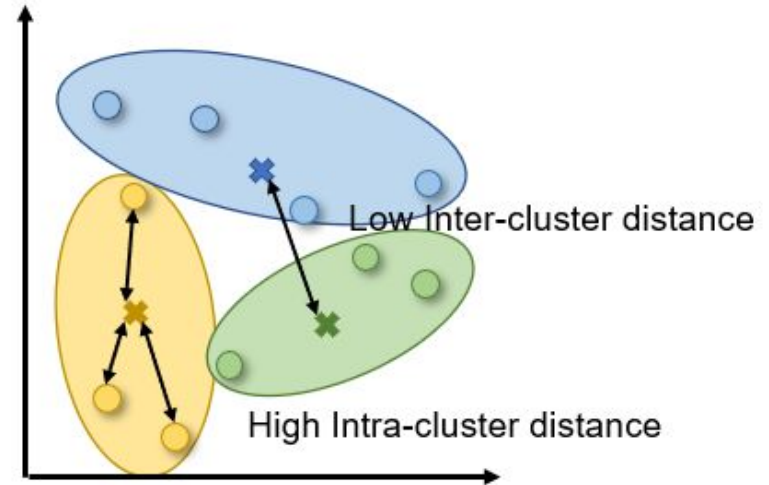
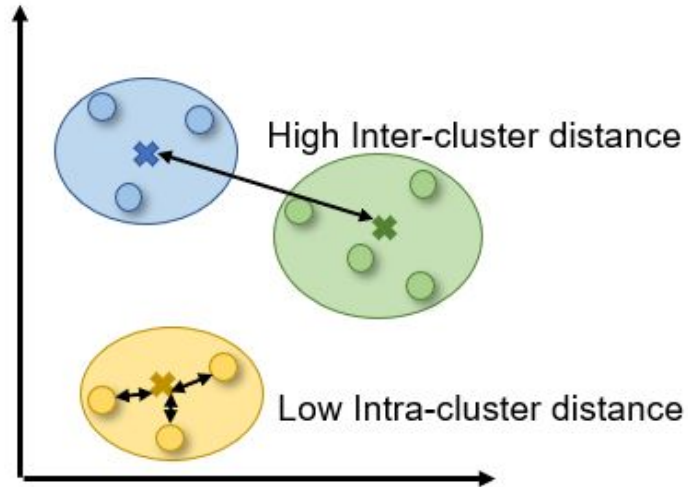
## Which Cut Gives “Good” Clusters?

- A good cluster has the following properties:
  - Only need to move a small number of edges
  - The two subgraphs have similar number of vertices / edges. I.e., they are “balanced in sizes”

## Examples of clusters :(



## Examples of Clusters :D



---

# Spectral Approach of a “Good” Cluster



## Volume of a Subgraph

- Volume = sum of the degrees of all vertices in the subgraph
- A “larger” graph has more vertices -> larger volume
- A “denser” graph has more edges -> larger volume
- This is a very good classifier of good clusters: BOTH sides needs to be large and dense (“balanced”)



## Conductance of a Cut

- The conductance is computed by the “edge flux” divided by the minimum volume of the cut and its “complement cut”
- Measures whether a cut is a “good” cut in the following sense:
  - Deleting a small number of edges – Minimizing edge flux
  - “Balanced in size” – maximizing the smaller one of the two volumes





## Conductance of a Graph

- The conductance of a graph is simply given by the minimum conductance of any possible cuts of a graph.
- A “good” cluster yields the minimum conductance

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# How to Find “Good” Clusters?



## Naive Way: Consider All Possible Clusters

- Consider every nonempty subset  $S$  of  $V$  as potential candidates of our “good” cluster
- Calculate the conductance of each candidate
- Return the candidate with the smallest conductance
- Runtime:  $O(2^{|V|}) := ($




## A Better Way: Spectral Clustering

- Our good old friend: Laplacian Matrix  $L$
- Laplacian matrix gives information about edge flux. Very useful
- Want to find a vector  $x$  such that  $x^T L x / x^T x$  is minimized
- Fact: the smallest non-zero eigenvector  $x$  gives  $\min(x^T L x / x^T x)$ 
  - Be careful!  $L$  has a 0 eigenvector.



## Algorithm (Spectral Clustering)

- Find a good approximation of the eigenvector  $x$  of the second smallest eigenvalue of  $L$  (can use the power iteration method)
- Sort the vertices of  $V$  according to  $x$ :  $x(v_1) \leq x(v_2) \leq \dots \leq x(v_n)$
- Iterate over the cuts  $S_i$  with each time adding  $v_i$  to the cut. For each iteration calculate the conductance of  $S_i$
- Return minimum of conductance



# Demonstration of the Algorithm

```
graph.py x
graph.py > ...
45     if conductance < min_conductance:
46         min_conductance = conductance
47         cluster = S
48     update_graph(graph, pos, set(vertices[:i]))
49     print("Minimum conductance:", min_conductance)
50     print("Cluster:", cluster)
51     update_graph(graph, pos, cluster, conductance)
52
53
54     plt.ioff()
55     plt.show()
56     plt.close()
57     return min_conductance, cluster
58
59 n = 20
60 p = 0.1
61 while True:
62     graph = nx.erdos_renyi_graph(n, p)
63     if is_connected(graph):
64         break
65
66 pos = nx.spring_layout(graph)
67 min_conductance, cluster = spectral_partitioning_a
68
69
```



# Applications

- PageRank: classification of websites in to topics
- Community detection: detect communities or groups of nodes with similar connectivity patterns in social networks, biological networks, and other complex networks.
- Recommendation Systems: Spectral clustering can be used in recommendation systems to group similar items or users based on their preferences, ratings, or behavior.



# References

GPT4 (for the algorithm implementation in python)

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