

CS630 Homework 5

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1 import networkx as nx
2 import numpy as np
3 from scipy.linalg import eigh
4
5 G = nx.read_gml("karate.gml", label="id")
6 mapping = {node: node - 1 for node in G.nodes()}
7 G = nx.relabel_nodes(G, mapping)
8 n = G.number_of_nodes()
9
10
11
12 A = np.zeros((n, n))
13 for u, v in G.edges():
14     A[u, v] = 1
15     A[v, u] = 1
16 for i in range(n):
17     A[i, i] = A[i, :].sum()
18
19
20 D = A.sum(axis=1)
21 D2 = np.diag(1 / np.sqrt(D))
22 N = D2 @ A @ D2
23
24
25 eigenvals, eigenvecs = eigh(N)
26 lambda2 = eigenvals[-2]
27 vec2 = eigenvecs[:, -2]
28
29 lambda2, vec2
30
31
32 def conductance(cut, vol, A):
33     Sc = ~cut
34     cut_weight = 0
35     for i in range(len(cut)):
36         if cut[i]:
37             for j in range(len(cut)):
38                 if Sc[j]:
39                     cut_weight += A[i, j]
40     vol_S = vol[cut].sum()
41     vol_Sc = vol[Sc].sum()
42     if vol_S == 0 or vol_Sc == 0:
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43     return np.inf
44     return cut_weight / min(vol_S, vol_Sc)
45
46 order = np.argsort(vec2)
47 vol = A.sum(axis=1)
48 best_cond = np.inf
49 best_cut = None
50 best_k = None
51 for k in range(1, n):
52     S = np.zeros(n, dtype=bool)
53     S[order[:k]] = True
54     c = conductance(S, vol, A)
55     if c < best_cond:
56         best_cond = c
57         best_cut = S.copy()
58         best_k = k
59 print("Spectral clustering:")
60 print(f"lambda_2 = {lambda2:.4f}")
61 print(f"1-lambda_2 = {1 - lambda2:.4f}")
62 print(f"best k = {best_k} with conductance {best_cond:.4f}")
63
64
65
66
67 # cheegar's inequality
68 # phi^2/16 vs 1-lambda_2 vs phi
69 print("Cheegar's inequality:")
70 print(f"phi^2/16 = {best_cond**2 / 16:.4f}")
71 print(f"1-lambda_2 = {1 - lambda2:.4f}")
72 print(f"phi = {best_cond:.4f}")
73
74
75 random_conds = []
76 for trial in range(10):
77     while True:
78         rand_assign = np.random.randint(0, 2, size=n).astype(bool)
79         if rand_assign.sum() > 0 and (n - rand_assign.sum()) > 0:
80             break
81         random_conds.append(conductance_custom(rand_assign, A, m))
82
83 print("Baseline random cut conductance:")
84 for i, rc in enumerate(random_conds):
85     print(f"\ttrial {i+1}: {rc:.4f}")
86
87 avg_random = np.mean(random_conds)
88 print(f"\taverage = {avg_random:.4f}")

```

Solution to problem 2. First suppose G has k connected components. For a $n = |V|$ -dimensional vector v , consider the quadratic term

$$v^T L v = \frac{1}{2} \sum_{(i,j) \in E} \left(\frac{v(i)}{\sqrt{d(i)}} - \frac{v(j)}{\sqrt{d(j)}} \right)^2.$$

Therefore, if v is an eigenvector of L with eigenvalue 0 we have $v^T L v = 0$, which forces each edge $(i, j) \in E$

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D = A.sum(axis=1)
D2 = np.diag(1 / np.sqrt(D))
N = D2 @ A @ D2

eigenvals, eigenvcs = eigh(N)
lambda2 = eigenvals[-2]
vec2 = eigenvcs[-2]

lambda2, vec2

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Spectral clustering:
lambda_2 = 0.9339
1-lambda_2 = 0.0661
best k = 16 with conductance 0.0658

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print(phi - best_cond, phi)
Cheeger's inequality:
phi^2/16 = 0.0003
1-lambda_2 = 0.0661
phi = 0.0658

```

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0.0s
Baseline random cut conductance:
trial 1: 0.1058
trial 2: 0.1202
trial 3: 0.1431
trial 4: 0.1192
trial 5: 0.1209
trial 6: 0.1391
trial 7: 0.1354
trial 8: 0.1202
trial 9: 0.0897
trial 10: 0.1179
average = 0.1212

```

to have $v(i)/\sqrt{d(i)} = v(j)/\sqrt{d(j)}$. This shows that $v(i)/\sqrt{d(i)}$ is constant on each connected component. This implies that every connected component contributes one independent eigenvector with eigenvalue 0. Thus the multiplicity of eigenvalue 0 of L is precisely k .

Conversely, suppose N has k eigenvalues equal to 0. The same equality $v(i)/\sqrt{d(i)} = v(j)/\sqrt{d(j)}$ must hold. Thus, any such v must be constant on every connected component. Since we have k linearly independent eigenvectors of L with eigenvalues 0, we must have precisely k connected components.

Solution to problem 3. If $|S| \geq k + 1$ then clearly nothing in C can recover all of S by intersection. Hence the VC dimension is bounded by k .

On the other hand, a symmetry argument implies that the VC dimension is also bounded by $n - k$.

To see this dimension can be attained, WLOG assume $k \leq n/2$. Let S be any set of size k . For any subset $A \subset S$, one can extend A to a k -element subset by adding $k - |A|$ elements from $X \setminus S$.

Solution to problem 4. (1) First consider a diamond, e.g., $\{(0, \pm 1), (\pm 1, 0)\}$. It is clear that such 4 points can be shattered by axis-aligned rectangles. On the other hand, for any set S of 5 points, assuming we can break ties arbitrarily, there exists one point with rightmost x -coordinate, one with leftmost x -coordinate, and two with largest/smallest y -coordinate. At least one out of the 5 points is not covered by the previous 4 extremes. Call it p . Then $S \setminus \{p\}$ cannot be obtained by intersecting S with some axis-aligned rectangle, since the intersection must include a continuum of x -values, making it impossible to include points with largest and smallest x -values but not p .

(2) Immediately we see the 3 vertices of an equilateral triangle can be easily shattered by the set of axis aligned squares.

To see that a 4-point set cannot, consider a four-point set S . Let p_1 be the rightmost point (w.r.t. x -coordinate), p_2 leftmost, p_3 topmost (w.r.t. y -coordinate), and p_4 bottom-most. WLOG assume $p_{1,x} - p_{2,x} \geq p_{3,y} - p_{4,y}$, i.e., the x -span is larger, making the set more horizontal. Then no square C can satisfy $C \cap S = \{p_1, p_2\}$, for they have to also include either p_3, p_4 , or both.

(3) Intuitively, to construct a 5-point set, consider one point s very far away, while the remaining 4 points p_1, p_2, p_3, p_4 are aligned along an arc, close to each other, and form a convex cone with c . For example, given $\epsilon > 0$, consider on the complex plane $c = (0, 0)$ and $p_k = e^{2k\epsilon\pi i}$ for $k \in \{0, 1, 2, 3\}$. It's trivial to shatter the $\{p_i\}$. On the other hand, for sufficiently small ϵ , it's also easy to shatter anything consisting of both c and some p_i 's. (Sorry, too lazy to draw all diagrams now...) No idea how to prove the 6-point case, except

I assume it will involve some results in convex analysis

(4) The d -dimensional simplex S can clearly be shattered. Suppose A is a subset of these $d + 1$ points. The binary vector corresponding to the complement of A (i.e. entry has 0 if corresponding dimension is not chosen) defines a half space that separates A against $S \setminus A$.

To show that a set of size $d + 2$ cannot be shattered we use Radon's theorem, which claims that any such set can be partitioned into two sets A, B whose convex hulls intersect. Therefore, any halfspace containing A must contain the convex hull of A and therefore something from B . This implies A, B cannot be shattered, completing the proof that the set of d -dimensional half spaces has VC-dim $d + 1$.

Solution to problem 5. If the VC dimension of S is ∞ then the result trivially holds. Otherwise, if a set A is shattered by R' then it certainly is shattered by R : $R' \subset R$ implies $\{A \cap C' \mid C' \in R'\} \subset \{A \cap C \mid C \in R\}$. Therefore, $\text{VCdim}(R') \leq \text{VCdim}(R)$.