

lecture

October 15, 2025

1 HC

beat greedy baseline of $\frac{1}{3} - \alpha p x$. this is average linkage or random
this alg will give $\geq .336 OPT$

$$\sum_{ij \in E} w_{ij} (n - |T_{ij}|) = \sum_{ij \in E} w_{ij} \sum_{k \neq i, j} \mathbb{1}\{k \text{ not a leaf of } T_{ij}\} \quad (1)$$

consider

$$y_{ijk} = w_{ij} \mathbb{1}\{k \text{ not a leaf of } T_{ij}\} \quad (2)$$

it only atkes on values of $w_{ij}, 0$

$$y_{ij} = \sum_{k \neq i, j} y_{ijk} \quad (3)$$

2 ALG1

random always

for a given triplet i, j, k , the probability $P[ij|k]^1$ is $\frac{1}{3}$.

3 ALG 2

SDP first, random next

solve sdp for hc, look for vectors at mid level, $t^* = \lfloor n/2 \rfloor$, then we set $x_{ij}^t = x_{ij}^{t^*}$.

do hyperplane rounding to partition into S, \bar{S}

random always S, and sdp \bar{S} .

recursive max uncut bisection gives 0.585.

¹see lec 5

remember x_{ij}^t , at level t each cluster is at most size t .

sdp objective,

$$\max \sum_{t=1}^n \sum_{ij \in E} w_{ij} (1 - x_{ij}^t) \quad (4)$$

constraints:

$$\sum_{i \neq j} x_{ij}^t \geq n - t, \forall i, \forall t \quad (5)$$

$$x_{ij}^{t+1} \leq x_{ij}^t, \forall ij \in E, \forall t, x_{ij}^{(0)} = 1 \quad (6)$$

to convert to sdp, $x^t + ij \frac{1}{2} \|v_i^t - v_j^t\|_2^2, v_i^{((t))} \in \mathbb{R}^n$

4 case 1

$$OPT < (1 - \epsilon_1)(n - 2) \sum_{ij \in E} w_{ij} \quad (7)$$

if ϵ_1 is large then just use random always

5 case2

if $SDP \geq OPT \geq (1 - \epsilon_1)W$,

consider 3 cases

ϵ_{ij} ij is together, ϵ_{ijk} ijk all together $\epsilon_{ij|k}$ ij together but k split

the final one is $\frac{\theta_{ik} + \theta_{jk} - \theta_{ij}}{2\pi}$

the expectation of 3 is

$$E[y_{ijk}] = \frac{w_{ij}}{3} P[\epsilon_{ijk}] + w_{ij} P[\epsilon_{ij|k}] \quad (8)$$

$$E[y_{ij}] = \sum_{k \neq i, j} E[y_{ijk}] \quad (9)$$

because

$$P[\epsilon_{ij}] = 1 - P[\epsilon_{ijk}] - P[\epsilon_{ij|k}] \quad (10)$$

we get that 8 is also equal to

$$\frac{w_{ij}}{3} P[\epsilon_{ij}] + \frac{2w_{ij}}{3} P[\epsilon_{ij|k}] \quad (11)$$

and 9 is

$$E[y_{ij}] = \frac{w_{ij}}{3} \left((n-2)P[\epsilon_{ij}] + 2 \sum_{k \neq i, j} P[\epsilon_{ij|k}] \right) \quad (12)$$