

Welcome back...

Metric spaces.

A metric space X , $d(i,j)$ where $d(i,j) \leq d(i,k) + d(k,j)$,
 $d(i,j) = d(j,i)$, and $d(i,j) \geq 0$.

Which are metric spaces?

- (A) X from R^d and $d(\cdot, \cdot)$ is Euclidean distance.
- (B) X from R^d and $d(\cdot, \cdot)$ is squared Euclidean distance.
- (C) X - vertices in graph, $d(i,j)$ is shortest path distances in graph.
- (D) X is a set of vectors and $d(u,v)$ is $u \cdot v$.

Input to TSP, facility location, some layout problems, ..., metric labelling.

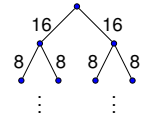
Hard problems. Easier to solve on trees. Dynamic programming on trees.

Approximate metric on trees?

Approximate metric using a tree.

Tree metric:

X is nodes of tree with edge weights
 $d_T(i,j)$ shortest path metric on tree.



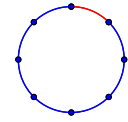
Hierarchically well separated tree metric:

Tree weights are geometrically decreasing.

Probabilistic Tree embedding.

Map X into tree.

- (i) No distance shrinks. (dominating)
- (ii) Every distance stretches $\leq \alpha$ in expectation.



Distance 1 goes to $n-1$!
Bummer.

Map metric onto tree?

Fix it up chappie!

For cycle, remove a random edge get a tree.

Stretch of edge: $\frac{n-1}{n} \times 1 + \frac{1}{n} \times (n-1) \approx 2$

General metrics?

Probabilistic Tree embedding.

Probabilistic Tree embedding.

Map X into tree.

- (i) No distance shrinks (dominating).
- (ii) Every distance stretches $\leq \alpha$ in expectation.

Today: the tree will be Hierarchically well-separated (HST).
 Elements of X are leaves of tree.

Later: use spanning tree for graphical metrics.

The Idea:

HST \equiv recursive decomposition of metric space.

Decompose space by diameter $\approx \Delta$ balls.
 Recurse on each ball for $\Delta/2$.

Use randomness in
 selection of ball centers.
 the \approx diameter of the balls.

Algorithm

Algorithm: (X, d) , $\text{diam}(X) \leq D$, $|X| = n$, $d(i,j) \geq 1$

1. π - random permutation of X .
2. Choose β in $[\frac{3}{8}, \frac{1}{2}]$.
 def subtree(S, Δ):
 $T = []$
 if $\Delta < 1$ return $[S]$
 foreach i in π :
 if $i \in S$
 $B = \text{ball}(i, \beta\Delta)$; $S = S/B$
 $T.append(B)$
 return map (λx : subtree($x, \Delta/2$), T);
3. subtree(X, D)

Tree has internal node for each level of call. Tree edges have weight Δ to children.

Claim 1: $d_T(x,y) \geq d(x,y)$.

When $\Delta \leq d(x,y)$, x and y must be in different balls, so cut at $\text{lvl } \Delta \geq d(x,y)/2$.

$\rightarrow d_T(x,y) \geq \Delta + \Delta \geq d(x,y)$

Analysis: idea

Claim: $E[d_T(x,y)] = O(\log n)d(x,y)$.

Cut at level $\Delta \rightarrow d_T(x,y) \leq 4\Delta$. (Level of subtree call.)

$Pr[\text{cut at level } \Delta]$?

Would like it to be $\frac{d(x,y)}{\Delta}$.

\rightarrow expected length is $\sum_{\Delta=D/2^i} (4\Delta) \frac{d(x,y)}{\Delta} = 4 \log D \cdot d(x,y)$.

Why should it be $\frac{d(x,y)}{\Delta}$?

smaller the edge the less likely to be on edge of ball.

larger the delta, more room inside ball.

random diameter jiggles edge of ball.

$\rightarrow Pr[x,y \text{ cut by ball} | x \text{ in ball}] \approx \frac{d(x,y)}{\beta\Delta}$

The problem?

Could be cut by many different balls.

For each probability is good, but could be hit by many.

random permutation to deal with this

Analysis: (x, y)

Would like $Pr[x, y \text{ cut by ball} | x \text{ in ball}] \leq \frac{8d(x, y)}{\Delta}$
(Only consider cut by x , factor 2 loss.)

At level Δ

At some point x is in some Δ level ball.

Renumber nodes in order of distance from x .

If $d(x, y) \geq \Delta/8$, $\frac{8d(x, y)}{\Delta} \geq 1$, so claim holds trivially.

j can only cut (x, y) if $d(j, x) \in [\Delta/4, \Delta/2]$ (else (x, y) entirely in ball),
Call this set X_Δ .

$j \in X_\Delta$ cuts (x, y) if..

$d(j, x) \leq \beta\Delta$ and $\beta\Delta \leq d(j, y) \leq d(j, x) + d(x, y)$

$\rightarrow \beta\Delta \in [d[j, x], d(j, x) + d(x, y)]$.

occurs with prob. $\frac{d(x, y)}{\Delta/8} = \frac{8d(x, y)}{\Delta}$.

And j must be before any $i < j$ in $\pi \rightarrow$ prob is $\frac{1}{j}$

$\rightarrow Pr[j \text{ cuts } (x, y)] \leq \left(\frac{1}{j}\right) \frac{8d(x, y)}{\Delta}$

$d_T(x, y)$ if cut level Δ is 4Δ .

$\rightarrow E[d_T(x, y)] = \sum_{\Delta=\frac{\Delta}{2^i}} \sum_{j \in X_\Delta} \left(\frac{1}{j}\right) 32d(x, y)$

And Now For Something...

Completely Different.

The pipes are distinct!

$$E[d_T(x, y)] = \sum_{\Delta=\frac{\Delta}{2^i}} \sum_{j \in X_\Delta} \left(\frac{1}{j}\right) 32d(x, y)$$

Recall X_Δ has nodes with $d(x, j) \in [\Delta/4, \Delta/2]$

"Listen Stash, the pipes are distinct!!"

Uh.. well X_Δ is distinct from $X_{\Delta/2}$.

$$E[d_T(x, y)] = \sum_{\Delta=\frac{\Delta}{2^i}} \sum_{j \in X_\Delta} \left(\frac{1}{j}\right) 32d(x, y)$$

$$\leq \sum_j \left(\frac{1}{j}\right) 32d(x, y)$$

$$\leq (32 \ln n) (d(x, y)).$$

Claim: $E[d_T(x, y)] = O(\log n) d(x, y)$

Expected stretch is $O(\log n)$.

We gave an algorithm that produces a distribution of trees.

The expected stretch of any pair is $O(\log n)$.

Example Problem: clustering.

- ▶ Points: documents, dna, preferences.
- ▶ Graphs: applications to VLSI, parallel processing, image segmentation.

Metric Labelling

Input: graph $G = (V, E)$ with edge weights, $w(\cdot)$, metric labels (X, d) , and costs for mapping vertices to labels $c : V \times X$.

Find an labeling of vertices, $\ell : V \rightarrow X$ that minimizes

$$\sum_{e=(u, v)} c(e) d(\ell(u), \ell(v)) + \sum_v c(v, \ell(v))$$

Idea: find HST for metric (X, d) .

Solve the problem on a hierarchically well separated tree metric.

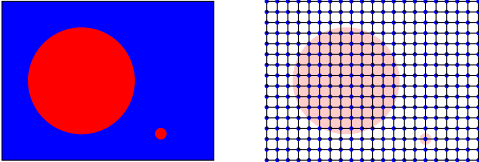
Kleinberg-Tardos: constant factor on uniform metric.

Hierarchically well separated tree, "geometric", constant factor.

$\rightarrow O(\log n)$ approximation.

Image example.

Image Segmentation



Which region? Normalized Cut: Find S , which minimizes

$$\frac{w(S, \bar{S})}{w(S) \times w(\bar{S})}$$

Ratio Cut: minimize

$$\frac{w(S, \bar{S})}{w(S)}$$

$w(S)$ no more than half the weight. (Minimize cost per unit weight that is removed.)

Either is generally useful!

Action of M .

v - assigns weights to vertices.

Mv replaces v_i with $\frac{1}{d} \sum_{e=(i,j)} v_j$.

Eigenvector with highest value? $v = \mathbf{1}$. $\lambda_1 = 1$.

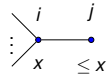
$\rightarrow v_i = (M\mathbf{1})_i = \frac{1}{d} \sum_{e=(i,j)} 1 = 1$.

Claim: For a connected graph $\lambda_2 < 1$.

Proof: Second Eigenvector: $v \perp \mathbf{1}$. Max value x .

Connected \rightarrow path from x valued node to lower value.

$\rightarrow \exists e = (i,j)$, $v_i = x$, $v_j < x$.



$$(Mv)_i \leq \frac{1}{d}(x + x \dots + v_j) < x.$$

Therefore $\lambda_2 < 1$. □

Claim: Connected if $\lambda_2 < 1$.

Proof: Assign +1 to vertices in one component, $-\delta$ to rest.

$x_j = (Mx_j) \Rightarrow$ eigenvector with $\lambda = 1$.

Choose δ to make $\sum_j x_j = 0$, i.e., $x \perp \mathbf{1}$. □

Edge Expansion/Conductance.

Graph $G = (V, E)$,

Assume regular graph of degree d .

Edge Expansion.

$$h(S) = \frac{|E(S, V-S)|}{d \min\{|S|, |V-S|\}}, h(G) = \min_S h(S)$$

Conductance.

$$\phi(S) = \frac{|E(S, V-S)|}{d|S||V-S|}, \phi(G) = \min_S \phi(S)$$

Note $n \geq \max\{|S|, |V-S|\} \geq n/2$

$$\rightarrow h(G) \leq \phi(G) \leq 2h(S)$$

Rayleigh Quotient

$$\lambda_1 = \max_x \frac{x^T M x}{x^T x}$$

In basis, M is diagonal.

Represent x in basis, i.e., $x_i = x \cdot v_i$.

$$x M x = \sum_i \lambda_i x_i^2 \leq \lambda_1 \sum_i x_i^2 = \lambda_1 x^T x$$

Tight when x is first eigenvector. □

Rayleigh quotient.

$$\lambda_2 = \max_{x \perp \mathbf{1}} \frac{x^T M x}{x^T x}$$

$$x \perp \mathbf{1} \leftrightarrow \sum_i x_i = 0.$$

Example: 0/1 Indicator vector for balanced cut, S is one such vector.

$$\text{Rayleigh quotient is } \frac{|E(S, S)|}{|S|} = h(S).$$

Rayleigh quotient is less than $h(S)$ for any balanced cut S .

Find balanced cut from vector that achieves Rayleigh quotient?

Spectra of the graph.

$M = A/d$ adjacency matrix, A

Eigenvector: $v - Mv = \lambda v$

Real, symmetric.

Claim: Any two eigenvectors with different eigenvalues are orthogonal.

Proof: Eigenvectors: v, v' with eigenvalues λ, λ' .

$$v^T M v' = v^T (\lambda' v') = \lambda' v^T v'$$

$$v^T M v' = \lambda v^T v' = \lambda v^T v'$$

Distinct eigenvalues \rightarrow orthonormal basis. □

In basis: matrix is diagonal..

$$M = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

Cheeger's inequality.

Rayleigh quotient.

$$\lambda_2 = \max_{x \perp \mathbf{1}} \frac{x^T M x}{x^T x}$$

Eigenvalue gap: $\mu = \lambda_1 - \lambda_2$.

Recall: $h(G) = \min_{S, |S| \leq |V|/2} \frac{|E(S, V-S)|}{|S|}$

$$\frac{\mu}{2} = \frac{1-\lambda_2}{2} \leq h(G) \leq \sqrt{2(1-\lambda_2)} = \sqrt{2\mu}$$

Hmmm..

Connected $\lambda_2 < \lambda_1$.

$h(G)$ large \rightarrow well connected $\rightarrow \lambda_1 - \lambda_2$ big.

Disconnected $\lambda_2 = \lambda_1$.

$h(G)$ small $\rightarrow \lambda_1 - \lambda_2$ small.

Easy side of Cheeger.

Small cut \rightarrow small eigenvalue gap.

$$\frac{h}{2} \leq h(G)$$

Cut S : $i \in S : v_i = |V| - |S|$, $i \in \bar{S} v_i = -|S|$.

$$\sum_i v_i = |S|(|V| - |S|) - |S|(|V| - |S|) = 0$$

$\rightarrow v \perp \mathbf{1}$.

$$v^T v = |S|(|V| - |S|)^2 + |S|^2(|V| - |S|) = |S|(|V| - |S|)(|V|).$$

$$v^T M v = \frac{1}{d} \sum_{e=(i,j)} x_i x_j.$$

Same side endpoints: like $v^T v$.

Different side endpoints: $-|S|(|V| - |S|)$

$$v^T M v = v^T v - (2|E(S, \bar{S})| |S|(|V| - |S|))$$

$$\frac{v^T M v}{v^T v} = 1 - \frac{2|E(S, \bar{S})|}{|S|}$$

$$\lambda_2 \geq 1 - 2h(S) \rightarrow h(G) \geq \frac{1 - \lambda_2}{2}$$

See you ...

Thursday.