

Linnal - Meshulam model of random complex.

↳ Erdős-Rényi model of random graph!

Take n points, $\forall 2$ vertices (i,j)

$P((i,j) \text{ is an edge}) = p.$

$X(n,p)$ - "the" random graph on n vertices

with every edge appear w/ prob. $p.$

$\frac{2}{9} \binom{12}{3} < 25^6$
 $\frac{n(n-1)(n-2)}{6} < 9 \cdot 12^3$
 $\frac{n(n-1)(n-2)}{6} < 27 \cdot 25^3$
 $n = 24$

Thus if $p < \frac{\log n}{n} - \epsilon$ then $X(n,p)$ is a.s. not connected

if $p > \frac{\log n}{n} + \epsilon$, then a.s. connected.

L-M

↳ Take n pts, take all pairs, i.e. all edges.

$X_2(n,p)$ = each tr. is taken ind. w/ prob. $p.$

When $H_1(X_2(n,p), \mathbb{F}_2) = \text{dor.}?$

Thus if $p < 2 \frac{\log n}{n} - \epsilon \Rightarrow$ a.s. $\neq 0$

$p > 2 \frac{\log n}{n} + \epsilon \Rightarrow$ a.s. $= 0$

Meshulam - Wallach

d -dim case.

$\leq d \dots \Rightarrow H_{d-1}(\dots) \neq 0$

$> d \dots \Rightarrow = 0$

Uriya First.

22/10

One-dimensional expanders (expander graphs)

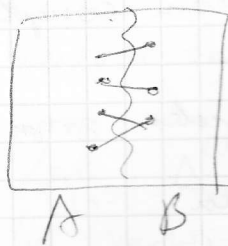
Def Graph = 1-dim complex, $X = G = (V, E)$

$X(0) = V, X(1) = E$

X -graph, the Cheeger constant:

$$h(X) = \min_{\substack{A \sqcup B = V \\ \min\{|A|, |B|\} \geq \epsilon}} \frac{|E(A, B)|}{\min\{|A|, |B|\}}$$

X is an ϵ -expander $\Leftrightarrow h(X) \geq \epsilon$



Defn The adj matrix of X :

$$A = (a_{ij})_{i,j \in V} \quad a_{ij} = \# \text{ edges from } i \text{ to } j$$

A is symmetric $\Rightarrow A$ is diagonalizable with real spectrum

$$-k \leq \lambda_0 \leq \lambda_1 \leq \dots \leq \lambda_{n-1} = k$$

\uparrow if X is k -regular \uparrow if X is k -regular

Remark when X is k -regular λ_0, λ_{n-2} control how fast a random walk on X converges to the uniform distribution on V

P -prob on the vertices of X

$$P \xrightarrow{\text{one step}} \frac{1}{k} AP$$

$$p = \sum d_i v_i, \text{ where } Av_i = \lambda_i v_i$$

$$\left(\frac{1}{k} A\right)^r p = \frac{1}{k^r} \sum d_i \lambda_i^r v_i \sim k^{-r} \lambda_{n-1}^r v_{n-1}$$

$(|\lambda_{n-2}|, |\lambda_0| < \lambda_{n-1})$

since $\left(\frac{\lambda_i}{k}\right)^r \leq 1$

this is a constant vector

Def Re Laplacian

Let $C_i(X) = L_2(X|V)$ $i=0,1$
 Hilbert space

Fix an order $v_1 \leq v_2 \leq \dots \leq v_n$ on V

Define $S_0: C_1(X) \rightarrow C_0(X)$

$$(S\psi)(e) = \psi(e^+) - \psi(e^-), \text{ where } e^+ \leq e^-$$

We also have a dual map

$$Q(X) \xrightarrow{S^*} C_0(X)$$

$$(\delta^* \varphi)(v) = \sum_{v=e^+} f(e) - \sum_{v=e^-} f(e)$$

The Laplacian of X is

$$\Delta = \Delta^* = \delta^* \delta: C_0(X) \rightarrow C_0(X)$$

Claim: Δ is independent of the ordering ^{on} V

Proof let us compute:

$$(\Delta \varphi)_v = \delta^* \delta(\varphi)(v) = \sum_{v=e^+} (\delta \varphi)(e) - \sum_{v=e^-} (\delta \varphi)(e) =$$

$$= \sum_{v=e^+} (\varphi(e^+) - \varphi(e^-)) - \sum_{v=e^-} (\varphi(e^+) - \varphi(e^-)) =$$

$$= \sum_{v \in E} \varphi(v) - \varphi(v) \deg(v) \leftarrow \text{clearly indep. of the ordering on } V. \quad \square$$

$$(\Delta \varphi)_v = \sum_{\substack{u \in V \\ u \sim v}} (\varphi(v) - \varphi(u)) = \deg(v) \cdot \varphi(v) - (\Delta \varphi)(v)$$

Since $\Delta = \delta^* \delta$, Δ is symm. and positive definite
 \Rightarrow diagonalizable w/ non-negative real spectrum

$$0 = \lambda_0 \leq \lambda_1 \leq \dots \leq \lambda_{n-1}$$

Dfn $\lambda(X) = \lambda_1(\Delta)$ is called the spectral gap of X .

Note when X is k -regular, $\lambda(X) = \lambda_{n-1}(A) - \lambda_{n-2}(A)$

Thm $\frac{k(X)^2}{2m} \leq \lambda(X) \leq 2k(X)$, $m = \max_v \deg v$

Informally: being a good expander \sim having large spectral gap.

Lemma $L_2^0(V) = \{ \varphi \in L_2(V) : \sum \varphi(v) = 0 \}$

Then $\lambda(X) = \inf_{0 \neq \varphi \in L_2^0(V)} \frac{\|\delta\varphi\|_2^2}{\|\varphi\|_2^2}$

Proof wlog assume $\|\varphi\|_2 = 1$

$$\begin{aligned} \|\delta\varphi\|_2^2 &= \langle \delta\varphi, \delta\varphi \rangle = (\text{inner prod of } L_2(V)) = \\ &= \langle \varphi, \delta^* \delta\varphi \rangle = \langle \varphi, \Delta\varphi \rangle \end{aligned}$$

The inf of $\langle \varphi, \Delta\varphi \rangle$ is the lowest e.v. of Δ on $L_2^0(V)$ which is $\lambda_1(\Delta) = \lambda(X)$ \square

Explanation of (*): Δ is self-adjoint, diagonalizable

$\Rightarrow L_2^0(V)$ has orthonormal basis $\psi_1, \dots, \psi_{n-1}$
with $\Delta\psi_i = \lambda_i \psi_i$

where $\varphi = \sum_{i=1}^{n-1} d_i \psi_i$

$$\begin{aligned} \langle \varphi, \Delta\varphi \rangle &= \langle \sum d_i \psi_i, \sum d_i \lambda_i \psi_i \rangle = \\ &= \sum d_i^2 \lambda_i \langle \psi_i, \psi_i \rangle = \sum_{i=1}^{n-1} d_i^2 \lambda_i = \star \\ &\quad (\text{keep in mind: } \sum d_i^2 = 1) \end{aligned}$$

inf $\star = \lambda_1$. \square

Proof of $(\lambda(X) \leq 2h(X))$:

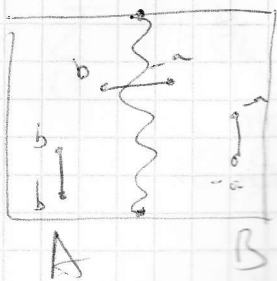
Recall $h(X) = \min_{A \cup B = V} \frac{|E(A, B)|}{\min\{|A|, |B|\}}$

let $V = A \cup B$, $a = |A|$, $b = |B|$, w.l.o.g. $a \leq b$

Define $\varphi \in L_2^0(V)$ by
$$\varphi(v) = \begin{cases} b & v \in A \\ -a & v \in B \end{cases}$$

$$\text{Compute } \frac{\|\delta\varphi\|_2^2}{\|\varphi\|_2^2} = \frac{\sum_e |\delta\varphi(e)|^2}{\sum_v |\varphi(v)|^2} = \frac{\sum_e |\varphi(e^+) - \varphi(e^-)|^2}{a \cdot b^2 + b \cdot a^2}$$

$$= \frac{n^2 \cdot \# \text{ edges from } A \text{ to } B}{nab} = n \frac{|E(A, B)|}{a \cdot b} = \frac{n |E(A, B)|}{a \cdot b} \leq$$



$$\leq \frac{n |E(A, B)|}{a \cdot \binom{n}{2}} \leq \frac{2 \cdot |E(A, B)|}{|A|} =$$

same $\frac{n}{b} \leq \frac{1}{2}$

$$= 2 \cdot \frac{|E(A, B)|}{\min(|A|, |B|)}$$

This is true for all partitions $V = A \cup B$, so

$$\lambda(x) = \inf \frac{\|S\phi\|^2}{\|\phi\|^2} \leq 2h(x)$$

Proof of (2):

Let $g \in L_2^0(V)$ be the e.f. of Δ with e.v. $\lambda_1(\Delta)$ and with $\|g\| = 1$.

Define

$$f = \max\{g, 0\}, \quad f(v) = \begin{cases} g(v) & , g(v) \geq 0 \\ 0 & , \text{otherwise} \end{cases}$$

$V^+ = \{v \in V \mid g(v) > 0\}$. w.l.o.s. $|V^+| \leq \frac{1}{2}|V|$, otherwise replace g with $-g$.

$$\begin{aligned} \|Sf\|^2 &= \langle Sf, Sf \rangle = \langle f, \Delta f \rangle = \sum_{v \in V} f(v) \Delta f(v) \\ &= \sum_{v \in V^+} g(v) \Delta f(v) = \sum_{v \in V^+} g(v) \sum_{w \in V} \underbrace{(f(v) - f(w))}_{g(v)} \underbrace{\frac{1}{g(w)}}_V \leq \\ &\leq \sum_{v \in V^+} g(v) \sum_{w \in V} (g(v) - g(w)) = \sum_{v \in V^+} g(v) \Delta g(v) \\ &= \sum_{v \in V^+} g(v)^2 \lambda_1 = \lambda_1 \|f\|^2 \end{aligned}$$

Conclusion I: $\|Sf\|^2 \leq \lambda_1 \|f\|^2$

$$\textcircled{A} := \sum_{e \in E} |f(e^+)|^2 - |f(e^-)|^2 =$$

$$= \sum_e |f(e^+) + f(e^-)| \cdot |f(e^+) - f(e^-)| \leq$$

$$\leq \left(\sum_e |f(e^+) + f(e^-)|^2 \right)^{\frac{1}{2}} \left(\sum_e |f(e^+) - f(e^-)|^2 \right)^{\frac{1}{2}} =$$

$$\leq \left(\sum_e (2f(e^+)^2 + 2f(e^-)^2) \right)^{1/2} \cdot \|Sf\| \leq$$

\nearrow same
 $(a+b)^2 \leq 2a^2 + 2b^2$

\nearrow by the conclusion

$$\leq \sqrt{2} \left(\sum f(e^+)^2 + f(e^-)^2 \right)^{1/2} \cdot \|Sf\| \leq$$

$$\leq \sqrt{2} \left(m \cdot \sum_{v \in V} f(v)^2 \right)^{1/2} \cdot \|Sf\|^2 = \sqrt{2m} \|f\| \cdot \|Sf\| \leq$$

$$\leq \sqrt{2m} \cdot \sqrt{\lambda_1} \cdot \|f\|^2$$

Conclusion II: $(\star) \leq \sqrt{2m\lambda_1} \cdot \|f\|^2$

Let $0 \leq \beta_0 < \beta_1 < \dots < \beta_n$ be the values of f .

Let $L_i = \{v \in V \mid f(v) \geq \beta_i\}$

$$V = L_0 \supset L_1 \supset L_2 \supset \dots \supset L_n$$

$$|L_1| \leq \frac{1}{2} |V|, \text{ since } |L_1| = |V^T|$$

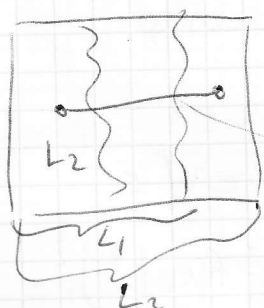
$$(\star) = \sum_{e \in E} |f(e^+) - f(e^-)|^2 =$$

$$= \sum_{i=1}^n \sum_{x \sim y} (f(x)^2 - f(y)^2) =$$

$$\begin{matrix} f(x) = \beta_i \\ f(y) < \beta_i \end{matrix}$$

$$\beta_i^2 - \beta_j^2 = \beta_i^2 - \beta_{i-1}^2 + \beta_{i-1}^2 - \beta_{i-2}^2 + \dots + (\beta_{j+1}^2 - \beta_j^2)$$

$$= \sum_{i=1}^n \sum_{e \in \partial L_i} (\beta_i^2 - \beta_{i-1}^2)$$



contributes $\beta_1^2 - \beta_0^2$ to $\sum_{e \in \partial L_1}$
 $\beta_2^2 - \beta_1^2$ to $\sum_{e \in \partial L_2}$

$$= \sum_{i=1}^r |E(L_i, V \setminus L_i)| \cdot (\beta_i^2 - \beta_{i-1}^2)^2 \geq$$

$$\geq \sum_{i=1}^r h(x) \cdot |L_i| \cdot (\beta_i^2 - \beta_{i+1}^2) =$$

$$\stackrel{\text{same}}{|L_i| \leq \frac{1}{2}|V|} \quad |L_i| \leq \frac{1}{2}|V|$$

$$= h(x) \left[|L_r| \cdot \beta_r^2 + \sum_{i=1}^r \beta_i^2 \cdot (|L_i| - |L_{i+1}|) \right] =$$

$$\# \{v \in V \mid f(v) = \beta_i\}$$

$$= h(x) \sum_{v \in V} f(v)^2 = h(x) \cdot \|f\|^2 \leftarrow \text{conclusion II}$$

$$\Rightarrow \left(\text{II} + \text{III} \right) \Rightarrow \|f\|^2 h(x) \leq \sqrt{2m\lambda_1} \cdot \|f\|^2$$

$$\Downarrow$$

$$\lambda_1(x) \geq \frac{h(x)^2}{2m} \quad \square$$

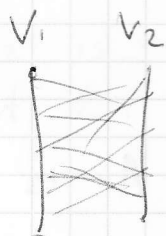
Lubotzky

29/10/13

X - a finite graph $A = A_X = \text{adj matrix of } A$

If X is k -regular then k is the largest e.v. with mult 1 iff X is connected and $-k$ is an e.v. $\Leftrightarrow X$ is bipartite.

\Leftarrow



$$|E| = |V_2| \cdot k = |V_1| \cdot k \Rightarrow |V_1| = |V_2|$$

$$f(x) = \begin{cases} 1 & x \in V_1 \\ -1 & x \in V_2 \end{cases} \Rightarrow A f = (-k) f$$

Moreover, in this case the $\text{spec}(A)$ is sym, i.e.

if $\lambda \in \text{Spec}$ then $-\lambda \in \text{Spec}$, because if

$f \in L^2(V)$ with $Af = \lambda f$, then

$$\bar{f}(x) = \begin{cases} f(x) & x \in V_1 \\ -f(x) & x \in V_2 \end{cases}$$

$$A \bar{f} = (-\lambda) \bar{f}$$