

Today

Spectral Methods (Part III)

- Cayley Graphs
→ Spectrum
- Random Walk
Matrix

CS5.205.1

Toolkit in TCS

- Lecture #22

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Cayley Graphs:

G -group (finite set of elements
w/ a binary operation

$$G = (S, \cdot)$$

$$\cdot : G \times G \rightarrow G$$

$$a \cdot b \mapsto a \cdot b$$

(1) \cdot has an identity elt

$$\exists e \in G, \quad e \cdot a = a \cdot e = a \\ \forall a \in G$$

(2) Every elt has an inverse

$$\forall a \in G, \exists b \in G, \quad a \cdot b = b \cdot a = e$$

(3) Associativity: $(a \cdot b) \cdot c = a \cdot (b \cdot c)$

If furthermore \cdot is commutative,

then G is an Abelian group

eg: (1) $G = \mathbb{Z}/p\mathbb{Z}$, $\cdot = +$

(2) $G = \{0,1\}^n$; $\cdot = +$ ^{xs}

$$(x_1 \dots x_n) + (y_1 \dots y_n) = (x_1 + y_1, \dots, x_n + y_n)$$

↳ non operation.

Non-Abelian Groups.

(3) $GL_n(\mathbb{R})$ - set of non-singular $n \times n$ matrices w/ real entries } infinite

\cdot - product

(4) $GL_n(\mathbb{Z}/p\mathbb{Z})$

$S \subseteq G$ - set of generators for G
 if every element in G can be obtained by taking elts in S and applying the group operation

eg: (1) $G = (\{0,1\}^n, +)$

$S = \{e_1, \dots, e_n\}$ $e_i = (0 \dots 1 \dots 0)$
 \downarrow
 i^{th} location

(2) $G = (\mathbb{Z}/n\mathbb{Z}, +)$
 n -natural number.

$S = \{1\}$

\overline{S} is closed under inverse if
 $\forall s \in S \Rightarrow s^{-1} \in S.$

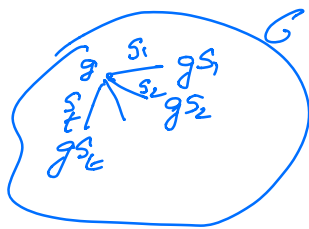
Cycle	\cong	Boolean hypercube
C_n - n cycle		H_n
$V = \mathbb{Z}/n\mathbb{Z}$		$V = \{0,1\}^n$
$(a,b) \in E$		$(a,b) \in E$
if $a = b + 1$		if $a = b + e_c$ for some $c \in [n]$
$\cdot \Rightarrow$		
$b = a + 1$		

Cayley Graph : G -group ; S -set of generators of G .
 (S is closed under inverse)

Cay (G, S) :

$V = G$

$E = \{(g, gs) \mid g \in G, s \in S\}$



Since S is closed under inverse $(g, gs) \in E \Rightarrow (gs, g) \in E$

$\text{Cay}(G, S)$ - undirected graph

$$C_n = \text{Cay}(\langle \mathbb{Z}/n\mathbb{Z}, + \rangle, \{1, -1\})$$

$$H_n = \text{Cay}(\langle \{0, 1\}^n, + \rangle, \{e_1, \dots, e_n\})$$

Heisenberg Group:

$$H_3 = \left\{ \begin{bmatrix} 1 & x & y \\ 0 & 1 & z \\ 0 & 0 & 1 \end{bmatrix} \mid x, y, z \in \mathbb{Z} \right\}$$

Group Operation = +

$$T = \left\{ \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} \right\}$$

$$S = T \cup T^{-1}$$

Cayley Graphs over Abelian Groups

Character:

$$\chi: G \rightarrow \mathbb{C} \text{ (complexes)}$$

$$\chi(g_1 g_2) = \chi(g_1) \cdot \chi(g_2)$$

eg: $G = (\mathbb{Z}/p\mathbb{Z}, +)$; $\chi_k: G \rightarrow \mathbb{C}$
 p - prime ; $a \rightarrow e^{\frac{2\pi i a k}{p}}$
for each $k \in \mathbb{Z}/n\mathbb{Z}$

$$G = (\{0,1\}^n, +) \quad ; \quad \chi_s: G \rightarrow \mathbb{C}$$

$$(x_1 \dots x_n) \mapsto \prod_{i \in S} (-1)^{x_i}$$

$S \subseteq [n]$ (parity fn)

$$\mathcal{F} = \{f: G \rightarrow \mathbb{C}\}$$

\mathcal{F} is $|G|$ -dimensional vector space.

Consider the inner product on \mathcal{F}

$$\langle f, h \rangle = \sum_{g \in G} \overline{f(g)} h(g).$$

Proposition:

- (1) χ, χ' - character
- $\Rightarrow \chi \cdot \chi'$ - character
- $\Rightarrow \chi^{-1}$ - character

- (2) $\chi_0: G \rightarrow \mathbb{C}$ } trivial character.
 $g \mapsto 1$

Set of characters form a group.

$$(3) \quad \sum_{g \in G} \chi(g) = \begin{cases} |G|, & \chi \text{ - trivial} \\ 0, & \chi \text{ - non-trivial} \end{cases}$$

Pf. $\sum_{g \in G} \chi(g)$ (χ is not trivial)

$$= \sum_{g \in G} \chi(gg') \quad \exists g' \in G, \chi(g') \neq 1$$

$$= \sum_{g \in G} \chi(g) \chi(g')$$

$$= \chi(g') \sum_{g \in G} \chi(g)$$

$$(4) \langle \chi, \chi' \rangle = \begin{cases} |G| & \text{if } \chi = \chi' \\ 0 & \text{otherwise} \end{cases}$$

$$\langle \chi, \chi' \rangle = \sum_g \overline{\chi(g)} \chi'(g)$$

$$= \sum_g \chi'(g) \chi(g)$$

$$= \sum (\chi^{-1} \chi')(g)$$

$$= \begin{cases} |G| & \text{if } \chi^{-1} \chi' = \text{trivial} \\ 0 & \text{otherwise} \end{cases}$$

Characters are orthogonal

Abelian Group.

Group of characters \cong Group \mathcal{G}

For Abelian groups

Characters form an orthogonal basis for \mathbb{C}^G .

Back to Cayley Graph

$H = \text{Cay}(G, S)$ · S -set of generators
of G closed under
inverses.

A_H - adjacency matrix

χ -character.

$$\begin{aligned}(A_H \chi)(g) &= \sum_{s \in S} \chi(gs) = \sum_{s \in S} \chi(g)\chi(s) \\ &= \chi(g) \underbrace{\left(\sum_{s \in S} \chi(s) \right)}_{\text{independent of } g}\end{aligned}$$

Hence χ is an eigen vector of A_H .
w/ eigen value $\sum_{s \in S} \chi(s)$.

- For any Cayley graph on G
(irrespective of what S)
the set of e.vectors = set of characters

$$\chi \text{ is a e.vec of } \begin{cases} A_H & \text{w/ e.vd } \sum_{s \in S} \chi(s) \\ L_H & \text{w/ e.vd } |S| - \sum_{s \in S} \chi(s) \end{cases}$$

Random Walk Matrix

G - undirected (possibly weighted) graph.
- non-negative weights.

Random Walk: From a vertex u go to vertex v
w/ prob proportional to $A(u, v)$

$$\text{Prob}[u \rightarrow v] = \frac{A(u, v)}{\sum_{v' \in V} A(u, v')}$$

- Random Walk Matrix

$$D_G^{-1} A_G(u, v) = \frac{A_G(u, v)}{\text{deg}(u)} \left. \vphantom{\frac{A_G(u, v)}{\text{deg}(u)}}} \right\} \text{on weighted graphs}$$
$$= P_n[u \rightarrow v]$$

$$W(u, v) = P_n[u \rightarrow v]$$

For unweighted graphs. $W_G = D_G^{-1} A_G$

Even for weighted graphs

$$\deg(v) = \sum_{u \in V} A(v, u)$$

$$P_n[u, \rightarrow v] = \frac{A_G(u, v)}{\deg(u)}$$

$$D_G = \text{Diag}(\deg(v))$$

Random Walk Matrix (arising from a weighted undirected graph)

$$W_G = D_G^{-1} A_G$$

Obs: A general random walk matrix.

$$\left. \begin{array}{l} W \in \mathbb{R}^{n \times n} \\ \forall u, v, W(u, v) \geq 0 \\ \forall u, \sum_{v \in V} W(u, v) = 1 \end{array} \right\}$$

Not all r.w matrices come from weighted undirected graphs.

Right Multiplication

$$W \in \mathbb{R}^{V \times V} \text{ - r.w matrix}$$

$$f: V \rightarrow \mathbb{R} \quad ; \quad f \in \mathbb{R}^V$$

Wf - Right multiplication by f

$$(Wf)(u) = \sum_{v \in V} W(u, v) f(v)$$

Right multiplication - corresponds
to averaging accg
to random
walk

$$W\mathbb{1} = \mathbb{1}$$

$\mathbb{1}$ is a right eigenvector
w/ eigenvalue 1.

Left Multiplication:

$$p: V \rightarrow \mathbb{R}$$

$$(pW)(v) = \sum_{u \in V} p(u) W(u, v)$$

Suppose p is a prob. dist on
vertices.

$$(pW)(v) = \text{Prob}[\text{RW lands on } v \text{ when started accord to } p]$$

$$p \xrightarrow{\text{1st step}} pW \xrightarrow{\text{2nd step}} pW^2 \rightarrow pW^3$$

Suppose there exists a prob dist π

$$\text{i.e., } \pi = \pi W$$

π - stationary distribution

- left eigenvector of W w/ λ -value 1

For weighted graphs,

$$\left. \begin{array}{l} \pi \sim \text{deg} \\ \rightarrow \text{stationary} \end{array} \right\}$$

$$\pi(v) = \frac{\text{deg}(v)}{\sum_{u \in V} \text{deg}(u)} = \frac{\text{deg}}{\Delta}$$

Δ - Total degree

$$\text{deg } W = \text{deg}$$

Eigen Spectrum of Random Walk Matrix:

π -Inner Product

$$\langle f, g \rangle_{\pi} = \sum_{v \in V} \pi(v) f(v) g(v)$$

When is W self-adjoint under \langle, \rangle_π

What is adjoint W^* ?

$$\langle f, Wg \rangle = \langle W^*f, g \rangle, \quad \forall f, g \in \mathbb{R}^V$$

$$f = \mathbb{1}(v_1) \quad ; \quad g = \mathbb{1}(v_2)$$

$$\begin{aligned} \langle f, Wg \rangle &= \sum_v \pi(v) f(v) (Wg)(v) \\ &= \pi(v_1) Wg(v_1) \end{aligned}$$

$$= \pi(v_1) W(v_1, v_2)$$

$$\begin{aligned} \langle W^*f, g \rangle &= \pi(v_2) (W^*f)(v_2) \\ &= \pi(v_2) W^*(v_2, v_1) \end{aligned}$$

Hence, W^* is the matrix

$$W^*(v_2, v_1) = \frac{\pi(v_1) W(v_1, v_2)}{\pi(v_2)}$$

random walk matrix corresp
to time-reversal:

$$P \mapsto PW$$

$$P = P W W^* \leftarrow P W$$

If we want W to be self-adjoint

$$W^* = W.$$

$$W(v_2, v_1) = \frac{\pi(v_1) W(v_1, v_2)}{\pi(v_2)} \quad \forall v_1, v_2$$

Nicer: $\pi(v_2) W(v_2, v_1) = \pi(v_1) W(v_1, v_2)$

\hookrightarrow Detailed Balance condition

\rightarrow Reversible Random Walk

Random Walks arising from weighted (undirected) graphs.

$$\pi(v_i) = \frac{\deg(v_i)}{\sum \Delta} \quad W(v_1, v_2) = \frac{A(v_1, v_2)}{\deg(v_1)}$$

$$\pi(v_1) W(v_1, v_2) = \frac{A(v_1, v_2)}{\sum \Delta} = \pi(v_2) W(v_2, v_1)$$

$\underbrace{\hspace{10em}}_{\text{undirected}}$

Spectral Theorem guarantees an orthonormal eigen decomposition

for any self-adjoint RW matrices,
(under π -dist).

RW matrices arising from weighted
undirected graphs are
self-adjoint
(reversible)
(detailed balance).

→ Eigen decomposition.

$v_1, \dots, v_n \in \mathbb{R}^n$ w/ eigen values
 $\omega_1, \dots, \omega_n \in \mathbb{R}$

Obs.: (1) $\mathbb{1}$ w/ e-value 1 is a
right e-vector.

(2) All e-values have absolute
value ≤ 1

(3) $\mathbb{1}$ is the right eigenvector
corresponding to 1
w/ largest e-value

Right Eigen space of W

$$\begin{aligned}
 \mathbb{I} &= v_1, v_2, \dots & v_n \\
 \mathbb{I} &= \omega_1, \omega_2, \dots & \omega_n \geq -\mathbb{I}
 \end{aligned}$$

Left Eigen Space of W

- Next lecture.