Sheet 3: The Adjacency matrix

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All graphs are finite on this sheet. Let G be a directed graph on the vertex set $\{1, \ldots, n\}$. Let us define the *adjacency matrix* $A = \operatorname{Adj}(G)$ by setting

 $A_{i,j}$ = number of edges from i to j in G.

So, we allow multiple edges and even loops in G.

Exercise 1 Express the following in linear algebra terms, using A:

- 1) the degrees of a vertex;
- 2) the number of edges in G;
- 3) e(X, Y) for $X, Y \subseteq V(G)$.

One of the main reasons why we look at the adjacency (or neighboring) relation as a matrix is the following correspondence between matrix multiplication and walks in G.

Definition 2 A directed walk of length n in G is a sequence of directed edges e_1, \ldots, e_n such that $e_i^+ = e_{i+1}^-$ ($1 \le i \le n-1$). The walk is a loop (or returning), if $e_1^- = e_n^+$.

Note that we redefine the notion of walk here: it is a sequence of edges rather than vertices.

Theorem 3 For every k > 0, $(A^k)_{i,j}$ equals the number of directed walks of length k from i to j.

Proof. Let's prove by induction.

When k = 1, $(A)_{ij}$ is indeed the number of directed walks of length 1, (which is just directed edge) from i to j by definition.

Assume true for k-1.

Denote the ijth entry of A by a_{ij} and ijth entry of A^{k-1} by b_{ij}

$$(A^k)_{i,j} = (A \cdot A^{k-1})_{i,j} = \sum_{\alpha=1}^n a_{i\alpha} b_{\alpha j}$$
 (1)

Thus, for fixed α , we multiply together $a_{i\alpha}b_{\alpha j}$, which means multiply number of directed walks of length 1 (which are directed edges) from i to α and the number of directed walks of length k-1 from α to j. After this, we sum over α , which run through 1 to n, and the result $\sum_{\alpha=1}^{n}a_{i\alpha}b_{\alpha j}$ is indeed number of directed walks of length k.

Thus, we have proved the theorem using induction. ■

Corollary 4 For every k > 0, the trace $tr(A^k)$ equals the number of loops of length k in G.

Proof. Since $(A^k)_{ii}$ equals the number of directed walks of length k from i to i, (i.e., a loop), we could sum over the n entries in the diagonal to obtain all the loops of length k in G. This is exactly $tr(A^k)$.

Now assume that G is undirected.

This turns A to be a symmetric real matrix. Using the spectral theorem, it follows that A admits an orthonormal eigenbasis $b_0(G), b_1(G), \ldots, b_{n-1}(G)$ with real eigenvalues $\lambda_0(G) \geq \lambda_1(G) \geq \ldots \geq \lambda_{n-1}(G)$. That is, we have

$$Ab_i = \lambda_i b_i \ (0 \le i < n).$$

Note that the λ_i are well defined, but the b_i are not. Also:

Lemma 5 The eigenvalues are graph invariants, that is, isomorphic graphs have the same eigenvalues.

Proof. Isomorphic graphs have the same eigenvalues. Since Isomorphic graphs are structurally the same, they have the exact same adjacency matrix, thus the same eigenvalues.

Exercise 6 *Compute* λ_i *and* b_i *for the triangle.*

One way to visualize the adjacency matrix as an operator is as follows. Write real numbers on the vertices of G, call this function f. Now A acts by taking all neighbors of the vertex x, add up the f-values there and write it to the position x. This will be the value of Af at x:

$$(Af)(x) = \sum_{(x,y)\in E(G)} f(y).$$

Actually this is how we will *define* the adjacency operator for infinite graphs. Using this image, one can prove.

Theorem 7 Let G be an undirected graph with maximal degree d. Then $|\lambda_i(G)| \le d$ $(0 \le i < n)$. When G is d-regular, we have $\lambda_0(G) = d$.

Hint: take an eigenvector. Find a particular vertex for it..

Proof. Let G be undirected graph with max degree d.

First we show $|\lambda_k| := |\lambda_k(G)| \le d$ for $k \in \{0, \dots, n-1\}$. Let λ_k be an eigenvalue of adjacency matrix A_G , and let

$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \tag{2}$$

be a corresponding eigenvector. Choose x_i the max component $\{x_1, \ldots, x_n\}$. WLOG $x_i > 0$. We have

$$|\lambda_k x_i| = A \cdot x \tag{3}$$

$$= \left| \sum_{j=1}^{n} a_{ij} x_j \right| \tag{4}$$

$$\leq \left| \sum_{j=1}^{n} a_{ij} x_i \right|$$
(5)

$$\leq |dx_i|$$
 (6)

$$=|x_i|d\tag{7}$$

since summing over row gives the degree, which is less than the degree bound. Thus, we have $|\lambda_k||x_i| \leq |x_i|d$ so $|\lambda_k| \leq d$ as desired.

Now, when G is d-regular $\implies \lambda_0(G) = d$.

We need an eigenvector that has d as its eigenvalue.

Consider the vector

$$x = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} \tag{8}$$

with n copies of 1. Since G is d-regular, every row of A_G sum to d. Thus,

$$A_G \cdot x = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} d \\ \vdots \\ d \end{bmatrix}$$
 (9)

Thus, d is an eigenvalue. Since all eigenvalues are $\leq d$, d is $\lambda_0(G)$, the biggest.

Exercise 8 Assume that G is undirected and connected with maximal degree d. Then $\lambda_0(G) = d$ if and only if G is d-regular.

Proof. G is d regular $\implies \lambda_0(G) = d$. We already proved this in the previous theorem. Let G undirected, connected, max degree is d. $\lambda_0(G) = d \implies G$ is d-regular. Let

$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \tag{10}$$

be an eigenvector for eigenvalue d. Again, WLOG, pick a max component $x_i >$ among the entries x_i, \ldots, x_n . With $A_G = [a_{ij}]$ we have

$$|dx_i| = |\sum_{j=1}^n a_{ij} x_j|$$
 (11)

$$\leq |\sum_{j=1}^{n} a_{ij} x_i| \tag{12}$$

$$\leq |dx_i| \tag{13}$$

meaning we have equal signs above. so x is a vector of constants of value x_i . Thus, we could rescale to have

$$x = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} \tag{14}$$

so $\sum_{j=1}^{n} a_{ij} = d$, thus, vertex i, corresponding the row

$$\begin{bmatrix} a_{i1} & \dots & a_{in} \end{bmatrix} \tag{15}$$

has degree d, and since G connected, we could apply

$$x = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} \tag{16}$$

to other row i to obtain v_i has degree d. Thus, G is d-regular.

Lemma 9 Let G be a d-regular undirected graph. Then the multiplicity of d as an eigenvalue of G equals the number of connected components of G.

The eigenvalue -d also comes into the picture naturally.

Proof. Let G be d-regular, undirected, graph that has n vertices. \Longrightarrow multiplicity of d as an eigenvalue of G = # of connected components of G. Let G have i connected components C_i with $i \in \{1, \ldots, k\}$ for some k. We define vectors X^i corresponding to C_i in this way:

$$X_j^i = \begin{cases} 1 \text{ if } v_j \in C_i \\ 0 \text{ else} \end{cases}$$
 (17)

$$j \in \{1, \dots, n\} \tag{18}$$

The adjacency matrix of G, A_G looks like this:

$$A_{G} \cdot X_{2} = \begin{bmatrix} C_{1} & 0 & \dots & 0 \\ 0 & C_{2} & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & & C_{n} \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix} = d \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}$$
(19)

since d-regular row sum to d. We could see that X^i are eigenvectors of eigenvalue d. And x^i 's are orthogonal to each other. Thus, the multiplicity of d is at least the number of connected components of G.

We want to show that those X^i actually span the eigenbasis of d, so there are no more eigenvectors of d that we didn't already take into account.

So let y be an eigenvector of d, we want to express it as a linear combination of X^i 's.

$$i\text{-th row of } A_G \cdot \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_n \end{bmatrix} = i\text{-th row of } d \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_n \end{bmatrix}$$
 (20)

We assume $|y_i|$ is the max component in vector y, and $y_i \in C_i$.

$$\sum_{(i,j)\in \text{Edge Set}} y_j = dy_i \tag{21}$$

LHS has d summands, since G is d-regular. Also,

$$|y_j| \le |y_i| \forall j \tag{22}$$

so in fact all y_j are y_i . So those underlying vertices v_j are all connected to the v_i , so they are in component C_i . Thus, for vector y, the "block" corresponding to C_i could be expressed as a linear combination of X^i , like αX^i .

Lemma 10 Let G be a d-regular undirected, connected graph. Then $\lambda_{n-1} = -d$ if and only if G is bipartite.

Proof. We know from Ex 8 that since G is d-regular, undirected, connected, d is an eigenvalue. Want To Show: G bipartite $\implies -d$ is also eigenvalue. (Then $\lambda_{n-1} = -d$ since $|\lambda_i(G)| \leq d$ eigenvalue cannot get smaller than -d.)

Since G is bipartite, A_G could be expressed as:

$$\begin{bmatrix} 0 & M \\ M^T & 0 \end{bmatrix} \tag{23}$$

for some matrix M that is $k \times (n-k)$. Now let

$$\begin{bmatrix} x_1 \\ \vdots \\ x_k \\ x_{k+1} \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} X \\ Y \end{bmatrix}$$

$$(24)$$

be an eigenvector of eigenvalue d.

Therefore, we get

$$d \begin{bmatrix} X \\ Y \end{bmatrix} = \begin{bmatrix} 0 & M \\ M^T & 0 \end{bmatrix} \cdot \begin{bmatrix} X \\ Y \end{bmatrix} = \begin{bmatrix} MY \\ M^T X \end{bmatrix}$$
 (25)

Thus:

$$dX = MY (26)$$

$$dY = M^T X (27)$$

Claim: -d is eigenvalue with eigenvector

$$\begin{bmatrix} X \\ -Y \end{bmatrix} = \begin{bmatrix} x_1 \\ \vdots \\ x_k \\ -x_{k+1} \\ \vdots \\ -x_n \end{bmatrix}$$
 (28)

so we have

$$\begin{bmatrix} 0 & M \\ M^{T} & 0 \end{bmatrix} \begin{bmatrix} X \\ -Y \end{bmatrix} = \begin{bmatrix} M(-Y) \\ M^{T}X \end{bmatrix} = \begin{bmatrix} -MY \\ M^{T}X \end{bmatrix} = \begin{bmatrix} -dX \\ dY \end{bmatrix} = \begin{bmatrix} -dX \\ -d(-Y) \end{bmatrix} = -d \begin{bmatrix} X \\ -Y \end{bmatrix} = -d \begin{bmatrix} x_{1} \\ \vdots \\ x_{k} \\ -x_{k+1} \\ \vdots \\ -x_{n} \end{bmatrix}$$
(29)

Thus, -d is eigenvalue as wanted.

 $\lambda_{n-1} = -d$, -d is also eigenvalue $\implies G$ bipartite.

Exercise 11 Let G be a d-regular undirected graph. What is the multiplicity of -d as an eigenvalue of G?

Now we look at some simple examples.

Exercise 12 Compute the eigenvalues and eigenvectors for the cycle of length n.

Hint: What are the eigenvalues for the directed cycle of length n?

Exercise 13 Compute the eigenvalues and an orthonormal eigenbasis for the complete graph on d+1 points.

Hint: What happens to the eigenvalues and eigenvectors of A if you add a scalar matrix to A?