

Adjacency matrices, interlacing and the Perron-Frobenius theorem

We recall

Definition. The adjacency matrix M_G of a graph is given by

$$(M_G \vec{x})(v_1) = \sum_{\substack{v_2 \\ v_1 v_2}} \vec{x}(v_2)$$

The notation deserves a moment of explanation. M_G is a $v \times v$ matrix, so it operates on vectors $\vec{x} \in \mathbb{R}^v$, returning vectors $(M_G \vec{x}) \in \mathbb{R}^v$.

If v_1 is some vertex in G , and \vec{y} is any vector in \mathbb{R}^v , then

" $\vec{y}(v_1)$ " is the component of \vec{y}

(2)

in whatever position corresponds to vertex v_1 .

The sum $\sum_{v_1 \rightarrow v_2}$ is the sum over all v_2 which are connected to v_1 by an edge.

Definition. The eigenvalues of M_G are denoted $\mu_1 \geq \dots \geq \mu_n$.

This is different from our convention that the eigenvalues of L_G are denoted $\lambda_1 \leq \dots \leq \lambda_n$.

Let's see why this is a good idea.

(3)

Definition. A graph G is d -regular if every vertex has degree d .

Proposition. If G is a d -regular graph then $\lambda_i = d - \mu_i$.

Proof. We know $L_G = D_G - M_G$. If G is d -regular then $D_G = dI$. So the i -th eigenvector $\vec{\Psi}_i$ of L_G has

$$M_G \vec{\Psi}_i = (D_G - L_G) \vec{\Psi}_i = (dI - L_G) \vec{\Psi}_i = (d - \lambda_i) \vec{\Psi}_i$$

and is an eigenvector of M_G with eigenvalue $d - \lambda_i$. Since this is true for all $\vec{\Psi}_1, \dots, \vec{\Psi}_r$, these must be all of the r eigenvalues of M_G . Further, since $\lambda_1 \leq \dots \leq \lambda_n$, we have $d - \lambda_1 \geq \dots \geq d - \lambda_n$ and so $\mu_i = d - \lambda_i$ as claimed. \square

(4)

There's a suggestive corollary of this result:

Cor. If G is d -regular, then $\mu_1 = d$ and the corresponding eigenvector is constant.

Proof. We know $\lambda_1 = 0$, with $\vec{\psi}_1$ constant...

So what if G is not d -regular? Well,

Lemma. If d_{ave} is the average degree of a vertex in G and d_{max} is the maximum degree of a vertex in G ,

$$d_{ave} \leq \mu_1 \leq d_{max}.$$

Proof. We know from Courant-Fischer¹ (5)

$$\begin{aligned}\mu_1 &= \max_{\vec{x}} \frac{\langle \vec{x}, M_G \vec{x} \rangle}{\langle \vec{x}, \vec{x} \rangle} \\ &\geq \max_{\vec{1}} \frac{\langle \vec{1}, M_G \vec{1} \rangle}{\langle \vec{1}, \vec{1} \rangle} \\ &= \frac{\langle \vec{1}, \begin{bmatrix} d_1 \\ \vdots \\ d_r \end{bmatrix} \rangle}{\sqrt{\quad}} = \frac{\sum_i d_r}{\sqrt{\quad}} = \text{dave.}\end{aligned}$$

The interesting step here is

$$(M_G \vec{1})(v_1) = \sum_{v_1 \leftrightarrow v_2} \vec{1}(v_2) = \sum_{v_1 \leftrightarrow v_2} 1 = \underset{\substack{\uparrow \\ \text{degree}}}{d(v_1)}.$$

1. This is a standard trick for extracting information about eigenvalues - use the Rayleigh quotient with a cleverly chosen vector. We'll see this a lot.

Now suppose $\vec{\Phi}_1$ is the eigenvector of M_G with eigenvalue μ_1 . w.l.o.g. we can assume that some entry of $\vec{\Phi}_1$ is positive. Suppose v_{max} has $\vec{\Phi}_1(v_{max}) \geq \vec{\Phi}_1(v)$ for all v . Now

$$\mu_1 = \frac{(M_G \vec{\Phi}_1)(v_{max})}{\vec{\Phi}_1(v_{max})} = \frac{\sum_{v_{max} \rightarrow v_2} \vec{\Phi}_1(v_2)}{\vec{\Phi}_1(v_{max})}$$

$$= \sum_{v_{max} \rightarrow v_2} \frac{\vec{\Phi}_1(v_2)}{\vec{\Phi}_1(v_{max})}$$

$$\leq \sum_{v_{max} \rightarrow v_2} 1 = d(\vec{v}_{max}) \leq d_{max}. \quad \square$$

We note that this might be a little stronger, if $d(\vec{v}_{max}) \neq d_{max}$.
than claimed

If G is d -regular, then $d_{\text{ave}} = d_{\text{max}} = d$, ⑦

so this implies our previous lemma.

In fact, we can get a converse result!

Lemma. If G is connected and $\mu_1 = d_{\text{max}}$, then G is d_{max} -regular.

Proof. If $\mu_1 = d_{\text{max}}$ then (by our previous argument) we have

$$\sum_{v_{\text{max}} \rightarrow v_2} \frac{\vec{\Phi}_1(v_2)}{\Phi_1(v_{\text{max}})} = \sum_{v_{\text{max}} \rightarrow v_2} 1 = d(v_{\text{max}}) = d_{\text{max}}.$$

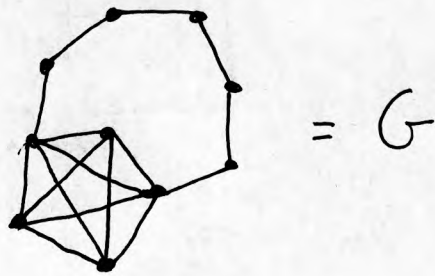
Therefore $\vec{\Phi}_1(v_2) = \vec{\Phi}_1(v_{\text{max}})$ for all v_2 joined to v_{max} by an edge and $d(v_{\text{max}}) = d_{\text{max}}$.

8

But then all these v_2 are also maxes for φ_1 and we can repeat the argument to show their degrees are also d_{\max} .

Since G is connected, this eventually covers the entire graph. \square

Can we do even better? For instance, consider



which has a high average degree some places, but a low average degree in others.

Cauchy's Interlacing Theorem.

Let A be an $n \times n$ symmetric matrix and B be the $(n-1) \times (n-1)$ submatrix obtained by deleting the same row and column.¹

Let $\alpha_1 \geq \dots \geq \alpha_n$ and $\beta_1 \geq \dots \geq \beta_{n-1}$ be the eigenvalues of A and B . Then

$$\alpha_1 \geq \beta_1 \geq \alpha_2 \geq \beta_2 \geq \dots \geq \beta_{n-1} \geq \alpha_n.$$

Proof. There's a proof in the book, which is basically to apply Courant-Fischer. \square

1. This is called a principal submatrix, as you might remember.

Definition. If S is a subset of the vertices of a graph G then $G(S)$, called "the subgraph induced by S " is the graph with vertex set S and edge set $\{v_1 \rightarrow v_2 \mid v_1, v_2 \in S \text{ and } v_1 \rightarrow v_2 \text{ an edge of } G\}$.

Definition. If M is an $n \times n$ symmetric matrix and $S \subset \{1, \dots, n\}$ then $M(S)$ is the symmetric submatrix with rows and columns in S .

Lemma. For every $S \subset V(G)$, let $d_{\text{ave}}(S)$ be the average ^{vertex} degree of $G(S)$. Then

$$d_{\text{ave}}(S) \leq \mu_1$$

(where μ_1 is $\mu_1(G)$).

Proof. If M_G is the adjacency matrix of G , then $M(S)$ is the adjacency matrix of $G(S)$.

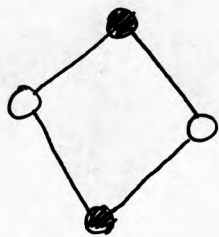
Now if $(\mu_S)_1$ is the largest eigenvalue of $G(S)$, by Cauchy interlacing,

$$(\mu_S)_1 \leq \mu_1$$

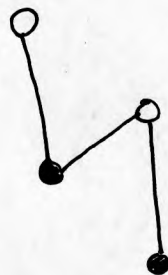
But we already know that

$$d_{\text{ave}}(G(S)) \leq (\mu_S)_1 \leq \mu_1. \quad \square$$

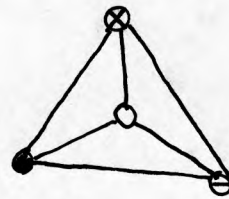
Definition. A coloring of a graph is an assignment of colors to vertices so that every pair of vertices joined by an edge have different colors.



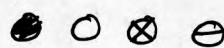
2 colors



2 colors



4 colors



Definition. A graph is K-colorable if it can be colored with only K colors.

The chromatic number $\chi(G)$ of a graph is the smallest K so that G is K-colorable.

Lemma. $\chi(G) \leq d_{\max} + 1$.

Proof. Assign colors to vertices one by one, and proceed by induction, with the hypothesis "no two adjacent vertices have the same color".

Base case. 0 ~~vertices~~ vertices colored. True.

Inductive step. Choose an uncolored vertex. It has at most d_{\max} neighbors with at most d_{\max} different colors. ~~Use~~ At least one color is left unused. Color the ~~new~~ vertex with an unused color. \square

Wilf's Theorem. $\chi(G) \leq \lfloor \mu_1 \rfloor + 1$.

Proof. We will induct on the number of vertices in the graph.

Base case. The one vertex graph with no edges has $\chi(G) = 1$ and $\mu_1 = \lambda_1 + 0 = 0$, so $\chi(G) \leq \lfloor \mu_1 \rfloor + 1 = 1$.

Inductive step. Suppose the theorem is true for all graphs with $v-1$ vertices and let G be a graph with v vertices.

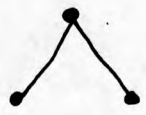
we have $d_{ave} \leq \lfloor \mu_1 \rfloor$ and since $d_{ave} \leq \mu_1 \leq d_{max}$, there is a vertex v_i of G with degree $\leq \lfloor \mu_1 \rfloor$.

Let $S = V(G) - v$. By Cauchy interlacing, ^{vertices of G}

$$\mu_1(G(S)) \leq \mu_1(G)$$

so $\lfloor \mu_1(G(S)) \rfloor \leq \lfloor \mu_1(G) \rfloor$. Further, by induction, $G(S)$ has a coloring with at most $\lfloor \mu_1(G(S)) \rfloor + 1 \leq \lfloor \mu_1(G) \rfloor + 1$ colors.

Now v has degree at most $\lfloor \mu_1(G) \rfloor$ so no more than $\lfloor \mu_1(G) \rfloor$ of the colors are used by neighbors of v , and at least one is left to color v . \square

Example.  has $\mu_1 = \sqrt{2}$, but $d_{max} = 2$.

Thus Wilf's theorem tells us that

$$\chi(G) \leq \lfloor \sqrt{2} \rfloor + 1 = 2, \text{ while } d_{max} + 1 = 3.$$

Perron-Frobenius Theorem.

Let G be a connected (weighted) graph,
 let M_G be its adjacency matrix, and
 let $\mu_1 \geq \dots \geq \mu_r$ be the eigenvalues
 of M_G . Then

a. μ_1 has a strictly positive eigenvector

b. $\mu_1 \geq -\mu_n$

c. $\mu_1 > \mu_2$

In order to prove this, we'll need

Lemma. Let G be a connected graph,
 let M_G be its adjacency matrix and
 let some non-negative ^{nonzero} $\vec{\phi}$ be an eigenvector
 of M_G . Then $\vec{\phi}$ is strictly positive.

Proof. If the set $\{a \in V(G) \mid \vec{\Phi}(a) = 0\}$ is nonempty, there is some edge $b \rightarrow c$ so that $\vec{\Phi}(b) = 0$ and $\vec{\Phi}(c) > 0$.

After all, we know $\{c \in V(G) \mid \vec{\Phi}(c) > 0\}$ is nonempty (b/c $\vec{\Phi}$ is nonzero) and G is connected.

Now

$$\nu \vec{\Phi}(b) = (M\vec{\Phi})(b) = \sum_{b \rightarrow v} \vec{\Phi}(v) \geq \vec{\Phi}(c) > 0$$

The contradiction proves $\{a \in V(G) \mid \vec{\Phi}(a) = 0\}$ must have been empty. \square

Proof. (of Perron-Frobenius)

Let $\vec{\Phi}_1$ be an eigenvector of M_G with eigenvalue μ_1 , and assume $\|\vec{\Phi}_1\| = 1$. Let $\vec{x} = |\Phi_1|$.

We claim \vec{x} is an eigenvector of M_G with eigenvalue μ_1 .

Now $\langle \vec{x}, \vec{x} \rangle = \langle \vec{\Phi}_1, \vec{\Phi}_1 \rangle = 1$. Further,

$$\mu_1 = \langle \vec{\Phi}_1, M_G \vec{\Phi}_1 \rangle = \sum_{\substack{a,b \\ \text{in } V(G)}} M_G(a,b) \vec{\Phi}_1(a) \vec{\Phi}_1(b)$$

$$\leq \sum_{\substack{a,b \text{ in} \\ V(G)}} M_G(a,b) |\vec{\Phi}_1(a)| |\vec{\Phi}_1(b)|$$

$$= \langle \vec{x}, M_G \vec{x} \rangle$$

Since μ_1 is the maximum possible

Rayleigh quotient $\langle \vec{v}, M_G \vec{v} \rangle$ for any unit vector \vec{v} , the Courant-Fischer theorem says that $\langle \vec{x}, M_G \vec{x} \rangle = \mu_1$ and that \vec{x} is an eigenvector of M_G with eigenvalue μ_1 .

Since \vec{x} is non-negative, our lemma tells us \vec{x} is positive, proving a.

We now claim $\mu_1 \geq -\mu_n$.

Let $\vec{\Phi}_n$ be the eigenvector of M_G with eigenvalue μ_n , and let $\vec{y} = |\vec{\Phi}_n|_V$. Now
and $\|\vec{y}\| = \|\vec{\Phi}_n\| = 1$

$$\begin{aligned} |\mu_n| &= |\langle \vec{\Phi}_n, M_G \vec{\Phi}_n \rangle| \\ &= \left| \sum_{\substack{a,b \\ \in V(G)}} M_G(a,b) \vec{\Phi}_n(a) \vec{\Phi}_n(b) \right| \end{aligned}$$

$$\leq \sum_{\substack{a, b \\ \in V(G)}} \underbrace{M_G(a, b)}_{\rightarrow \text{already } \geq 0 \text{ by defn.}} |\vec{\Phi}_n(a)| |\vec{\Phi}_n(b)|$$

$$= \sum_{\substack{a, b \\ \in V(G)}} M_G(a, b) \vec{y}(a) \vec{y}(b)$$

$$= \langle \vec{y}, M \vec{y} \rangle \leq \mu_1.$$

Rayleigh quotient of \vec{y}
since $\|\vec{y}\| = 1$

which completes the proof of b.

We now claim $\mu_2 < \mu_1$. Suppose that $\vec{\Phi}_2$ is an eigenvector of M_G of eigenvalue μ_2 , which is orthogonal to $\vec{\Phi}_1$ and has $\|\vec{\Phi}_2\| = 1$. As before, let $\vec{y} = |\vec{\Phi}_2|$, and note $\|\vec{y}\| = \|\vec{\Phi}_2\|$.

We have

$$\mu_2 = \langle \vec{\Phi}_2, M_G \vec{\Phi}_2 \rangle = \sum_{\substack{a,b \\ \in V(G)}} M_G(a,b) \vec{\Phi}_2(a) \vec{\Phi}_2(b)$$

$$\leq \sum_{\substack{a,b \\ \in V(G)}} M_G(a,b) |\vec{\Phi}_2(a)| |\vec{\Phi}_2(b)|$$

$$= \sum_{\substack{a,b \\ \in V(G)}} M_G(a,b) \vec{y}(a) \vec{y}(b)$$

$$= \langle \vec{y}, M_G \vec{y} \rangle \leq \mu_1.$$

Now if $\mu_2 = \mu_1$, then (by Courant-Fischer)

\vec{y} is a non-negative eigenvector of M_G of eigenvalue μ_1 , and hence \vec{y} is strictly positive by our Lemma. This means that

$\vec{\Phi}_2$ has no zero entries.

Since $\vec{\Phi}_2$ is orthogonal to $\vec{\Phi}_1$ (which has all positive entries), we know $\vec{\Phi}_2$ has both positive and negative entries. Since G ~~was~~ is connected, there must be some edge $a \rightarrow b$ with $\Phi_2(a) < 0 < \Phi_2(b)$.

But then

$$M_G(a,b) \Phi_2(a) \Phi_2(b) = \Phi_2(a) \Phi_2(b)$$

$$< 0 < |\Phi_2(a)| |\Phi_2(b)| = y(a) y(b)$$

$$= M_G(a,b) y(a) y(b)$$

and $\mu_2 < \mu_1$. The contradiction proves that $\mu_2 \neq \mu_1$ (and hence that $\mu_2 < \mu_1$), proving c. \square

Now that we have the Perron-Frobenius theorem, we can give a purely spectral characterization of ~~the~~ graphs with $\chi(G) = 2$ (also called bipartite graphs).

Prop. If G is connected and $\mu_n = -\mu_1$, then G is bipartite.

Proof. In the proof of P-F, we let $\vec{\phi}_n$ be the ~~eigenvector~~ eigenvector of M_G with eigenvalue μ_n , and considered

$\vec{y} = |\vec{\phi}_n|$, with $\|\vec{\phi}_n\| = \|\vec{y}\| = 1$. We saw

$$|\mu_n| = |\langle \vec{\phi}_n, M_G \vec{\phi}_n \rangle|$$

$$= \left| \sum_{a \rightarrow b} \vec{\phi}_n(a) \vec{\phi}_n(b) \right| \leq \sum_{a \rightarrow b} |\vec{\phi}_n(a)| |\vec{\phi}_n(b)|$$

$$\leq \sum_{a \rightarrow b} \vec{y}(a) \vec{y}(b) = \langle \vec{y}, M_G \vec{y} \rangle \leq \mu_1$$

If all these inequalities are equalities, then all $\vec{\Phi}_n(a)\vec{\Phi}_n(b)$ must have the same sign.

Further, \vec{y} must have Rayleigh quotient μ_1 and hence be a non-negative (\Rightarrow strictly positive) eigenvector of M_G of eigenvalue μ_1 .

Since \vec{y} is strictly positive, $\vec{\Phi}_n$ has no zero values and (since $\vec{\Phi}_n$ is orthogonal to $\vec{\Phi}_1$) must have positive and negative values. As before, some edge $a \leftrightarrow b$ must join the positive vertices to the negative ones, and for this edge $\vec{\Phi}_n(a)\vec{\Phi}_n(b) < 0$.

But this means $\vec{\Phi}_n(a) \vec{\Phi}_n(b) < 0$ for every edge $a \rightarrow b$. We conclude that

$$\{ a \mid \vec{\Phi}_n(a) > 0 \} \text{ and } \{ a \mid \vec{\Phi}_n(a) < 0 \}$$

are a 2-coloring of G . \square

We even have a (stronger) converse!

Prop. If G is bipartite, ~~\mathbb{R}~~ then the eigenvalues $\mu_1 \geq \dots \geq \mu_n$ are symmetric around zero.

Proof. The definition of bipartite (or 2-colorable) is that $V(G)$ can be partitioned into S and T so that every edge joins a vertex in S to a vertex in T .

Now suppose $\vec{\Phi}$ is an eigenvector of M_G with eigenvalue μ . Define \vec{X} by

$$\vec{X}(a) = \begin{cases} \vec{\Phi}(a), & \text{if } a \in S \\ -\vec{\Phi}(a), & \text{if } a \in T \end{cases}$$

Now if $a \in S$,

$$\begin{aligned} (M_G \vec{X})(a) &= \sum_{a \rightarrow b} M_G(a,b) \vec{X}(b) \\ &= \sum_{a \rightarrow b} M_G(a,b) (-\vec{\Phi}(b)) \quad (\text{since } b \in T) \\ &= - \sum_{a \rightarrow b} M_G(a,b) \vec{\Phi}(b) \\ &= -(M_G \vec{\Phi})(a) \\ &= -\mu \vec{\Phi}(a) = -\mu \vec{X}(a). \end{aligned}$$

Similarly, if $a \in T$, $(M_G \vec{X})(a) = -\mu \vec{X}(a)$. Thus \vec{X} is an eigenvector of M_G with eigenvalue $-\mu$, completing the proof! \square