

Chapter 2

Spectra of Finite Graphs

2.1 Characteristic Polynomials

Let $G = (V, E)$ be a finite graph on $n = |V|$ vertices. Numbering the vertices, we write down its adjacency matrix in an explicit form of $n \times n$ matrix, say A . The characteristic polynomial of A is defined as usual by

$$\varphi_A(x) = |xI - A| = \det(xI - A),$$

where I is the $n \times n$ identity matrix. It is noted that $\varphi_A(x)$ is determined independently of the numbering. In fact, let A' be the adjacency matrix obtained by a different numbering. It follows from Proposition 1.20 that $A' = S^{-1}AS$ with an $n \times n$ permutation matrix S . Then,

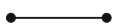
$$\begin{aligned}\varphi_{A'}(x) &= |xI - A'| = |xI - S^{-1}AS| \\ &= |S^{-1}(xI - A)S| = |S|^{-1}|xI - A||S| = \varphi_A(x).\end{aligned}$$

We call $\varphi_A(x)$ the *characteristic polynomial* of G and denote it by $\varphi(x; G)$.

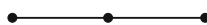
Example 2.1 Here are a few simple examples:



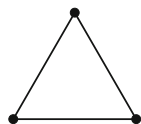
$$x$$



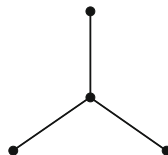
$$x^2 - 1$$



$$x^3 - 2x$$



$$x^3 - 3x - 2 = (x + 1)^2(x - 2)$$



$$x^4 - 3x^2 = x^2(x^2 - 3)$$

By definition the characteristic polynomial of a graph $G = (V, E)$ on $n = |V|$ vertices is of the form:

$$\varphi(x; G) = x^n + c_1 x^{n-1} + c_2 x^{n-2} + c_3 x^{n-3} + \cdots + c_{n-1} x + c_n. \quad (2.1)$$

The coefficients represent combinatorial characteristics, see e.g., Bapat [13, Sect. 3.2] and Biggs [17, Chap. 7]. We only mention the following

Theorem 2.2 *For the coefficients of the characteristic polynomial (2.1) we have:*

- (1) $c_1 = 0$.
- (2) $-c_2 = |E|$, the number of edges.
- (3) $-c_3 = 2\Delta$, where Δ is the number of triangles in G .

Proof Let $A = [a_{ij}]$ be the adjacency matrix of G in an explicit form of $n \times n$ matrix after numbering the vertices. Since the diagonal elements of A vanish, the characteristic polynomial of G is given by

$$\varphi(x; G) = |xI - A| = \begin{vmatrix} x & -a_{12} & \cdots & -a_{1n} \\ -a_{21} & x & \cdots & -a_{2n} \\ \vdots & & \ddots & \vdots \\ -a_{n1} & \cdots & \cdots & x \end{vmatrix}.$$

For simplicity, the matrix in the right-hand side is denoted by $B = [b_{ij}]$. Then,

$$\begin{aligned} \varphi(x; G) &= |B| = \sum_{\sigma \in \mathfrak{S}(n)} \operatorname{sgn}(\sigma) b_{1\sigma(1)} b_{2\sigma(2)} \cdots b_{n\sigma(n)} \\ &= \sum_{k=0}^n \sum_{\substack{\sigma \in \mathfrak{S}(n) \\ |\operatorname{supp} \sigma| = k}} \operatorname{sgn}(\sigma) b_{1\sigma(1)} b_{2\sigma(2)} \cdots b_{n\sigma(n)}, \end{aligned}$$

where $\operatorname{supp} \sigma = \{1 \leq i \leq n \mid \sigma(i) \neq i\}$. Since the indeterminate x appears only in the diagonal of B , comparing with (2.1) we have

$$f_k(x) \equiv \sum_{\substack{\sigma \in \mathfrak{S}(n) \\ |\operatorname{supp} \sigma| = k}} \operatorname{sgn}(\sigma) b_{1\sigma(1)} b_{2\sigma(2)} \cdots b_{n\sigma(n)} = c_k x^{n-k}. \quad (2.2)$$

That $c_1 = 0$ is apparent since there is no permutation σ with $|\operatorname{supp} \sigma| = 1$. For c_2 we note that the permutations σ with $|\operatorname{supp} \sigma| = 2$ are parametrized as $\sigma = (i j)$ ($1 \leq i < j \leq n$). Taking $\operatorname{sgn}(\sigma) = -1$ into account, we obtain

$$f_2(x) = \sum_{1 \leq i < j \leq n} (-1)(-a_{ij})(-a_{ji})x^{n-2} = - \sum_{1 \leq i < j \leq n} a_{ij} x^{n-2}$$

where we used $a_{ij}a_{ji} = a_{ij}^2 = a_{ij}$, and hence $c_2 = -|E|$. Similarly, for c_3 we need to consider permutations σ satisfying $|\text{supp } \sigma| = 3$. Those permutations are parametrized as $\sigma = (i j k)$ and $\sigma = (i k j)$ for $1 \leq i < j < k \leq n$. Noting that $\text{sgn}(\sigma) = 1$ for such cyclic permutations, we have

$$f_3(x) = - \sum_{1 \leq i < j < k \leq n} (a_{ij}a_{jk}a_{ki} + a_{ik}a_{kj}a_{ji})x^{n-3}.$$

Since $a_{ij}a_{jk}a_{ki} = a_{ik}a_{kj}a_{ji} = 1$ only if i, j, k form a triangle, we get $-c_3 = 2\Delta$. \square

Exercise 2.3 Let A be the adjacency matrix of a finite graph $G = (V, E)$. Show that

$$\text{Tr } A = 0, \quad \text{Tr } A^2 = 2|E|, \quad \text{Tr } A^3 = 6\Delta.$$

2.2 Spectra

Let $G = (V, E)$ be a finite graph on $n = |V|$ vertices and A the adjacency matrix. The characteristic polynomial of G is factorized as

$$\varphi(x; G) = |xI - A| = \prod_{j=1}^n (x - \lambda_j), \quad (2.3)$$

where $\lambda_1, \dots, \lambda_n$ are real numbers since A is a real symmetric matrix. Changing notation, let $\lambda_1 < \lambda_2 < \dots < \lambda_s$ be the distinct eigenvalues of A . Then (2.3) becomes

$$\varphi(x; G) = |xI - A| = \prod_{i=1}^s (x - \lambda_i)^{m_i}, \quad (2.4)$$

where $m_i \geq 1$ is called the (*algebraic*) *multiplicity*¹ of an eigenvalue λ_i .

Definition 2.4 Let $G = (V, E)$ be a finite graph and $\varphi(x; G)$ its characteristic polynomial in the form (2.4). The *spectrum* of G is by definition the array

$$\text{Spec}(G) = \begin{pmatrix} \lambda_1 & \lambda_2 & \dots & \lambda_s \\ m_1 & m_2 & \dots & m_s \end{pmatrix}. \quad (2.5)$$

Each λ_i is called an *eigenvalue* of G and m_i its *multiplicity*.

Proposition 2.5 For finite graphs G and G' , $G \cong G'$ implies $\text{Spec}(G) = \text{Spec}(G')$.

¹Let $\varphi_A(x) = |xI - A|$ be the characteristic polynomial of a matrix $A \in M(n, \mathbb{C})$. Then $\lambda \in \mathbb{C}$ is an eigenvalue of A if and only if $\varphi_A(\lambda) = 0$. In that case the *algebraic multiplicity* of the eigenvalue λ is defined to be the multiplicity of λ as a zero of the polynomial $\varphi_A(x)$. While, the *geometric multiplicity* is defined to be the dimension of the eigenspace associated to λ .

Proof Straightforward from Proposition 1.21. \square

Example 2.6 The spectra of the five graphs in Example 2.1 are given as follows:

$$\begin{pmatrix} 0 \\ 1 \end{pmatrix}, \quad \begin{pmatrix} -1 & 1 \\ 1 & 1 \end{pmatrix}, \quad \begin{pmatrix} -\sqrt{2} & 0 & \sqrt{2} \\ 1 & 1 & 1 \end{pmatrix}, \quad \begin{pmatrix} -1 & 2 \\ 2 & 1 \end{pmatrix}, \quad \begin{pmatrix} -\sqrt{3} & 0 & \sqrt{3} \\ 1 & 2 & 1 \end{pmatrix}.$$

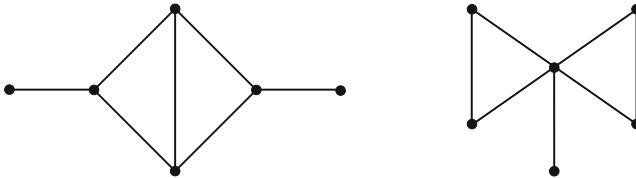
The converse assertion of Proposition 2.5 is not valid, namely, there are non-isomorphic graphs that have the same spectra. The following two graphs have a common characteristic polynomial: $\varphi(x) = x^5 - 4x^3 = x^3(x - 2)(x + 2)$.



In general, graphs with common spectra are called *cospectral* graphs. It is known that the above is a unique pair of cospectral graphs among 34 non-isomorphic graphs on 5 vertices. For a concise account on small graphs see e.g., Godsil–McKay [54], Harary *et al.* [62], Johnson–Newman [82], and for more recent and detailed results see Cvetković *et al.* [39–41]. There are many attempts to distinguish cospectral graphs by means of another matrices.

Exercise 2.7 (Baker [10]) Show that the following two graphs have a common characteristic polynomial:

$$\begin{aligned} \varphi(x) &= x^6 - 7x^4 - 4x^3 + 7x^2 + 4x - 1 \\ &= (x - 1)(x + 1)^2(x^3 - x^2 - 5x + 1). \end{aligned}$$

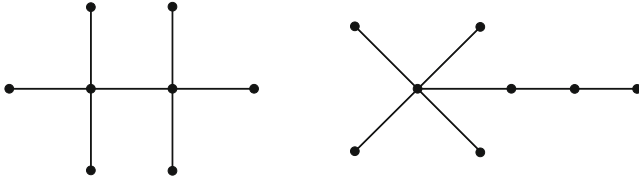


In fact, this is a unique pair of cospectral connected graphs among 112 non-isomorphic connected graphs on 6 vertices.

Exercise 2.8 (Collatz–Sinogowitz [37])

Show that the following two trees have a common characteristic polynomial:

$$\varphi(x) = x^8 - 7x^6 + 9x^4 = x^4(x^4 - 7x^2 + 9).$$



Exercise 2.9 Examine by characteristic polynomials that graphs $G = (V, E)$ with $|V| \leq 4$ are uniquely specified by their spectra.

Proposition 2.10 For a finite graph G let $s(G)$ denote the number of distinct eigenvalues of G . Then we have $s(G) = \dim \mathcal{A}(G)$.

Proof Let $\lambda_1 < \dots < \lambda_s$ be the different eigenvalues of the adjacency matrix A , $s = s(G)$. After diagonalization by a suitable orthogonal matrix U we obtain

$$U^{-1}AU = \text{diag}[\underbrace{\lambda_1 \cdots \lambda_1}_{m_1} \cdots \underbrace{\lambda_s \cdots \lambda_s}_{m_s}] \equiv D.$$

Then $\{I, D, D^2, \dots, D^{s-1}\}$ is linearly independent, while $\{I, D, D^2, \dots, D^{s-1}, D^s\}$ is not since $(D - \lambda_1 I) \cdots (D - \lambda_s I) = O$, where I is the identity matrix. Therefore the algebra $U^{-1}\mathcal{A}(G)U$ is of dimension s , and so is $\mathcal{A}(G)$. \square

Corollary 2.11 For a connected finite graph G we have

$$s(G) = \dim \mathcal{A}(G) \geq \text{diam}(G) + 1.$$

Proof Immediate from Propositions 2.10 and 1.26. \square

2.3 Spectra of K_n , C_n and P_n

Lemma 2.12 Let J be the $n \times n$ matrix with entries being all one. Then J/n is a projection of rank one. Therefore the eigenvalues of J are n with multiplicity one and 0 with multiplicity $n - 1$.

Proof That J/n is a projection follows from $J^2 = nJ$ and $J^* = J$. Let $\{e_k; 1 \leq k \leq n\}$ be the canonical basis of \mathbb{C}^n and set $f = \sum_{k=1}^n e_k$. Then we have $Jf = nf$ and $J(e_1 - e_k) = 0$ for $2 \leq k \leq n$. Since $\{f, e_1 - e_k; 2 \leq k \leq n\}$ is a basis of \mathbb{C}^n , the eigenvalues of J are n with multiplicity one and 0 with multiplicity $n - 1$. In particular, J/n is a projection onto the one-dimensional subspace spanned by f . \square

Theorem 2.13 For the complete graph K_n with $n \geq 1$ we have

$$\varphi(x; K_n) = (x + 1)^{n-1}(x - n + 1), \quad \text{Spec}(K_n) = \begin{pmatrix} -1 & n-1 \\ n-1 & 1 \end{pmatrix}.$$

Proof Since the adjacency matrix A of K_n is given by $A = J - I$, the assertion is immediate from Lemma 2.12. \square

Lemma 2.14 *For $n \geq 2$ let W be the $n \times n$ circulant permutation matrix defined by*

$$W = \begin{bmatrix} 0 & & & 1 \\ 1 & 0 & & \\ & 1 & 0 & \\ & & \ddots & \ddots \\ & & & 1 & 0 \end{bmatrix}.$$

The eigenvalues of W are $1 = \zeta^0, \zeta, \dots, \zeta^{n-1}$, where $\zeta = \exp(2\pi i/n)$.

Proof Let $\{e_k; 1 \leq k \leq n\}$ be the canonical basis of \mathbb{C}^n and set

$$f_l = \sum_{k=1}^n \zeta^{-lk} e_k, \quad 0 \leq l \leq n-1.$$

Since $W e_k = e_{k+1}$ for $1 \leq k \leq n$ with the understanding that $e_{n+1} = e_1$, we have

$$W f_l = \sum_{k=1}^n \zeta^{-lk} W e_k = \sum_{k=1}^n \zeta^{-lk} e_{k+1} = \zeta^l \sum_{k=1}^n \zeta^{-l(k+1)} e_{k+1} = \zeta^l f_l.$$

Namely, ζ^l is an eigenvalue of W with an eigenvector f_l . Since $\{f_l; 0 \leq l \leq n-1\}$ forms a basis of \mathbb{C}^n , $\{\zeta^l; 0 \leq l \leq n-1\}$ exhaust the eigenvalues of W . \square

Theorem 2.15 *The spectrum of the cycle C_{2m+1} for $m \geq 1$ is given by*

$$\text{Spec}(C_{2m+1}) = \left(\begin{array}{cccc} 2 & \cdots & 2 \cos \frac{2l\pi}{2m+1} & \cdots & 2 \cos \frac{2m\pi}{2m+1} \\ 1 & \cdots & 2 & \cdots & 2 \end{array} \right), \quad 1 \leq l \leq m. \quad (2.6)$$

The spectrum of the cycle C_{2m} for $m \geq 2$ is given by

$$\text{Spec}(C_{2m}) = \left(\begin{array}{cccc} 2 & \cdots & 2 \cos \frac{2l\pi}{2m} & \cdots & -2 \\ 1 & \cdots & 2 & \cdots & 1 \end{array} \right), \quad 1 \leq l \leq m-1. \quad (2.7)$$

Proof We keep the same notations as in Lemma 2.14. Note first that the adjacency matrix of C_n is given as $A = W + W^T = W + W^{-1}$. Then for $0 \leq l \leq n-1$ we have

$$A f_l = W f_l + W^{-1} f_l = (\zeta^l + \zeta^{-l}) f_l = \left(2 \cos \frac{2l\pi}{n} \right) f_l.$$

Therefore, the eigenvalues of A are

$$2 \cos \frac{2l\pi}{n}, \quad 0 \leq l \leq n-1, \quad (2.8)$$

among which the same values appear at most twice since

$$2 \cos \frac{2l\pi}{n} = 2 \cos \frac{2(n-l)\pi}{n}.$$

The assertion follows after picking up different values from (2.8). \square

Finally, we consider the path P_n ($n \geq 1$). After natural numbering the vertices, we may take the adjacency matrix of the form:

$$A_n = \begin{bmatrix} 0 & 1 & & & \\ 1 & 0 & 1 & & \\ & 1 & 0 & 1 & \\ & & \ddots & \ddots & \ddots \\ & & & 1 & 0 & 1 \\ & & & & 1 & 0 \end{bmatrix}.$$

Lemma 2.16 *The characteristic polynomials $\varphi_n(x) = \varphi(x; P_n)$ of the path P_n verify the following recurrence relation:*

$$\varphi_1(x) = x, \quad \varphi_2(x) = x^2 - 1, \quad (2.9)$$

$$\varphi_n(x) = x\varphi_{n-1}(x) - \varphi_{n-2}(x), \quad n \geq 3 \quad (2.10)$$

Proof Relations in (2.9) are immediate. Let $n \geq 3$. We have

$$\begin{aligned} \varphi_n(x) &= \begin{vmatrix} x & -1 & & & \\ -1 & x & -1 & & \\ & -1 & x & -1 & \\ & & \ddots & \ddots & \ddots \\ & & & -1 & x & -1 \\ & & & & -1 & x \end{vmatrix} = x\varphi_{n-1}(x) + \begin{vmatrix} -1 & -1 & & & \\ & x & -1 & & \\ & -1 & x & -1 & \\ & & \ddots & \ddots & \ddots \\ & & & -1 & x & -1 \\ & & & & -1 & x \end{vmatrix} \\ &= x\varphi_{n-1}(x) - \varphi_{n-2}(x), \end{aligned}$$

where we applied cofactor expansion twice, first along the first row and then along the first column. \square

Chebyshev polynomials of the second kind. Using elementary knowledge of trigonometric functions we see easily that for $n = 0, 1, 2, \dots$ there exists a polynomial $U_n(x)$ such that

$$U_n(\cos \theta) = \frac{\sin(n+1)\theta}{\sin \theta}. \quad (2.11)$$

In fact, $U_n(x)$ is determined by the recurrence relation:

$$U_0(x) = 1, \quad U_1(x) = 2x, \quad U_{n+1}(x) - 2xU_n(x) + U_{n-1}(x) = 0. \quad (2.12)$$

These polynomials $\{U_n(x)\}$ are called the *Chebyshev polynomials of the second kind*. First a few are given by

$$\begin{aligned} U_0(x) &= 1, & U_1(x) &= 2x, & U_2(x) &= 4x^2 - 1, & U_3(x) &= 8x^3 - 4x, \\ U_4(x) &= 16x^4 - 12x^2 + 1, & U_5(x) &= 32x^5 - 32x^3 + 6x. \end{aligned}$$

Theorem 2.17 $\varphi(x; P_n) = U_n(x/2)$ for $n \geq 1$.

Proof Set $\varphi_n(x) = \varphi(x; P_n)$ for $n \geq 1$ and $\varphi_0(x) = 1$. By Lemma 2.16 we have

$$\varphi_0(x) = 1, \quad \varphi_1(x) = x, \quad \varphi_n(x) = x\varphi_{n-1}(x) - \varphi_{n-2}(x), \quad n \geq 2. \quad (2.13)$$

On the other hand, we see from (2.12) that $\{U_n(x/2); n \geq 0\}$ fulfills the same recurrence relation as in (2.13) together with the initial conditions. Therefore $\varphi_n(x) = U_n(x/2)$ for all $n \geq 0$. \square

Theorem 2.18 The spectrum of the path P_n is given by

$$\text{Spec}(P_n) = \left(\begin{array}{ccccc} 2 \cos \frac{\pi}{n+1} & \cdots & 2 \cos \frac{k\pi}{n+1} & \cdots & 2 \cos \frac{n\pi}{n+1} \\ 1 & \cdots & 1 & \cdots & 1 \end{array} \right), \quad 1 \leq k \leq n.$$

Proof First we need to find the zeroes of the Chebyshev polynomial of the second kind. It follows from (2.11) that $U_n(\cos \theta) = 0$ for

$$\theta = \theta_k = \frac{k\pi}{n+1}, \quad k = 1, 2, \dots, n.$$

Hence, $U_n(x_k) = 0$ for $x_k = \cos \theta_k$, $1 \leq k \leq n$. Since $U_n(x)$ is a polynomial of degree n and x_1, x_2, \dots, x_n are mutually distinct, they exhaust the zeroes of $U_n(x)$ and each x_k is simple. Then by Theorem 2.17 the zeroes of $\varphi(x; P_n)$ are all simple and given by $2x_1, \dots, 2x_n$, which gives $\text{Spec}(P_n)$ as desired. \square

Exercise 2.19 Let $G = (V, E)$ be a graph. Let $a \in V$ be a vertex with $\deg(a) = 1$ and $b \in V$ a unique vertex adjacent to a . Prove that

$$\varphi(x; G) = x\varphi(x; G \setminus \{a\}) - \varphi(x; G \setminus \{a, b\}),$$

where $G \setminus W$ is a short-hand notation for $G[V \setminus W]$, the induced subgraph spanned by $V \setminus W$, where $W \subset V$.

Exercise 2.20 Using Theorem 2.18, prove the following identity:

$$\prod_{k=1}^m 2 \cos \frac{k\pi}{2m+1} = 1, \quad m \geq 1.$$

Exercise 2.21 (see Corollary 2.11) Show that $s(G) = \dim \mathcal{A}(G) = \text{diam}(G) + 1$ holds for $G = K_n$ with $n \geq 1$ and for $G = P_n$ with $n \geq 1$. Give a small graph G such that $s(G) > \text{diam}(G) + 1$.

Exercise 2.22 Show that the characteristic polynomial and the spectrum of the complete bipartite graph $K_{k,l}$ are given by

$$\varphi(x; K_{k,l}) = x^{k+l-2}(x^2 - kl), \quad \text{Spec}(K_{k,l}) = \begin{pmatrix} -\sqrt{kl} & 0 & \sqrt{kl} \\ 1 & k+l-2 & 1 \end{pmatrix}$$

2.4 Bounds of Spectra

We first recall useful results known under the name of Perron–Frobenius theory, for further details see Horn–Johnson [80, Chap. 8] and Seneta [123]. For a matrix $A \in M(n, \mathbb{C})$ the *spectral radius* is defined to be the maximal modulus of its eigenvalue and is denoted by $\rho(A)$.

Proposition 2.23 Let $A = [a_{ij}] \in M(n, \mathbb{C})$ and assume that $a_{ij} \geq 0$ for all i, j . Then $\rho(A)$ is an eigenvalue of A and there exists a nonzero vector $\xi = [\xi_i] \in \mathbb{C}^n$ such that $A\xi = \rho(A)\xi$ and $\xi_i \geq 0$ for all i .

A matrix $A \in M(n, \mathbb{C})$ is called *reducible* if there is a permutation matrix S such that $S^T A S$ is partitioned in the form:

$$S^T A S = \begin{bmatrix} B & C \\ O & D \end{bmatrix}, \quad B \in M(r, \mathbb{C}), \quad D \in M(n-r, \mathbb{C}), \quad 1 \leq r \leq n-1.$$

A matrix $A \in M(n, \mathbb{C})$ is called *irreducible* if it is not reducible.

Proposition 2.24 Let $A = [a_{ij}] \in M(n, \mathbb{C})$ and assume that $a_{ij} \geq 0$ for all i, j . If A is irreducible, $\rho(A)$ is an algebraically simple eigenvalue of A . Moreover, there exists a vector $\xi = [\xi_i] \in \mathbb{C}^n$ such that $A\xi = \rho(A)\xi$ and $\xi_i > 0$ for all i . (This vector is called the *Perron vector*.)

Definition 2.25 The *spectral radius* of a finite graph G is defined to be the spectral radius of the adjacency matrix, and is denoted by $\rho(G)$.

As a direct consequence from Propositions 2.23 and 2.24 we have the following

Theorem 2.26 *Let G be a finite graph. Then $\rho(G)$ coincides with the maximal eigenvalue of G . If G is a finite connected graph, $\rho(G)$ is simple and there exists an eigenvector with positive entries.*

Bounds of $\rho(G)$ is interesting. We start with a simple upper bound.

Theorem 2.27 *For a finite graph $G = (V, E)$ it holds that*

$$\rho(G) \leq \left\{ \frac{2|E|(|V| - 1)}{|V|} \right\}^{1/2}. \quad (2.14)$$

Proof Let A be the adjacency matrix of G and $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$ the eigenvalues of A , where $n = |V|$. Then $\rho(G) = \lambda_1$. Since $\text{Tr } A = 0$ and $\text{Tr } A^2 = 2|E|$, we have

$$\sum_{k=1}^n \lambda_k = 0, \quad \sum_{k=1}^n \lambda_k^2 = 2|E|.$$

Then, applying the Schwartz inequality, we obtain

$$\lambda_1^2 = \left\{ \sum_{k=2}^n (-\lambda_k) \right\}^2 \leq (n-1) \sum_{k=2}^n \lambda_k^2 = (n-1)(2|E| - \lambda_1^2),$$

from which we obtain $n\lambda_1^2 \leq 2(n-1)|E|$. Then (2.14) follows immediately. \square

Some statistics concerning the degrees of vertices play an interesting role. We set

$$d_{\max}(G) = \max\{\deg(x); x \in V\}, \quad d_{\min}(G) = \min\{\deg(x); x \in V\},$$

and define the *mean degree* of G by

$$\bar{d}(G) = \frac{1}{|V|} \sum_{x \in V} \deg(x) = \frac{2|E|}{|V|}. \quad (2.15)$$

Theorem 2.28 *For a finite graph $G = (V, E)$ it holds that*

$$d_{\min}(G) \leq \bar{d}(G) \leq \rho(G) \leq d_{\max}(G).$$

Proof Obviously, $d_{\min}(G) \leq \bar{d}(G) \leq d_{\max}(G)$. We first prove that $\bar{d}(G) \leq \rho(G)$. Let A be the adjacency matrix of G . From elementary linear algebra we know that

$$\lambda_{\min}(A) \leq \frac{\langle f, Af \rangle}{\langle f, f \rangle} \leq \lambda_{\max}(A) \quad \text{for all } f \in C(V), f \neq 0, \quad (2.16)$$

where $\lambda_{\max}(A)$ and $\lambda_{\min}(A)$ are the maximal and minimal eigenvalues of A , respectively. Now we take ψ to be a constant function defined by $\psi(x) = 1$ for all $x \in V$.

Then,

$$\langle \psi, A\psi \rangle = \sum_{x,y \in V} \overline{\psi(x)} (A)_{xy} \psi(y) = \sum_{x,y \in V} (A)_{xy} = 2|E|.$$

Since $\langle \psi, \psi \rangle = |V|$, we have

$$\frac{\langle \psi, A\psi \rangle}{\langle \psi, \psi \rangle} = \frac{2|E|}{|V|} = \bar{d}(G). \quad (2.17)$$

Combining (2.16) and (2.17), we obtain $\bar{d}(G) \leq \lambda_{\max}(A) = \rho(G)$, as desired.

Next we show that $\rho(G) \leq d_{\max}(G)$. We set $\lambda = \rho(G)$ for simplicity. Now we choose an eigenvector $f \in C(V)$ such that $Af = \lambda f$, $f \neq 0$, and $f(x) \in \mathbb{R}$ for all $x \in V$ (here we do not need to refer to a Perron vector). In that case, we may assume without loss of generality that $\alpha \equiv \max\{f(x) ; x \in V\} > 0$. Let $o \in V$ be chosen in such a way that $f(o) = \alpha$. Then,

$$\lambda \alpha = \lambda f(o) = Af(o) = \sum_{y \in V} (A)_{oy} f(y) \leq \alpha \deg(o).$$

Since $\alpha > 0$, we have $\lambda \leq \deg(o)$, which implies that $\rho(G) \leq d_{\max}(G)$. \square

Corollary 2.29 *If G is a finite regular graph with degree κ , we have $\rho(G) = \kappa$.*

Exercise 2.30 Show that $\min \text{Spec}(G) \geq -2$ for any line graph G of a finite graph.

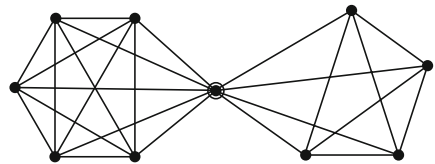
2.5 Star Products

Consider two graphs $G_i = (V_i, E_i)$ with distinguished vertex $o_i \in V_i$ for $i = 1, 2$. The *star product* $G_1 \star G_2 = (G_1, o_1) \star (G_2, o_2)$ is a graph obtaining by gluing G_1 and G_2 at the distinguished vertices (Fig. 2.1), for further study see Sect. 7.7.

Lemma 2.31 *Let $m \geq 1$ and $n \geq 1$ be natural numbers. Let A, B, C, D be $m \times m$ -, $m \times n$ -, $n \times m$ -, and $n \times n$ -matrices over the complex number field, respectively. If A is invertible, we have*

$$\det \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \det A \det(D - CA^{-1}B). \quad (2.18)$$

Fig. 2.1 Star product $K_6 \star K_5$



Similarly, if D is invertible, we have

$$\det \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \det D \det(A - BD^{-1}C) \quad (2.19)$$

Proof (2.18) follows immediately from the obvious identity:

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} A & O \\ C & I \end{bmatrix} \begin{bmatrix} I & A^{-1}B \\ O & D - CA^{-1}B \end{bmatrix},$$

where I stands for the identity matrix. The proof of (2.19) is similar. \square

Let B_1, B_2 and A be the adjacency matrices of G_1, G_2 and $G = G_1 \star G_2$, respectively. We number the vertices in such a way that $V_1 = \{1, 2, \dots, m\}$ and $V_2 = \{m, m+1, 2, \dots, m+n-1\}$, where o_1 and o_2 are given the common number m . Then the adjacency matrix A is of the form:

$$A = \begin{bmatrix} \boxed{B_1} & \\ & \boxed{B_2} \end{bmatrix} = \begin{bmatrix} B_1 & F \\ F^T & \tilde{B}_2 \end{bmatrix}, \quad (2.20)$$

where \tilde{B}_2 is the adjacency matrix of $G_2 \setminus \{o_2\}$, and F is an $m \times (n-1)$ matrix such that $(F)_{ij} = 0$ for all $1 \leq i \leq m-1$ and $1 \leq j \leq n-1$. By Lemma 2.31 we obtain

$$\begin{aligned} \varphi(x) &= \det(x - A) = \det \begin{bmatrix} x - B_1 & -F \\ -F^T & x - \tilde{B}_2 \end{bmatrix} \\ &= \det(x - B_1) \det(x - \tilde{B}_2 - F^T(x - B_1)^{-1}F). \end{aligned} \quad (2.21)$$

We define an $(n-1) \times (n-1)$ matrix \tilde{F} by

$$(\tilde{F})_{ij} = (F)_{mi}(F)_{mj}, \quad 1 \leq i, j \leq n-1.$$

By simple calculation, we have

$$F^T(x - B_1)^{-1}F = ((x - B_1)^{-1})_{mm}\tilde{F} = \det(x - B_1)^{-1} \det(x - \tilde{B}_1)\tilde{F}, \quad (2.22)$$

where \tilde{B}_1 is a principal submatrix of B_1 obtained by deleting m th row and column, in other words, \tilde{B}_1 is the adjacency matrix of $G_1 \setminus \{o_1\}$. Combining (2.21) and (2.22) we obtain

$$\varphi(x) = \det(x - B_1) \det(x - \tilde{B}_2 - \det(x - B_1)^{-1} \det(x - \tilde{B}_1)\tilde{F}).$$

Summing up, we have established the following

Theorem 2.32 *For $i = 1, 2$, let $G_i = (V_i, E_i)$ be a graph with a distinguished vertex $o_i \in V_i$. Let B_i be the adjacency matrix of G_i . Then the characteristic polynomial of the star product $G_1 \star G_2 = (G_1, o_1) \star (G_2, o_2)$ is given by*

$$\varphi(x; G_1 \star G_2) = \det(x - B_1) \det \left(x - \tilde{B}_2 - \frac{\det(x - \tilde{B}_1)}{\det(x - B_1)} \tilde{F} \right),$$

where \tilde{B}_i is the adjacency matrix of $G_i \setminus \{o_i\}$ and $\tilde{F} = [f_{ij}]$ is a matrix indexed by $(V_2 \setminus \{o_2\}) \times (V_2 \setminus \{o_2\})$ defined by $f_i = (B_2)_{o_2 i}$.

Corollary 2.33 *For the star product of the complete graphs we have*

$$\begin{aligned} \varphi(x; K_m \star K_n) \\ = (x + 1)^{m+n-4} ((x - m - n + 2)(x + 1)^2 + (m - 1)(n - 1)(x + 2)), \end{aligned}$$

where $m \geq 1$ and $n \geq 1$.

Proof Noting that \tilde{B}_i is the adjacency matrix of a complete graph and $\tilde{F} = J$, we need only to apply Theorem 2.32. \square

Notes. The spectral graph theory has a long history. There are many comprehensive books, some of which are Brouwer–Haemers [28], Cvetković–Doob–Sachs [41], Cvetković–Rowlinson–Simić [39, 40] and van Mieghem [136]. For spectra of the normalized Laplacian, see also Chung [35].

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