

### 4.3 Matrices related with Graphs

Let  $X$  be a graph on  $n$  vertices and let us fix a labeling of the vertices of  $X$ . Then the *adjacency matrix* of the graph  $X$ , on  $n$  vertices, denoted  $A(X) = [a_{ij}]$  (or  $A$ ), is an  $n \times n$  matrix with  $a_{ij} = 1$ , if the  $i$ -th vertex is adjacent to the  $j$ -th vertex and 0, otherwise. Note that another labeling of the vertices of  $X$  gives rise to another matrix  $B$  such that  $B = S^{-1}AS$ , for some permutation matrix  $S$  (for a permutation matrix  $S^t = S^{-1}$ ). Hence, we talk of the adjacency matrix of a graph  $X$  and we do not worry about the labeling of the vertices of  $X$ .

Clearly,  $A$  is a real symmetric matrix. Hence,  $A$  has  $n$  real eigenvalues,  $A$  is diagonalizable, and the eigenvectors can be chosen to form an orthonormal basis of  $\mathbb{R}^n$ . The eigenvalues, eigenvectors, the minimal polynomial and the characteristic polynomial of a graph  $X$  are defined to be that of its adjacency matrix.

The definition of the adjacency matrix of a graph indicates that there is one to one correspondence between graphs and its adjacency matrix. For the graph  $X$ , drawn in Figure 4.10, the adjacency matrix of  $X$  equals

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

For example, it can be easily shown that a graph  $X$  is disconnected if and only if there exists a permutation matrix  $P$  such that  $A(X) = \begin{bmatrix} A_{11} & \mathbf{0} \\ \mathbf{0} & A_{22} \end{bmatrix}$ , for some matrices  $A_{11}$  and  $A_{22}$ . Recall that a symmetric permutation of the rows and columns of  $A(X)$  corresponds to another labeling of the graph  $X$ .

We now state and prove a few results related to the adjacency matrix of a graph.

**Theorem 4.3.1.** *Let  $A$  be the adjacency matrix of a graph  $X = (V, E)$  with  $V = \{v_1, v_2, \dots, v_n\}$ . Then, for each positive integer  $k$ , the  $(i, j)$ -th entry of  $A^k$  gives the number of walks of length  $k$  from the vertex  $v_i$  to the vertex  $v_j$ .*

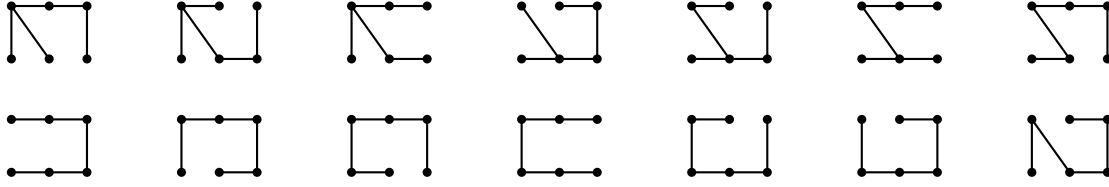
*Proof.* We prove the result by induction of  $k$ . The result is clearly true for  $k = 1$ . So, let the result be true for  $k$  and let us consider  $(A^{k+1})_{ij}$ , for  $1 \leq i \leq j \leq n$ . By definition,

$$(A^{k+1})_{ij} = \sum_{\ell=1}^n (A^k)_{i\ell} a_{\ell j} = \sum_{v_\ell \sim v_j} (A^k)_{i\ell}.$$

By induction hypothesis,  $(A^k)_{i\ell}$  gives the number of walks of length  $k$  from the vertex  $v_i$  to the vertex  $v_\ell$  and then a walk of length from  $v_\ell$  to  $v_j$ . Thus, the required result follows.  $\blacksquare$

One now immediately has the following corollary. We skip the proof as it is a direct application of Theorem 4.3.1.

**Corollary 4.3.2.** *Let  $A$  be the adjacency matrix of a graph  $X$  on  $n$  vertices. Then  $(A^2)_{i,j}$ , for  $i \neq j$ , gives the number of paths of length 2 from the vertex  $v_i$  to the vertex  $v_j$ . Also,  $(A^2)_{ii}$ , for  $1 \leq i \leq n$ , gives  $\deg(v_i)$ .*



The labeled spanning subgraphs of  $X$  (labeling not shown)

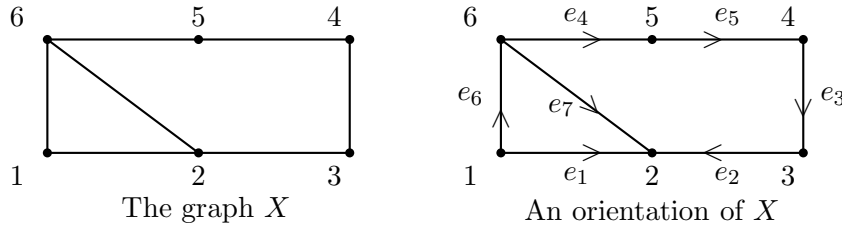


Figure 4.10: A graph on 6 vertices and its spanning trees

One also defines the  $\{-1, 0, 1\}$ -incidence (or the edge incidence) matrix of a graph  $X = (V, E)$ , denoted  $Q(X)$  or in short  $Q$ , as follows:

Give an orientation to each edge of  $X$  and let  $V = \{v_1, v_2, \dots, v_n\}$  and  $E = \{e_1, e_2, \dots, e_m\}$ . Then, the matrix  $Q = (q_{ij})$  is defined to be an  $n \times m$  matrix that has its rows and columns indexed by the elements of  $V$  and  $E$ , respectively with

$$q_{ij} = \begin{cases} 1 & e_j \text{ originates at the vertex } v_i, \\ -1 & e_j \text{ terminates at the vertex } v_i, \\ 0 & \text{otherwise.} \end{cases}$$

The matrix  $QQ^t$ , denoted  $L(X)$  or in short  $L$ , is called the Laplacian matrix of  $X$ . It can be easily verified that  $L = D - A$ , where  $D$  is the diagonal matrix with the  $i$ -th diagonal entry being  $\deg(v_i)$ . Consider the oriented graph  $X$  in Figure 4.10. Based on this orientation, the edge incidence matrix and the Laplacian matrix of  $X$ , respectively equal

$$Q = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ -1 & -1 & 0 & 0 & 0 & 0 & -1 \\ 0 & 1 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & -1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & -1 & 1 \end{bmatrix} \quad \text{and} \quad L = \begin{bmatrix} 2 & -1 & 0 & 0 & 0 & -1 \\ -1 & 3 & -1 & 0 & 0 & -1 \\ 0 & -1 & 2 & -1 & 0 & 0 \\ 0 & 0 & -1 & 2 & -1 & 0 \\ 0 & 0 & 0 & -1 & 2 & -1 \\ -1 & -1 & 0 & 0 & -1 & 3 \end{bmatrix}.$$

Note that depending on the orientation of the graph  $X$ , the incidence matrices may differ, but there is no change in its Laplacian matrix. Hence, there is again a one to one correspondence between the Laplacian matrix and the corresponding graph.

Let  $X$  be a graph on  $n$  vertices and let  $L$  be its Laplacian matrix. Then note that if we add the 2-nd, 3-rd and so on till  $n$ -th row of  $L$  to the first row of  $L$  then the first row of  $L$  equals the zero vector. Hence,  $\det(L) = 0$ . A similar statement is true for the columns as  $L$  is a symmetric matrix. The above argument also implies that 0 is an eigenvalue of  $L$  with corresponding eigenvector  $\mathbf{e} = (1, 1, \dots, 1)^t$ . Also,  $L = QQ^t$  and hence  $L$  is a positive semi-definite matrix. That is, all the eigenvalues of  $L$  are non-negative and there is a non-zero vector  $\mathbf{x}_0$  such that  $\mathbf{x}_0^t L \mathbf{x}_0 = 0$ . Using a result in matrix theory, it can be easily shown that the graph  $X$  is disconnected if and only if the multiplicity of the eigenvalue 0 is at least 2.