of G, the column of B corresponding to the oriented edge  $\sigma(\{v_i, v_j\})$  has zero entries except for a +1 and a -1 in position i and position j or vice-versa, so we have

$$z_i = z_i$$
.

An easy induction on the length of the path shows that if there is a path from  $v_i$  to  $v_j$  in G (unoriented), then  $z_i = z_j$ . Therefore, z has a constant value on any connected component of G. It follows that every vector  $z \in \text{Ker}(B^\top)$  can be written uniquely as a linear combination

$$z = \lambda_1 z^1 + \dots + \lambda_c z^c,$$

where the vector  $z^i$  corresponds to the *i*th connected component  $K_i$  of G and is defined such that

$$z_j^i = \begin{cases} 1 & \text{iff } v_j \in K_i \\ 0 & \text{otherwise.} \end{cases}$$

This shows that  $\dim(\operatorname{Ker}(B^{\top})) = c$ , and that  $\operatorname{Ker}(B^{\top})$  has a basis consisting of indicator vectors.

Since  $B^{\top}$  is a  $n \times m$  matrix, we have

$$m = \dim(\operatorname{Ker}(B^{\top})) + \operatorname{rank}(B^{\top}),$$

and since we just proved that  $\dim(\operatorname{Ker}(B^{\top})) = c$ , we obtain  $\operatorname{rank}(B^{\top}) = m - c$ . Since B and  $B^{\top}$  have the same rank,  $\operatorname{rank}(B) = m - c$ , as claimed.

**Definition 20.12.** Following common practice, we denote by  $\mathbf{1}$  the (column) vector (of dimension m) whose components are all equal to 1.

Since every column of B contains a single +1 and a single -1, the rows of  $B^{\top}$  sum to zero, which can be expressed as

$$B^{T} \mathbf{1} = 0.$$

According to Proposition 20.1, the graph G is connected iff B has rank m-1 iff the nullspace of  $B^{\top}$  is the one-dimensional space spanned by **1**.

In many applications, the notion of graph needs to be generalized to capture the intuitive idea that two nodes u and v are linked with a degree of certainty (or strength). Thus, we assign a nonnegative weight  $w_{ij}$  to an edge  $\{v_i, v_j\}$ ; the smaller  $w_{ij}$  is, the weaker is the link (or similarity) between  $v_i$  and  $v_j$ , and the greater  $w_{ij}$  is, the stronger is the link (or similarity) between  $v_i$  and  $v_j$ .

**Definition 20.13.** A weighted graph is a pair G = (V, W), where  $V = \{v_1, \ldots, v_m\}$  is a set of nodes or vertices, and W is a symmetric matrix called the weight matrix, such that  $w_{ij} \geq 0$  for all  $i, j \in \{1, \ldots, m\}$ , and  $w_{ii} = 0$  for  $i = 1, \ldots, m$ . We say that a set  $\{v_i, v_j\}$  is an edge iff  $w_{ij} > 0$ . The corresponding (undirected) graph (V, E) with  $E = \{\{v_i, v_j\} \mid w_{ij} > 0\}$ , is called the underlying graph of G.