

# Lecture 16

This is essentially [10, Lecture 2] with small modifications.

## 8.3.4 The graph Laplacian of some fundamental graphs

We now examine the eigenvalues and eigenvectors of the graph Laplacians of some fundamental graphs on  $n$  vertices  $V = \{1, \dots, n\}$ :

1. The complete graph  $K_n$ , for which  $E = \{(i, j) : i \neq j\}$ .
2. The star graph  $S_n$ , for which  $E = \{(1, i) : 2 \leq i \leq n\}$ .
3. The ring graph  $R_n$ , for which  $E = \{(i, i+1) : 1 \leq i < n\} \cup \{(1, n)\}$ .
4. The path graph  $P_n$ , for which  $E = \{(i, i+1) : 1 \leq i < n\}$ .

**Proposition 3.** *The graph Laplacian of  $K_n$  has eigenvalue 0 with multiplicity 1 and eigenvalue  $n$  with multiplicity  $n - 1$ .*

*Proof.* Since  $K_n$  is connected, the multiplicity of the 0 eigenvalue follows from Proposition 2.

To compute the remaining nonzero eigenvalues, let  $\varphi$  be any non-zero vector orthogonal to  $\mathbf{1}$ , which implies:

$$0 = \langle \varphi, \mathbf{1} \rangle = \sum_{i=1}^n \varphi[i]. \quad (49)$$

Without loss of generality we can take  $\varphi[1] \neq 0$ . Using (49), we then have:

$$L_{K_n} \varphi[1] = (n-1)\varphi[1] - \sum_{i=2}^n \varphi[i] = (n-1)\varphi[1] + \varphi[1] = n\varphi[1].$$

Thus any other eigenvector of  $L_{K_n}$  must have eigenvalue  $n$ . □

To analyze  $S_n$ , we first need the following lemma.

**Lemma 1.** *Let  $G = (V, E)$  be a graph, and let  $i$  and  $j$  be vertices of degree 1 that are both connected to another vertex  $k$ . Then the vector*

$$\varphi[\ell] = \begin{cases} 1 & \ell = i, \\ -1 & \ell = j, \\ 0 & \text{otherwise,} \end{cases} \quad (50)$$

*is an eigenvector of  $L_G$  with eigenvalue 1.*

*Proof.* Check on your own! □

Note that the existence of the eigenvector  $\varphi$  defined in (50) implies that  $\varphi'[i] = \varphi'[j]$  for every other eigenvector  $\varphi'$  with eigenvalue other than 1.

**Proposition 4.** *The graph Laplacian of  $S_n$  has eigenvalue 0 with multiplicity 1, eigenvalue 1 with multiplicity  $n - 2$ , and eigenvalue  $n$  with multiplicity 1.*

*Proof.* Check on your own! Use Proposition 2, Lemma 1, and the fact that

$$\text{Tr}(L_G) = \sum_{i=1}^n \lambda_i,$$

for any graph  $G$  (and in fact, any  $n \times n$  matrix with  $n$  eigenvalues). □

**Proposition 5.** *The graph Laplacian of  $R_n$  has eigenvectors*

$$\varphi_k[i] = \sin(2\pi ki/n), \tag{51}$$

$$\psi_k[i] = \cos(2\pi ki/n), \tag{52}$$

for  $1 \leq k \leq n/2$ . When  $n$  is even,  $\varphi_{n/2}$  is the all zero vector, so we only have  $\psi_{n/2}$ . Eigenvectors  $\varphi_k$  and  $\psi_k$  have eigenvalue  $2 - 2 \cos(2\pi k/n)$ .

*Proof.* We are going to use the following trigonometric identity:

$$\sin(\alpha \pm \beta) = \sin \alpha \cos \beta \pm \cos \alpha \sin \beta,$$

which implies:

$$\sin(\alpha - \beta) + \sin(\alpha + \beta) = 2 \sin \alpha \cos \beta.$$

Now observe that:

$$\begin{aligned} L_{R_n} \varphi_k[i] &= 2\varphi_k[i] - \varphi_k[i-1] - \varphi_k[i+1], \\ &= 2 \sin\left(\frac{2\pi ki}{n}\right) - \left[ \sin\left(\frac{2\pi ki}{n} - \frac{2\pi k}{n}\right) + \sin\left(\frac{2\pi ki}{n} + \frac{2\pi k}{n}\right) \right], \\ &= 2 \sin\left(\frac{2\pi ki}{n}\right) - 2 \cos\left(\frac{2\pi k}{n}\right) \sin\left(\frac{2\pi ki}{n}\right), \\ &= \left[ 2 - 2 \cos\left(\frac{2\pi k}{n}\right) \right] \sin\left(\frac{2\pi ik}{n}\right). \end{aligned}$$

The calculation for  $\psi_k$  is similar. □

**Proposition 6.** *The graph Laplacian of  $P_n$  has eigenvectors*

$$\phi_k[i] = \cos\left(\frac{\pi ki}{n} - \frac{\pi k}{2n}\right),$$

with corresponding eigenvalues:

$$\lambda_k = 2 - 2\cos(\pi k/n).$$

*Proof.* This proof is more involved than the previous ones. We will derive the eigenvectors and eigenvalues of  $L_{P_n}$  by treating  $P_n$  as a quotient of  $R_{2n}$ . To do so, let  $i$  be an arbitrary vertex of  $P_n$ ; identify it with the vertices  $i$  and  $2n + 1 - i$  of  $R_{2n}$ .

Now consider the vector:

$$\phi_k[i] = \cos\left(\frac{\pi ki}{n} - \frac{\pi k}{2n}\right).$$

Notice that:

$$\phi_k[i] = \phi_k[2n + 1 - i],$$

indeed:

$$\begin{aligned} \phi_k[2n + i - i] &= \cos\left(\frac{\pi k(2n + 1 - i)}{n} - \frac{\pi k}{2n}\right), \\ &= \cos\left(2\pi k + \frac{\pi k}{n} - \frac{\pi ki}{n} - \frac{\pi k}{2n}\right), \\ &= \cos\left(-\frac{\pi ki}{n} + \frac{\pi k}{2n}\right), \\ &= \cos\left(\frac{\pi ki}{n} - \frac{\pi k}{2n}\right), \\ &= \phi_k[i]. \end{aligned}$$

Furthermore,

$$\begin{aligned} \phi_k[i] &= \cos\left(\frac{2\pi ki}{2n} - \frac{\pi k}{2n}\right), \\ &= \cos\left(\frac{2\pi ki}{2n}\right) \cos\left(\frac{\pi k}{2n}\right) + \sin\left(\frac{2\pi ki}{2n}\right) \sin\left(\frac{\pi k}{2n}\right), \\ &= \varphi_k[i] \cos\left(\frac{\pi k}{2n}\right) + \psi_k[i] \sin\left(\frac{\pi k}{2n}\right), \end{aligned}$$

where  $\varphi_k$  and  $\psi_k$  are eigenvectors of  $R_{2n}$ , defined by (51) and (52), respectively. It follows that  $\phi_k$  is an eigenvector of  $L_{R_{2n}}$  with eigenvalue  $\lambda_k = 2 - 2 \cos(2\pi k/(2n)) = 2 - 2 \cos(\pi k/n)$ .

Now set:

$$\forall 1 \leq i \leq n, \quad v_k[i] = \phi_k[i].$$

We claim that  $v_k$  is eigenvector of  $L_{P_n}$  with eigenvalue  $\lambda_k$ . To see this, let  $1 < i < n$  and compute:

$$\begin{aligned} L_{P_n} v_k[i] &= 2v_k[i] - v_k[i-1] - v_k[i+1], \\ &= \frac{1}{2} \left( 2\phi_k[i] - \phi_k[i-1] - \phi_k[i+1] + \dots \right. \\ &\quad \left. \dots + 2\phi_k[2n+1-i] - \phi_k[2n+1-(i-1)] - \phi_k[2n+1-(i+1)] \right), \\ &= \frac{1}{2} \left( L_{R_{2n}} \phi_k[i] + L_{R_{2n}} \phi_k[2n+1-i] \right), \\ &= \frac{1}{2} \left( \lambda_k \phi_k[i] + \lambda_k \phi_k[2n+1-i] \right), \\ &= \lambda_k v_k[i]. \end{aligned}$$

For  $i = 1$  we have:

$$\begin{aligned} L_{P_n} v_k[1] &= v_k[1] - v_k[2], \\ &= 2v_k[1] - v_k[2] - v_k[1], \\ &= 2\phi_k[1] - \phi_k[2] - \phi_k[2n], \\ &= L_{R_{2n}} \phi_k[1], \\ &= \lambda_k v_k[1]. \end{aligned}$$

The calculation for  $i = n$  is similar. □

We have now seen that the  $k$ th eigenvector of the path graph alternates in sign  $k - 1$  times. This is consistent with our intuition that the Laplacian of the path graph is a discretization of a continuous string, and that its eigenvectors are approximations of its fundamental modes of vibration when its ends are free.

If this intuition is correct, then it should continue to be true even if we discretize a string whose material changes along its length. This corresponds to a weighted path graph.

**Exercises**

*Exercise 24.* Prove Lemma 1.

*Exercise 25.* Prove Proposition 4.

*Exercise 26.* Write a function that takes as inputs the vertices  $V$  and edges  $E$  of a graph  $G$ , and outputs the graph Laplacian  $L_G$  and its eigenvectors and eigenvalues. Use this code to verify the results of this lecture on the graphs  $K_n$ ,  $S_n$ ,  $R_n$ , and  $P_n$  (test several different values of  $n$  for each graph).

## References

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