ORIE 6334 Bridging Continuous and Discrete Optimization

Sep 30, 2019

Lecture 8

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1 The Matrix-Tree Theorem

In this lecture, we continue to see the usefulness of the graph Laplacian via its connection to yet another standard concept in graph theory, the spanning tree. Let A[i] be the matrix A with its ith row and column removed. We will give two different proofs of the following.

Theorem 1 (Kirchhoff's Matrix-Tree Theorem) The number of spanning trees in a graph G is given by $det(L_G[i])$, for any i.

For the first proof, we will need the following fact.

Fact 1 Let $A \in \mathbb{R}^{n \times n}$ and take E_{ii} to be the matrix with 1 in the $(i,i)^{th}$ entry and 0s elsewhere. Then,

$$\det(A + E_{ii}) = \det(A) + \det(A[i]).$$

This claim is evident if one considers the permutation sum definition of the determinant

$$\det(A = (a_{ij})) = \sum_{\pi} \operatorname{sgn}(\pi) \, a_{1\pi(1)} a_{2\pi(2)} \dots a_{n\pi(n)},$$

since we've increased the $(i, i)^{\text{th}}$ entry, a_{ii} , to $a_{ii} + 1$. Hence, we get the original determinant plus what is effectively a sum over all permutations of $[n] \setminus \{i\}$, avoiding the i^{th} row and column, i.e. $\det(A[i])$.

Proof of Theorem 1:

Our first proof will be by induction on the number of vertices and edges of the graph G. Base case: If G is an empty graph on two vertices, then

$$L_G = \left[\begin{array}{cc} 0 & 0 \\ 0 & 0 \end{array} \right],$$

so $L_G[i] = [0]$ and $det(L_G[i]) = 0$, as desired.

<u>Inductive step</u>: In what follows, let $\tau(G)$ denote the number of spanning trees in G, let G-e denote the graph with edge e removed, and let G/e denote the graph with edge e contracted. See below for an illustration of graph contraction.

If i is an isolated vertex, then G admits no spanning trees, and there are zeros along the ith row and column of L_G . Thus, $\det(L_G[i]) = \det(L_{G-i}) = 0$, as desired.

⁰These notes are slightly modified from a Fall 2016 version scribed by Faisal Alkaabneh. This lecture is derived from Cvetković, Rowlinson, and Simić, An Introduction to the Theory of Graph Spectra, Sections 7.1 and 7.2, and Godsil and Royle, Algebraic Graph Theory, Section 13.2.

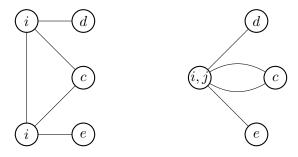


Figure 1: Example of contracting nodes i and j into a single node.

Otherwise, we can suppose that there exists an edge e = (i, j) incident to i. For any spanning tree T, either $e \in T$ or $e \notin T$. We note that $\tau(G/e)$ gives the number of trees T with $e \in T$, while $\tau(G - e)$ gives the number of trees T with $e \notin T$. Thus,

$$\tau(G) = \tau(G \backslash e) + \tau(G - e).$$

Note that the first term is G with one fewer edge, while the second has one fewer vertex, so these will serve as the basis of our induction.

First we try to relate L_G to L_{G-e} , and we observe that $L_G[i] = L_{G-e}[i] + E_{jj}$ (that is, if we remove edge e, then the resulting Laplacian only differs from $L_G[i]$ in the degree of j). Thus, by Fact 1,

$$\det(L_G[i]) = \det(L_{G-e} + E_{jj})$$

$$= \det(L_{G-e}[i]) + \det(L_{G-e}[i,j])$$

$$= \det(L_{G-e}[i]) + \det(L_G[i,j]),$$

where by $L_G[i,j]$ we mean L_G with both the *i*th and *j*th rows and columns removed; the last equality follows since, once we've removed both the *i*th and *j*th rows and columns, there's no difference between L_G and L_{G-e} for e = (i,j).

Now we will relate L_G to $L_{G/e}$. Suppose we contract i onto j (so that $L_{G/e}$ has no row or column corresponding to i). Then, $L_{G/e}[j] = L_G[i,j]$.

Thus, we have that

$$\det(L_G[i]) = \det(L_{G-e}[i]) + \det(L_{G/e}[j])$$

= $\tau(G - e) + \tau(G/e) = \tau(G)$.

where the second equation follows by induction; this completes the proof.

For the second proof of the theorem, we need the following fact which explains how to take the determinant of the product of rectangular matrices.

Fact 2 (Cauchy-Binet Formula) Let $A \in \mathbb{R}^{n \times m}$, $B \in \mathbb{R}^{m \times n}$, with $m \geq n$. Let A_S (respectively B_S) be the submatrices formed by taking the columns (respectively rows) indexed by $S \subseteq [m]$ of A (respectively B).

Let $\binom{[m]}{n}$ be the set of all size n subsets of [m]. Then,

$$\det(AB) = \sum_{S \in \binom{[m]}{n}} \det(A_S) \det(B_S).$$

Recall that $L_G = \sum_{(i,j) \in E} (e_i - e_j)(e_i - e_j)^T$. Thus, we can write $L_G = BB^T$ where $B \in \mathbb{R}^{m \times n}$ has one column per edge (i,j), with the column equal to $(e_i - e_j)$. (Since we can write $L_G = BB^T$, this is yet another proof that L_G is positive semidefinite.) Then, if B[i] denotes B with its i^i row omitted, we have $L_G[i] = B[i]B[i]^T$. We let $B_S[i]$ denote B[i] with just the columns indexed by $S \subseteq E$.

We also need the following lemma, whose proof we defer for a moment.

Lemma 2 For $S \subseteq E$, |S| = n - 1,

$$|\det(B_S[i])| = \begin{cases} 1, & \text{if } S \text{ is a spanning tree} \\ 0, & \text{otherwise.} \end{cases}$$

The second proof of the matrix-tree theorem now becomes very short.

Proof of Theorem 1:

$$\det(L_G[i]) = \det(B[i]B[i]^T)$$

$$= \sum_{S \in \binom{E}{n-1}} (\det(B_S[i]))(\det(B_S[i]))$$

$$= \tau(G),$$

where the second equation follows by the Cauchy-Binet formula, and the third by Lemma 2.

We can now turn to the proof of the lemma.

Proof of Lemma 2:

Assume without loss of generality that the edges of $B_S[i]$ are "directed" arbitrarily; that is, we can change the column corresponding to (i, j) from $e_i - e_j$ to $e_j - e_i$, since this only flips the sign of the determinant.

If $S \subseteq E$ with |S| = n - 1, and S is not a spanning tree, then it must contain a cycle. We can then direct the edges around this cycle. If we then sum the columns of $B_S[i]$ corresponding to the cycle, we obtain the 0 vector, which implies that the columns of $B_S[i]$ are linearly dependent, and thus $\det(B_S[i]) = 0$.

Now we suppose that S is a spanning tree and continue via induction on n.

Base case If n=2, then

$$B_S = \left[\begin{array}{c} 1 \\ -1 \end{array} \right],$$

so $B_S[i] = \pm 1$ and thus $\det(B_S[i]) = 1$.

<u>Inductive case:</u> Suppose the lemma statement is true for graphs of size n-1. Let $j \neq i$ be a leaf of the tree, and take (k,j) to be an edge incident to j. We exchange rows and

columns so that (k, j) is last column, and j is last row; this may flip sign of determinant, but that doesn't matter. Then,

Thus if we expand the determinant along the last row we get

$$|\det(B_S[i])| = |\det(B_{S-\{(k,j)\}}[i,j])| = 1.$$

The last equality follows by induction, since $S - \{(k, j)\}$ is a tree on the vertex set without j and we assumed that j is a leaf.

2 Consequences of the Matrix-Tree Theorem

Once we have the matrix-tree theorem, there are a number of interesting consequences, which we explore in this section. Given a square matrix $A \in \mathbb{R}^{n \times n}$, let A_{ij} be matrix without row i column j (so $A[i] = A_{ii}$). Consider the cofactor matrix $C = (c_{ij})$, with entry $c_{ij} = (-1)^{i+j} \det(A_{ij})$ defining the i, j cofactor of A. Then we define the adjugate $\operatorname{adj}(A) = C^T$. Next, we will need the following fact.

Fact 3

$$A \operatorname{adj}(A) = \det(A)I.$$

By the matrix-tree theorem, the (i,i) cofactor of L_G is equal to $\tau(G)$. But we can say something even stronger.

Theorem 3 Every cofactor of L_G is $\tau(G)$, so that

$$\operatorname{adj}(L_G) = \tau(G)J.$$

Proof:

If G is not connected, then $\tau(G) = 0$ and $\lambda_2(L_G) = 0 = \lambda_1(L_G)$, so the rank of L_G is at most n-2. Then $\det((L_G)_{ij}) = 0$, which implies that $\operatorname{adj}(L_G) = 0$, as desired.

If G is connected, the fact above gives that $L_G \operatorname{adj}(L_G) = 0$ (i.e. the zero matrix) since $\det(L_G) = 0$. Because G is connected, multiples of e are the only eigenvectors of L_G with eigenvalue 0, and every column of $\operatorname{adj}(L_G)$ must therefore be some multiple of e. But we know that for the i^{th} column of $\operatorname{adj}(L_G)$, its i^{th} entry is $\tau(G)$, so the column itself must be $\tau(G)e$, and the lemma statement follows.

We conclude with one more theorem.

Theorem 4 Let $0 = \lambda_1 \le \lambda_2 \le ... \le \lambda_n$ be the eigenvalues of L_G . Then

$$\tau(G) = \frac{1}{n} \prod_{i=2}^{n} \lambda_i.$$

Proof:

If G is not connected, then $\lambda_2 = 0$ and $\tau(G) = 0$, so the theorem holds. Otherwise, we will look at linear term of the characteristic polynomial in two different ways. From one perspective, the characteristic polynomial is

$$(\lambda - \lambda_1)(\lambda - \lambda_2)...(\lambda - \lambda_n) = \lambda(\lambda - \lambda_2)(\lambda - \lambda_3)...(\lambda - \lambda_n),$$

so the linear term is

$$(-1)^{n-1} \prod_{i=2}^{n} \lambda_i.$$

On the other hand, we know that the characteristic polynomial is defined as $\det(\lambda I - L_G)$. We'll use that for A, B of the same dimensions, $\det(A + B) = \sum_{S \subseteq [n]} \det A_S$, where A_S is the matrix A with the rows indexed by S replaced by the corresponding rows of B. For $A = \lambda I$, $B = -L_G$, the linear term in λ is given by

$$\sum_{S\subseteq[n],|S|=n-1} \det(\lambda I)_S = (-1)^{n-1} \sum_{i=1}^n \det(L_G[i]) = (-1)^{n-1} \cdot n \cdot \tau(G).$$

Therefore, $\tau(G) = \frac{1}{n} \prod_{i=1}^{n} \lambda_i$.