

A construction of cospectral graphs for the normalized Laplacian

Steve Butler*

Department of Mathematics
Iowa State University
Ames, IA 50011, U.S.A.
butler@iastate.edu

Jason Grout

Department of Mathematics and Computer Science
Drake University
Des Moines, IA 50311, U.S.A.
jason.grout@drake.edu

Submitted: Aug 23, 2010; Accepted: Nov 21, 2011; Published: Dec 12, 2011
Mathematics Subject Classification: 05C50

Abstract

We give a method to construct cospectral graphs for the normalized Laplacian by a local modification in some graphs with special structure. Namely, under some simple assumptions, we can replace a small bipartite graph with a cospectral mate without changing the spectrum of the entire graph. We also consider a related result for swapping out biregular bipartite graphs for the matrix $A + tD$.

We produce (exponentially) large families of non-bipartite, non-regular graphs which are mutually cospectral, and also give an example of a graph which is cospectral with its complement but is not self-complementary.

Keywords: normalized Laplacian; cospectral; bipartite subgraph swapping

1 Introduction

Spectral graph theory examines relationships between the structure of a graph and the eigenvalues (or spectrum) of a matrix associated with that graph. Different matrices are able to give different information, but all the common matrices have limitations. This is because there are graphs which have the same spectrum for a certain matrix but different structure—such graphs are called cospectral with respect to that matrix.

The following are some of the matrices studied in spectral graph theory:

- The adjacency matrix A . This is defined by $A(u, v) = 1$ when u and v are adjacent and 0 otherwise. The spectrum of the adjacency matrix can determine the number

*This work was done with support of an NSF Mathematical Sciences Postdoctoral Fellowship while at UCLA.

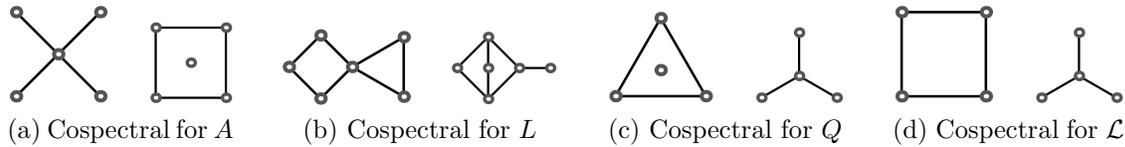


Figure 1: Examples of cospectral graphs for A , L , Q and \mathcal{L} .

of edges and if a graph is bipartite, but it cannot determine if a graph is connected (see [6] for more information). See Figure 1a, the “Saltire pair” [5, 15], for an example of two graphs which are cospectral with respect to the adjacency matrix but not any of the other matrices we will discuss.

- The Laplacian $L = D - A$. This matrix is also known as the combinatorial Laplacian and is found by taking the difference of the diagonal degree matrix D and the adjacency matrix. The spectrum of the Laplacian can determine the number of edges and the number of connected components, but it cannot determine if a graph is bipartite (see [6] for more information). See Figure 1b [10, 15], for an example of two graphs which are cospectral with respect to the Laplacian but not any of the other matrices we will discuss.
- The signless Laplacian $Q = D + A$. This matrix is found by taking the sum of the diagonal degree matrix D and the adjacency matrix. The spectrum of the signless Laplacian can determine the number of edges and the number of connected components which are bipartite, but it cannot determine if a graph is bipartite or connected (see [7, 8, 9] for more information). See Figure 1c [15], for an example of two graphs which are cospectral with respect to the signless Laplacian but not any of the other matrices we will discuss.
- The normalized Laplacian \mathcal{L} . This matrix is defined by $\mathcal{L} = D^{-1/2}LD^{-1/2}$, where by convention if we have an isolated vertex then it will contribute 0 to the spectrum. The spectrum of the normalized Laplacian is closely related to the spectrum of the probability transition matrix of a random walk. This spectrum can determine if a graph is bipartite and the number of connected components, but it cannot determine the number of edges (see [4] for more information). See Figure 1d for an example of two graphs which are cospectral with respect to the normalized Laplacian but not any of the other matrices we will discuss.

One way to understand what structure the spectrum of a matrix cannot identify is to study cospectral graphs. Cospectral graphs for the adjacency matrix [12, 13, 14, 15, 16] and the Laplacian matrix [15, 17, 20] have been studied, particularly for graphs with few vertices. Cospectral graphs for the signless Laplacian have been little studied beyond their enumeration and rules which apply to all of these matrices [15].

Little is also known about cospectral graphs for the normalized Laplacian compared to other matrices. Previously, the only cospectral graphs were bipartite (complete bipartite

graphs [4, 20] and bipartite graphs found by “unfolding” a small bipartite graph in two ways [2]) or were regular and cospectral for A (since a regular graph is cospectral for all of these four matrices if it is cospectral for any one of them).

In Table 1, we have listed the number of graphs with a cospectral mate for the normalized Laplacian matrix for graphs on nine or fewer vertices, counted using Sage (see Appendix A for example code). The number of cospectral graphs for the normalized Laplacian have also been counted by Wilson and Zhu [21], though they give percentages and not the count of how many have cospectral mates. We have also included similar counts for the other three matrices which come from [1, 15].

#vertices	#graphs	A	$L=D-A$	$Q=D+A$	\mathcal{L}
1	1	0	0	0	0
2	2	0	0	0	0
3	4	0	0	0	0
4	11	0	0	2	2
5	34	2	0	4	4
6	156	10	4	16	14
7	1044	110	130	102	52
8	12346	1722	1767	1201	201
9	274668	51039	42595	19001	1092

Table 1: Number of graphs with a cospectral mate for the various matrices

Given the relatively small number of graphs with a cospectral mate with respect to the normalized Laplacian, it is surprising that so little is known about forming cospectral graphs for that matrix. The problem is that some of the main tools that are used to form cospectral graphs for other matrices do not generalize to the normalized Laplacian. One such example is a technique known as switching, which is accomplished by replacing edges by non-edges and non-edges by edges between two subsets (see [14, 15, 18]; a simple example is shown in Figure 2). Given some basic assumptions, this is an easy way to construct cospectral graphs for the adjacency matrix. However, switching does not in general work for the normalized Laplacian, in particular, it will only be guaranteed to work when the degrees are unchanged (see [3]). So the graphs shown in Figure 2 are *not* cospectral with respect to the normalized Laplacian.



Figure 2: An example of two cospectral graphs for the adjacency matrix related by switching.

In Section 2 we will introduce a method to construct cospectral graphs for the normalized Laplacian. This construction will work similarly to switching in that we will make

a small local change to the graph by swapping in one bipartite graph with a cospectral mate and show the two graphs still share the same eigenvalues. In Section 3 we will show that if we add additional constraints to the bipartite graphs which are swapped then the resulting graphs are also cospectral with respect to A , L and Q . In Section 4 we will show how to use this construction to produce large families of graphs which are mutually cospectral. Finally, in Section 5 we will give some concluding remarks.

2 Swapping bipartite subgraphs

The method of finding cospectral graphs for the adjacency matrix by switching reduces to making a local change of the graph from one bipartite subgraph to its complement and showing that the spectrum is unchanged. We will consider something similar, namely a local change, but instead of replacing a bipartite subgraph with its complement we will swap out a bipartite subgraph for a cospectral mate.

For a subset W of the vertices V of G , we will let $G[W]$ be the induced subgraph of G on the vertex set W .

Theorem 1. *Let P_1 and P_2 be bipartite, cospectral graphs with respect to the normalized Laplacian on the vertex set $B \cup C$ such that all edges go between B and C and where all vertices in B have degree k for both graphs.*

Let G_1 be a graph on the vertices $A \cup A' \cup B \cup C$ where $G_1[A \cup A']$ is an arbitrary graph; $G_1[B]$ and $G_1[C]$ have no edges; there are no edges going between A and B , between A and C , or between A' and C ; $G_1[A' \cup B]$ is a complete bipartite graph; and $G_1[B \cup C] = P_1$. The graph G_2 is defined similarly except that $G_2[B \cup C] = P_2$. Then G_1 and G_2 are cospectral with respect to the normalized Laplacian.

If the dimension of the eigenspace associated with $\lambda = 1$ intersected with the subspace of vectors that are nonzero only on B is the same for both P_1 and P_2 , then the graphs H_1 and H_2 are also cospectral with respect to the normalized Laplacian, where H_1 only differs from G_1 in that $H_1[B]$ is the complete graph and similarly H_2 only differs from G_2 in that $H_2[B]$ is the complete graph.

An example of the construction described in Theorem 1 is shown in Figure 3. In this case P_1 and P_2 are the graphs $K_{1,1} \cup K_{1,6}$ and $K_{1,4} \cup K_{1,3}$ where we have put the degree 1 vertices into B . It is well known that the spectrum of a complete bipartite graph $K_{p,q}$ is $0^{[1]}, 1^{[p+q-2]}, 2^{[1]}$, where the exponent indicates multiplicity, hence these graphs are easily seen to be cospectral. Moreover it is easy to check that the dimension of the eigenspace for the eigenvalue 1 intersected with the subspace of vectors which are nonzero only on B is 5 for both graphs, i.e., we could also have put a complete graph on the vertices of B in the graphs shown in Figure 3 and still had a cospectral pair.

Additional examples of graphs P_1 and P_2 satisfying (both) the conditions on Theorem 1 are $P_1 = K_{\ell,p} \cup K_{\ell,q}$ and $P_2 = K_{\ell,r} \cup K_{\ell,s}$ where the vertices of degree ℓ are all placed in B and $p + q = r + s$. Two further examples are shown in Figures 4 and 5.

To prove Theorem 1 we will find it convenient to work with the harmonic eigenvectors of the normalized Laplacian. Namely, if $\mathbf{x} \neq \mathbf{0}$ is an eigenvector associated with the

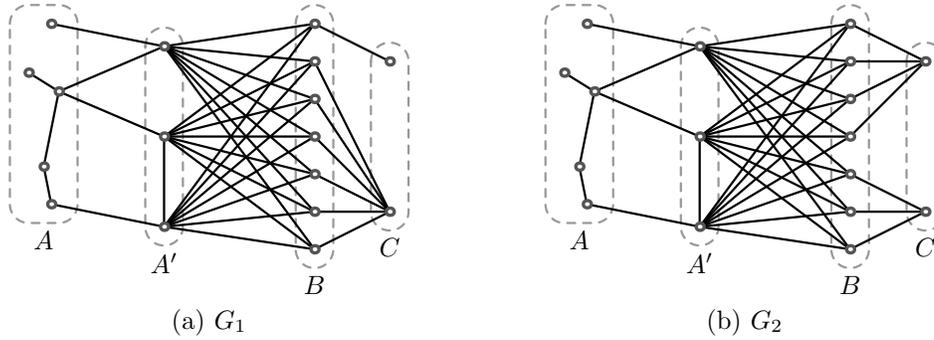


Figure 3: An example of cospectral graphs using the construction given in Theorem 1.



Figure 4: A pair of cospectral graphs satisfying the conditions of Theorem 1.

eigenvalue λ , i.e., $\mathcal{L}\mathbf{x} = \lambda\mathbf{x}$, then the corresponding harmonic eigenvector is $\mathbf{y} = D^{-1/2}\mathbf{x}$. This translates the relationship $\mathcal{L}\mathbf{x} = \lambda\mathbf{x}$ into $(D - A)\mathbf{y} = \lambda D\mathbf{y}$, which at a vertex v becomes

$$\sum_{u \sim v} \mathbf{y}(u) = (1 - \lambda)\mathbf{y}(v)d(v), \quad (1)$$

where $d(v)$ is the degree of the vertex v . We will say that two harmonic eigenvectors \mathbf{y}_1 and \mathbf{y}_2 are orthogonal if $\mathbf{y}_2^* D \mathbf{y}_1 = 0$, i.e., if the corresponding eigenvectors are orthogonal.

Before we begin the proof of Theorem 1 it will be useful to make some observations about bipartite graphs that will come up in the proof.



Figure 5: A pair of cospectral graphs satisfying the conditions of Theorem 1.

Lemma 2. Let P be a bipartite graph on the vertices $B \cup C$ where all edges go between B and C and the vertices in B all have degree k . Further, if \mathbf{x} is an eigenvector for the normalized Laplacian then we can write it as $\mathbf{x} = \mathbf{b} + \mathbf{c}$ where \mathbf{b} is the vector \mathbf{x} restricted to the vertices of B , and similarly for \mathbf{c} . Then the following hold:

- (a) If $\mathcal{L}\mathbf{x} = \mathbf{x}$, i.e., \mathbf{x} is an eigenvector for the eigenvalue 1, then $\mathcal{L}\mathbf{b} = \mathbf{b}$ and $\mathcal{L}\mathbf{c} = \mathbf{c}$. In other words, we can divide the eigenspace for the eigenvalue 1 between those which are nonzero only on the vertices of B and nonzero only on the vertices of C .
- (b) If $\mathbf{x} = \mathbf{b} + \mathbf{c}$ is an eigenvector associated with λ , then $\mathbf{b} - \mathbf{c}$ is an eigenvector associated with $2 - \lambda$.
- (c) If $\mathbf{x}_1 = \mathbf{1}_B + \mathbf{1}_C, \dots, \mathbf{x}_i = \mathbf{b}_i + \mathbf{c}_i, \dots, \mathbf{x}_m = \mathbf{1}_B - \mathbf{1}_C$ are a set of orthogonal harmonic eigenvectors of P , then for $1 < i < m$ we have \mathbf{b}_i is orthogonal to $\mathbf{1}_B$ and \mathbf{c}_i is orthogonal to $\mathbf{1}_C$. In particular, for $1 < i < m$ the sum of the entries of \mathbf{b}_i is 0.

Parts (a) and (b) easily follow from (1). For (c) we note that \mathbf{x}_i will be orthogonal to the vectors $\frac{1}{2}(\mathbf{x}_1 + \mathbf{x}_m) = \mathbf{1}_B$ and $\frac{1}{2}(\mathbf{x}_1 - \mathbf{x}_m) = \mathbf{1}_C$, also since the vertices in B are regular the result on the sum of the entries of \mathbf{b}_i follows by the definition of orthogonality.

Proof of Theorem 1. Let $\mathbf{x}_1 = \mathbf{1}_B + \mathbf{1}_C, \dots, \mathbf{x}_i = \mathbf{b}_i + \mathbf{c}_i, \dots, \mathbf{x}_m = \mathbf{1}_B - \mathbf{1}_C$ be a full set of orthogonal harmonic eigenvectors of P_1 associated with $\lambda_1 = 0, \dots, \lambda_i, \dots, \lambda_m = 2$ respectively. Let s denote the degree of a vertex in B for the graph G_1 and k the degree in P_1 . Then for $1 < i < m$ we have that $\widehat{\mathbf{x}}_i = \mathbf{b}_i + \sqrt{\frac{s}{k}}\mathbf{c}_i$ is a harmonic eigenvector for G_1 associated with eigenvalue $\gamma_i = 1 - (1 - \lambda_i)\sqrt{\frac{k}{s}}$. To see this we need to examine what happens for each vertex v in G_1 using (1).

- For $v \in A$: Then all of the entries of the vertex and its neighbors are all 0 and so (1) trivially holds.
- For $v \in A'$: All of the neighbors in A and A' are 0 while the sum of the entries in B will be 0 by Lemma 2(c), so both sides again are 0 and (1) holds.
- For $v \in B$: We note that all of the nonzero elements of $\widehat{\mathbf{x}}_i$ adjacent to v are in C , and further we have that for the harmonic eigenvector \mathbf{x}_i for P_1 that

$$\sum_{u \sim v} \mathbf{x}_i(u) = (1 - \lambda_i)\mathbf{x}_i(v)k.$$

So we have for G_1 that

$$\begin{aligned} \sum_{u \sim v} \widehat{\mathbf{x}}_i(u) &= \sqrt{\frac{s}{k}} \sum_{u \sim v} \mathbf{x}_i(u) = \sqrt{\frac{s}{k}}(1 - \lambda_i)\mathbf{x}_i(v)k = \sqrt{\frac{k}{s}}(1 - \lambda_i)\widehat{\mathbf{x}}_i(v)d(v) \\ &= (1 - \gamma_i)\widehat{\mathbf{x}}_i(v)d(v), \end{aligned}$$

so (1) again holds.

- For $v \in C$: We proceed similarly as we did for vertices in B and we get

$$\sum_{u \sim v} \widehat{\mathbf{x}}_i(u) = \sum_{u \sim v} \mathbf{x}_i(u) = (1 - \lambda_i) \mathbf{x}_i(v) d(v) = \sqrt{\frac{k}{s}} (1 - \lambda_i) \widehat{\mathbf{x}}_i(v) d(v) = (1 - \gamma_i) \widehat{\mathbf{x}}_i(v) d(v).$$

Next we note by Lemma 2(c) that $\mathbf{1}_B$ and $\mathbf{1}_C$ are orthogonal to each of the vectors $\widehat{\mathbf{x}}_i$ for $1 < i < m$. On P_1 , the dimension of the space orthogonal to the $\widehat{\mathbf{x}}_i$ for $1 < i < m$ restricted to P_1 is two. A basis for the orthogonal complement of $\text{span}\{\widehat{\mathbf{x}}_i\}$ is $\{\mathbf{1}_B, \mathbf{1}_C\}$, so any harmonic vector orthogonal to all of these *must* be a linear combination of $\mathbf{1}_B$ and $\mathbf{1}_C$. In particular, if we let \mathbf{y} be a harmonic eigenvector of G_1 that is orthogonal to all of the $\widehat{\mathbf{x}}_i$ for $1 < i < m$, then \mathbf{y} restricted to B and C is orthogonal to the $\widehat{\mathbf{x}}_i$, so we have $\mathbf{y}|_{B \cup C} = b\mathbf{1}_B + c\mathbf{1}_C$ for some constants b and c . In other words, the harmonic eigenvector \mathbf{y} is constant on the vertices of B and constant on the vertices of C .

Everything that we have done for G_1 carries over to G_2 . In particular, since P_1 and P_2 are cospectral, then the γ_i found by generalizing the harmonic eigenvectors of P_1 and P_2 remain the same. Furthermore, every other harmonic eigenvector orthogonal to the $\widehat{\mathbf{x}}_i$ must be constant on the vertices of B and the vertices of C . To finish off the first claim of the theorem, we now only need to observe that *any* harmonic eigenvector which is orthogonal to all of the $\widehat{\mathbf{x}}_i$ in G_1 is also a harmonic eigenvector for G_2 for the same eigenvalue. Again to see this we only need to consider what happens for each vertex.

- For $v \in A \cup A'$: Then all of the entries of the vertex and its neighbors are the same for both graphs, so (1) trivially holds.
- For $v \in B$: All of the neighbors in A' are the same, and we have the exact same number of neighbors in C as before with the same value and so again (1) holds.
- For $v \in C$: For the graph G_1 , let β be the fixed value of the vertices in B . We have

$$\sum_{u \sim v} \mathbf{y}(u) = d(v)\beta = (1 - \lambda)\mathbf{y}(v)d(v).$$

Of course the $d(v)$ terms cancel and we are left with $\beta = (1 - \lambda)\mathbf{y}(v)$. So even though the degree of the vertex in C might change, it will have no effect on this relationship and so (1) again holds.

In summary, we were able to find $m - 2$ harmonic eigenvectors for G_1 and G_2 that gave the same set of eigenvalues. For any other harmonic eigenvector orthogonal to these, the same harmonic eigenvector worked for both graphs and so the remaining set of eigenvalues also agreed. So we can conclude that G_1 and G_2 are cospectral with respect to the normalized Laplacian.

We now turn to the second statement of the theorem. Let s be the degree of a vertex in B in the graph H_1 . So let $\mathbf{x}_1 = \mathbf{1}_B + \mathbf{1}_C, \dots, \mathbf{x}_i = \mathbf{b}_i + \mathbf{c}_i, \dots, \mathbf{x}_m = \mathbf{1}_B - \mathbf{1}_C$ be a full set of orthogonal harmonic eigenvectors of P_1 associated with $\lambda_1 = 0, \dots, \lambda_i, \dots, \lambda_m = 2$

respectively. First, we consider the harmonic eigenvectors of the graph P_1 in the eigenspace corresponding to $\lambda = 1$. By Lemma 2(a) we can assume that the vectors \mathbf{x}_i are either of the form \mathbf{b}_i or \mathbf{c}_i . Let the corresponding harmonic eigenvector of H_1 be $\widehat{\mathbf{x}}_i = \mathbf{x}_i$, i.e., we simply expand \mathbf{x}_i to be zero outside of P_1 . Then we have two cases.

- If $\mathbf{x}_i = \mathbf{b}_i$: In this case, we know there are no problems for the vertices in A (every term in (1) is 0) or C (since (1) reduces to what was done in P_1). For the vertices in A' we can use Lemma 2(c) to see that both sides of (1) are 0. Finally, suppose that v is a vertex in B . Then we have

$$\sum_{u \sim v} \widehat{\mathbf{x}}_i(u) = \sum_{u \in B} \widehat{\mathbf{x}}_i(u) - \widehat{\mathbf{x}}_i(v) = -\widehat{\mathbf{x}}_i(v) = -\frac{1}{s}\widehat{\mathbf{x}}_i(v)s = -\frac{1}{s}\widehat{\mathbf{x}}_i(v)d(v).$$

In particular, we have that $1 - \gamma_i = -\frac{1}{s}$ is an eigenvalue, i.e., that $\gamma_i = 1 + \frac{1}{s}$ is an eigenvalue, for this harmonic eigenvector.

- If $\mathbf{x}_i = \mathbf{c}_i$: In this case we similarly know there are no problems for vertices in A , A' and B . For vertices in C , (1) reduces to what we had in P_1 and so we can conclude that this is a harmonic eigenvector for the eigenvalue $\gamma_i = 1$.

Now suppose that $\mathbf{x}_i = \mathbf{b}_i + \mathbf{c}_i$ is associated with $\lambda_i \neq 1$ for P_1 . Then we now will create two harmonic eigenvectors for H_1 , namely

$$\begin{aligned} \mathbf{y}_i^1 &= \mathbf{b}_i + t_1 \mathbf{c}_i = \mathbf{b}_i + \left(\frac{1 + \sqrt{1 + 4(1 - \lambda_i)^2 s k}}{2k(1 - \lambda_i)} \right) \mathbf{c}_i, \\ \mathbf{y}_i^2 &= \mathbf{b}_i + t_2 \mathbf{c}_i = \mathbf{b}_i + \left(\frac{1 - \sqrt{1 + 4(1 - \lambda_i)^2 s k}}{2k(1 - \lambda_i)} \right) \mathbf{c}_i. \end{aligned}$$

These will be associated with the eigenvalues of

$$\gamma_1 = \frac{2s + 1 - \sqrt{1 + 4(1 - \lambda_i)^2 s k}}{2s} \quad \text{and} \quad \gamma_2 = \frac{2s + 1 + \sqrt{1 + 4(1 - \lambda_i)^2 s k}}{2s},$$

respectively, for H_1 . Some simple computations show that the following relationships hold:

$$t_1(1 - \gamma_1) = t_2(1 - \gamma_2) = 1 - \lambda_i, \quad t_1(1 - \lambda_i)k - 1 = (1 - \gamma_1)s, \quad \text{and} \quad t_2(1 - \lambda_i)k - 1 = (1 - \gamma_2)s.$$

(Before we proceed to the next step of showing that these are indeed harmonic eigenvectors and eigenvalues we first make the observation that we are not creating more harmonic eigenvectors than we had before. This is because in Lemma 2(b) we can pair up harmonic eigenvectors and eigenvalues. In particular, we would have generated the same new harmonic eigenvectors and eigenvalues if we had used $2 - \lambda_i$ and $\mathbf{b}_i - \mathbf{c}_i$. So really we are taking pairs of eigenvectors and eigenvalues to new pairs of eigenvectors and eigenvalues.)

To verify that these are harmonic eigenvectors with the specified eigenvalues we need to examine what happens for each vertex v in H_1 using (1). We will step through \mathbf{y}_i^1 and γ_1 , the arguments for \mathbf{y}_i^2 and γ_2 are the same.

- For $v \in A$: Then all of the entries of the vertex and its neighbors are all 0 and so (1) trivially holds.
- For $v \in A'$: All of the neighbors in A and A' are 0 while the sum of the entries in B will be 0 by Lemma 2(c), so both sides again are 0 and (1) holds.
- For $v \in B$: We note that all of the nonzero elements of $\widehat{\mathbf{x}}_i$ adjacent to v are in B and C , and further we have that for the harmonic eigenvector \mathbf{x}_i for P_1 that

$$\sum_{\substack{u \sim v \\ u \in C}} \mathbf{x}_i(u) = (1 - \lambda_i)\mathbf{x}_i(v)k.$$

So we have for H_1 that

$$\begin{aligned} \sum_{u \sim v} \mathbf{y}_i^1(u) &= t_1 \sum_{\substack{u \sim v \\ u \in C}} \mathbf{x}_i(u) + \sum_{u \in B} \mathbf{x}_i(u) - \mathbf{x}_i(v) = t_1(1 - \lambda_i)\mathbf{x}_i(v)k - \mathbf{x}_i(v) \\ &= (t_1(1 - \lambda_i)k - 1)\mathbf{x}_i(v) = (1 - \gamma_1)\mathbf{y}_i^1(v)s = (1 - \gamma_1)\mathbf{y}_i^1(v)d(v). \end{aligned}$$

so (1) again holds.

- For $v \in C$: We proceed similarly as we did for vertices in B and we get

$$\sum_{u \sim v} \mathbf{y}_i^1(u) = \sum_{u \sim v} \mathbf{x}_i(u) = (1 - \lambda_i)\mathbf{x}_i(v)d(v) = t_1(1 - \gamma_1)\mathbf{x}_i(v)d(v) = (1 - \gamma_1)\mathbf{y}_i^1(v)d(v).$$

The remainder of this case now proceeds as before. Namely, everything that we did for H_1 carries over for H_2 (counting multiplicity of eigenvalues). Further, any other harmonic eigenvector orthogonal to the ones given must be constant on B and C and so any other harmonic eigenvector which works for H_1 also works for H_2 . Therefore we can conclude that they have the same set of eigenvalues, i.e., are cospectral with respect to the normalized Laplacian. \square

3 Swapping biregular bipartite subgraphs

The graphs generated by Theorem 1 will generally not give cospectral graphs for other matrices. So it is instructive to examine the proof and try to understand the point at which the fact that we were using the normalized Laplacian as compared to some other matrix came into play. The key is understanding how (1) remains true even when the degrees of vertices in C change—if the degree of a vertex in C changes, the $d(v)$ term on the right side of (1) will change proportionally to the sum on the left so that the equality still holds.

In the case when the degrees of vertices do not change, for example, when P_1 and P_2 are both (k, ℓ) -biregular, then the proof generalizes. Recall that a bipartite graph is (k, ℓ) -biregular if the vertices can be partitioned into $B \cup C$ where all edges go between B and C , the vertices in B all have degree k , and the vertices in C have degree ℓ . We now have the following theorem.

Theorem 3. *Let P_1 and P_2 be bipartite, cospectral with respect to the adjacency matrix, (k, ℓ) -biregular graphs on the vertex set $B \cup C$ such that all vertices in B have degree k and all vertices in C have degree ℓ and edges go between B and C . Further, the dimension of the eigenspace associated with $\lambda = 0$ for the adjacency matrix intersected with the subspace of vectors which are nonzero only on B is the same for both P_1 and P_2 .*

Let G_1 be a graph on the vertices $A \cup A' \cup B \cup C$ where $G_1[A \cup A']$ is an arbitrary graph; $G_1[B]$ and $G_1[C]$ have no edges; there are no edges going between A and B , between A and C , or between A' and C ; $G_1[A' \cup B]$ is a complete bipartite graph; and $G_1[B \cup C] = P_1$. The graph G_2 is defined similarly except that $G_2[B \cup C] = P_2$. Then G_1 and G_2 are cospectral with respect to the matrix $A + tD$ for t arbitrary.

Similarly, the graphs H_1 and H_2 are also cospectral with respect to $A + tD$ for t arbitrary, where H_1 only differs from G_1 in that $H_1[B]$ is the complete graph and similarly H_2 only differs from G_2 in that $H_2[B]$ is the complete graph.

Note when $t = 0$ then $A + tD = A$, when $t = 1$ then $A + tD = A + D = Q$, and when $t = -1$ then $A + tD = A - D = -L$, so that any such pairs of graphs are cospectral with respect to all of the matrices A , L , and Q . Further, these are also cospectral for \mathcal{L} since if $t = -\lambda - 1$, we have

$$\begin{aligned} \det(A + (-\lambda - 1)D) &= \det((A - D) - \lambda D) \\ &= \det(D^{1/2}(\mathcal{L} - \lambda I)D^{1/2}) \\ &= \left(\prod d_i\right) \det(\mathcal{L} - \lambda I). \end{aligned}$$

Since the two graphs have the same degree sequence, the product $\prod d_i$ is the same, so the characteristic polynomials also are the same.

In order to use this theorem, we must find two (k, ℓ) -biregular bipartite graphs which are cospectral and for which the dimensions of the eigenspace of 0 restricted to B agree on both graphs. There is one special case for which this is much easier, namely the theorem does not prohibit the possibility that P_1 and P_2 are the same graph (so trivially are cospectral so we only are reduced to checking the dimension of the eigenspaces). This is not so trivial as it might sound since we are distinguishing the two parts of the bipartite graph in the proof of the theorem, so that while P_1 and P_2 are the same graph we are attaching them in two different ways and so the resulting graphs might not be isomorphic.

As an example of this consider the graphs shown in Figure 6. These are both the same graph (just flipped) and a simple check shows that the dimension of the eigenspace associated with 0 and restricted to B is 2 in both graphs. Therefore Theorem 3 applies, and any two graphs to which these two graphs are attached are cospectral. But note that in P_2 there are two vertices in C with the same neighbors (marked), but there are no such pairs in P_1 . So the resulting graphs are *not* isomorphic as long as $A' \neq \emptyset$.

So for example if we consider the graphs H_1 and H_2 where $A \cup A' = \emptyset$ and we induce a complete graph on the vertices of B , then Theorem 3 shows that the resulting graphs are cospectral, and again since there are two vertices with degree 3 sharing common neighbors in one graph but not the other they are non-isomorphic. These graphs are shown in



Figure 6: A pair of cospectral graphs satisfying the conditions of Theorem 3.

Figure 7. These graphs also happen to be complements of one another, and so give an example of a graph which is cospectral with its complement but not self-complementary. (Note the graph shown in Figure 1c is another example of a graph which is cospectral with its complement, but in that case it is only cospectral with respect to Q and not with respect to A , L or \mathcal{L} .)



Figure 7: A pair of non-isomorphic cospectral graphs for A , L , Q and \mathcal{L} . (Note $\overline{H_1} = H_2$.)

The proof for Theorem 3 will proceed similar to the proof for Theorem 1, so we will skip some of the routine computations and provide an outline of the proof. Note that if \mathbf{x} is an eigenvector associated with eigenvalue λ , i.e., $(A + tD)\mathbf{x} = \lambda\mathbf{x}$, then at a vertex v we have

$$\sum_{u \sim v} \mathbf{x}(u) = (\lambda - td(v))\mathbf{x}(v). \tag{2}$$

We first start with some simple properties of bipartite graphs that will be helpful in the proof. The proofs are similar to Lemma 2 and we will omit them.

Lemma 4. *Let P be a bipartite (k, ℓ) -biregular graph on the vertices $B \cup C$ where all edges go between B and C . Further, if \mathbf{x} is an eigenvector for the adjacency matrix, then we can write it as $\mathbf{x} = \mathbf{b} + \mathbf{c}$ where \mathbf{b} is the vector \mathbf{x} restricted to the vertices of B , and similarly for \mathbf{c} . Then the following hold:*

- (a) *If $A\mathbf{x} = \mathbf{0}$, i.e., \mathbf{x} is an eigenvector for the eigenvalue 0, then $A\mathbf{b} = \mathbf{0}$ and $A\mathbf{c} = \mathbf{0}$. In other words, we can divide the eigenspace for the eigenvalue 0 between those which are nonzero only on the vertices of B and nonzero only on the vertices of C .*

- (b) If $\mathbf{x} = \mathbf{b} + \mathbf{c}$ is an eigenvector associated with λ , then $\mathbf{b} - \mathbf{c}$ is an eigenvector associated with $-\lambda$.
- (c) If $\mathbf{x}_1 = \sqrt{k}\mathbf{1}_B + \sqrt{\ell}\mathbf{1}_C, \dots, \mathbf{x}_i = \mathbf{b}_i + \mathbf{c}_i, \dots, \mathbf{x}_m = \sqrt{k}\mathbf{1}_B - \sqrt{\ell}\mathbf{1}_C$ are a set of orthogonal eigenvectors of P , then for $1 < i < m$ we have \mathbf{b}_i is orthogonal to $\mathbf{1}_B$ and \mathbf{c}_i is orthogonal to $\mathbf{1}_C$. In particular, for $1 < i < m$ the sum of the entries of \mathbf{b}_i is 0.

Proof of Theorem 3. We will first consider the case when the vertices on B induce an empty graph.

Let $\mathbf{x}_1 = \sqrt{k}\mathbf{1}_B + \sqrt{\ell}\mathbf{1}_C, \dots, \mathbf{x}_i = \mathbf{b}_i + \mathbf{c}_i, \dots, \mathbf{x}_m = \sqrt{k}\mathbf{1}_B - \sqrt{\ell}\mathbf{1}_C$ be a full set of orthogonal eigenvectors of P_1 for the adjacency matrix associated with the eigenvalues $\lambda_1 = \sqrt{k\ell}, \dots, \lambda_i, \dots, -\sqrt{k\ell}$ respectively. Let s denote the degree of a vertex in B in G_1 , and note by construction that ℓ will be the degree of a vertex in C in G_1 . We now show how to extend each of these eigenvectors to an eigenvector for G_1 . We have the following:

- For an eigenvector of the form $\mathbf{x} = \mathbf{b}$ (i.e., only nonzero on the vertices of B) and $\lambda = 0$ then $\mathbf{y} = \mathbf{b}$ is an eigenvector for $A + tD$ on G_1 associated with the eigenvalue of $\gamma = ts$.
- For an eigenvector of the form $\mathbf{x} = \mathbf{c}$ (i.e., only nonzero on the vertices of C) and $\lambda = 0$ then $\mathbf{y} = \mathbf{c}$ is an eigenvector for $A + tD$ on G_1 associated with the eigenvalue of $\gamma = t\ell$.
- For an eigenvector $\mathbf{x} = \mathbf{b} + \mathbf{c}$ associated with an eigenvalue $\lambda \neq 0$, we can construct two new eigenvectors, namely

$$\begin{aligned} \mathbf{y}_1 &= 2\lambda\mathbf{b} + \left(-t(s - \ell) + \sqrt{t^2(s - \ell)^2 + 4\lambda^2}\right)\mathbf{c} \quad \text{and} \\ \mathbf{y}_2 &= 2\lambda\mathbf{b} + \left(-t(s - \ell) - \sqrt{t^2(s - \ell)^2 + 4\lambda^2}\right)\mathbf{c}. \end{aligned}$$

These are associated with the eigenvalues of $A + tD$ for G_1

$$\gamma_1 = \frac{t(s + \ell) + \sqrt{t^2(s - \ell)^2 + 4\lambda^2}}{2} \quad \text{and} \quad \gamma_2 = \frac{t(s + \ell) - \sqrt{t^2(s - \ell)^2 + 4\lambda^2}}{2}$$

respectively.

(Note, as in the proof of Theorem 1, that we are not creating too many new eigenvectors and eigenvalues, i.e., by Lemma 4(b) we can pair up the eigenvalues and eigenvectors. In particular, $\mathbf{x} = \mathbf{b} + \mathbf{c}$ and λ will create the same new eigenvectors and eigenvalues as $\mathbf{x} = \mathbf{b} - \mathbf{c}$ and $-\lambda$.)

The verification that each of these is an eigenvector/eigenvalue pair for G_1 reduces to verifying (2) for each vertex, similarly as was done in Theorem 1. We will skip these routine computations.

Next we note by Lemma 4(c) that $\mathbf{1}_B$ and $\mathbf{1}_C$ are orthogonal to each of the new eigenvectors for $1 < i < m$. Since the dimension of the space orthogonal to the new

eigenvectors restricted to P_1 is two, any vector orthogonal to all of these new vectors *must* be a linear combination of $\mathbf{1}_B$ and $\mathbf{1}_C$ when restricted to P_1 . In particular, if we let \mathbf{y} be an eigenvector of G_1 that is orthogonal to all of the new eigenvectors, then \mathbf{y} restricted to B and C must be of the form $\mathbf{y}|_{B \cup C} = b\mathbf{1}_B + c\mathbf{1}_C$ for some constants b and c . In other words, the eigenvector \mathbf{y} is constant on the vertices of B and constant on the vertices of C .

Everything that we have done for G_1 carries over to G_2 . In particular, since P_1 and P_2 are cospectral, then the newly found γ_i are the same for both graphs, and further, every other eigenvector orthogonal to the ones found must be constant on the vertices of B and the vertices of C . To finish off the first case of the theorem, we now only need to observe that *any* eigenvector which is orthogonal to all of the new eigenvectors in G_1 is also an eigenvector for G_2 for the same eigenvalue. This again is done in the exact same way as in Theorem 1 and we will skip the computations here.

In summary, we were able to find $m - 2$ eigenvectors of $A + tD$ for G_1 and G_2 that gave the same set of eigenvalues. For any other eigenvector orthogonal to these, the same eigenvector worked for both graphs and so the remaining set of eigenvalues also agreed. So we can conclude that G_1 and G_2 are cospectral with respect to the matrix $A + tD$.

Now we turn to the case when the vertices on B induce a complete graph.

Using the same notation as in the previous case, we will extend the eigenvectors of P_1 for the adjacency matrix to eigenvectors of H_1 for $A + tD$. We have the following:

- For an eigenvector of the form $\mathbf{x} = \mathbf{b}$ (i.e., only nonzero on the vertices of B) and $\lambda = 0$ then $\mathbf{y} = \mathbf{b}$ is an eigenvector for $A + tD$ on H_1 associated with the eigenvalue of $\gamma = ts - 1$.
- For an eigenvector of the form $\mathbf{x} = \mathbf{c}$ (i.e., only nonzero on the vertices of C) and $\lambda = 0$ then $\mathbf{y} = \mathbf{c}$ is an eigenvector for $A + tD$ on H_1 associated with the eigenvalue of $\gamma = t\ell$.
- For an eigenvector $\mathbf{x} = \mathbf{b} + \mathbf{c}$ associated with an eigenvalue $\lambda \neq 0$, we can construct two new eigenvectors, namely

$$\mathbf{y}_1 = 2\lambda\mathbf{b} + (1 - t(s - \ell) + \sqrt{(1 - t(s - \ell))^2 + 4\lambda^2})\mathbf{c} \quad \text{and}$$

$$\mathbf{y}_2 = 2\lambda\mathbf{b} + (1 - t(s - \ell) - \sqrt{(1 - t(s - \ell))^2 + 4\lambda^2})\mathbf{c}.$$

These are associated with the eigenvalues of $A + tD$ for H_1

$$\gamma_1 = \frac{t(s + \ell) - 1 + \sqrt{(1 - t(s - \ell))^2 + 4\lambda^2}}{2} \quad \text{and}$$

$$\gamma_2 = \frac{t(s + \ell) - 1 - \sqrt{(1 - t(s - \ell))^2 + 4\lambda^2}}{2}$$

respectively.

(Note, as in the proof of Theorem 1, that we are not creating too many new eigenvectors and eigenvalues, i.e., by Lemma 4(b) we can pair up the eigenvalues and eigenvectors. In particular, $\mathbf{x} = \mathbf{b} + \mathbf{c}$ and λ will create the same new eigenvectors and eigenvalues as $\mathbf{x} = \mathbf{b} - \mathbf{c}$ and $-\lambda$.)

The verification that each of these is an eigenvector/eigenvalue pair for H_1 reduces to verifying (2) for each vertex, similarly as was done in Theorem 1. We will skip these routine computations.

The remainder of this case now proceeds as before. Namely, everything that we did for H_1 carries over for H_2 (counting multiplicity of eigenvalues). Further, any other eigenvector orthogonal to the ones given must be constant on B and C , so any other eigenvector for $A + tD$ which works for H_1 also works for H_2 . Therefore we can conclude that they have the same set of eigenvalues, i.e., are cospectral with respect to $A + tD$. \square

4 Large families of mutually cospectral graphs

We can use Theorem 1 to construct large families of mutually cospectral graphs for the normalized Laplacian that need not be bipartite nor have the same degree sequences.

Example 1. Let n, k be positive integers such that $m_1 + \dots + m_k = n$ is a partition of n into k positive integer parts. Let $FB(m_1, m_2, \dots, m_k)$ be the graph on $n + k$ vertices $b_1, \dots, b_n, v_1, \dots, v_k$, where the b_i induce a complete graph, each v_i is only adjacent to exactly m_i vertices from b_1, \dots, b_n , and each b_i is adjacent to exactly one of the v_i . The collection of all $FB(m_1, m_2, \dots, m_k)$ for partitions $m_1 + m_2 + \dots + m_k = n$ give a family of cospectral graphs for the normalized Laplacian, i.e., the spectrum is completely determined for the normalized Laplacian by n and k .

We have dubbed these graphs “fuzzy balls” since they consist of a complete graph (the b_i) with some extra vertices attached (the fuzz). In Figure 8, we have shown the cospectral family that corresponds to $n = 8$ and $k = 3$.

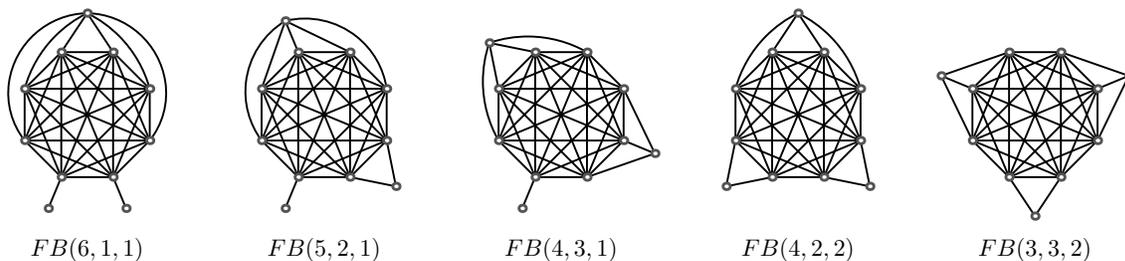


Figure 8: Fuzzy ball graphs for $n = 8$ and $k = 3$.

To see that these are all cospectral, we observe by Theorem 1 that $FB(m_1, m_2, \dots, m_k)$ is cospectral with $FB(m_1 + m_i - 1, m_2, \dots, m_{i-1}, 1, m_{i+1}, \dots, m_k)$ by using the pair of cospectral graphs $K_{1, m_1} \cup K_{1, m_i}$ and $K_{1, m_1 + m_i - 1} \cup K_{1, 1}$ where B induces a complete graph.

Therefore, applying this idea $k - 1$ times, we can conclude that for each partition of n into k parts we have $FB(m_1, m_2, \dots, m_k)$ is cospectral with $FB(n - k + 1, 1, \dots, 1)$. In particular, they are all mutually cospectral.

Example 2. Let n, k be positive integers such that $m_1 + \dots + m_k = n$ is a partition of n into k positive integer parts. Let $IS(m_1, m_2, \dots, m_k)$ be the graph on $n + k + 1$ vertices $a, b_1, \dots, b_n, v_1, \dots, v_k$, where the b_i and a induce a star graph with a as the central vertex, each v_i is only adjacent to exactly m_i vertices from b_1, \dots, b_n , and each b_i is adjacent to exactly one of the v_i . The collection of all $IS(m_1, m_2, \dots, m_k)$ for partitions $m_1 + m_2 + \dots + m_k = n$ give a family of cospectral graphs, i.e., the spectrum is completely determined by n and k .

We have dubbed these graphs “inflated stars”. In Figure 9, we have shown the cospectral family that corresponds to $n = 8$ and $k = 3$.

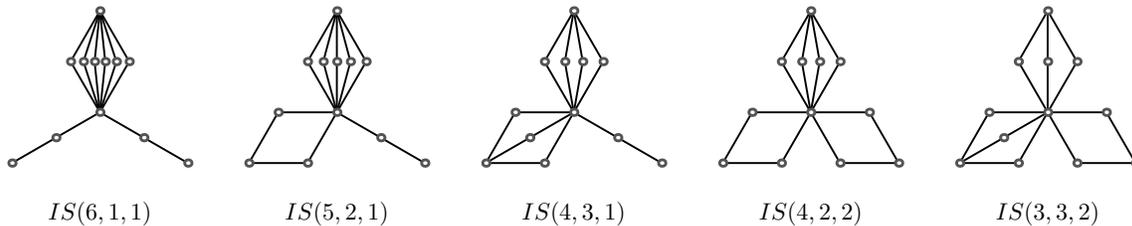


Figure 9: Inflated star graphs for $n = 8$ and $k = 3$.

Again, to see that these are all cospectral, we observe by Theorem 1 that the graph $IS(m_1, m_2, \dots, m_k)$ is cospectral with $IS(m_1 + m_i - 1, m_2, \dots, m_{i-1}, 1, m_{i+1}, \dots, m_k)$. Therefore, applying this idea $k - 1$ times, we can conclude that for each partition of n into k parts, we have $IS(m_1, m_2, \dots, m_k)$ is cospectral with $IS(n - k + 1, 1, \dots, 1)$. In particular, they are all mutually cospectral.

In both of these examples, we are using partitions to form large families. The number of partitions grows subexponentially with n , however it is possible to use Theorems 1 or 3 to construct large mutually cospectral families that grow exponentially with n , either with respect to the normalized Laplacian, or more generally with respect to $A + tD$.

Theorem 5. For n large, there exists a family of $2^{\lfloor n/7 \rfloor}$ non-isomorphic, mutually cospectral graphs with respect to the normalized Laplacian.

Proof. Find a graph G on $\lfloor n/7 \rfloor$ vertices that has a trivial automorphism group (for $n \geq 42$ this is easy to do). For each vertex v of G , attach one of the two inflated stars shown in Figure 10 by identifying v with the vertex marked v in the inflated star. This constructs $2^{|G|} = 2^{\lfloor n/7 \rfloor}$ graphs on $7\lfloor n/7 \rfloor \leq n$ vertices. We note that for any one of these graphs we can easily recover G and identify which attachment was done at v and so these graphs are all non-isomorphic. Further, by Theorem 1 changing our choice of widget at any single vertex does not change the spectrum and so all of the graphs are cospectral. \square

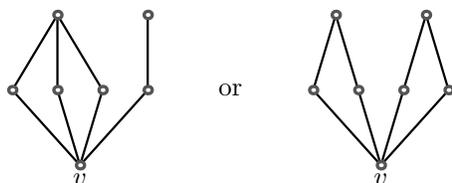


Figure 10: The two different inflated stars to add to vertices v in Theorem 5.

We have not tried to optimize the construction. For example, if instead of using the family of inflated stars with $n = 4$ and $k = 2$ we used the family of inflated stars with $n = 21$ and $k = 5$ (of which there are 101 such non-isomorphic members), then we get a family with rate of growth $101^{\lfloor n/26 \rfloor} \approx 1.19423^n$ which is faster than $2^{\lfloor n/7 \rfloor} \approx 1.10409^n$.

We similarly have the following result for graphs cospectral with respect to $A + tD$.

Theorem 6. *For n large, there exists a family of $2^{\lfloor n/13 \rfloor}$ non-isomorphic, mutually cospectral graphs with respect to the matrix $A + tD$.*

Proof. Find a graph G on $\lfloor n/13 \rfloor$ vertices that has a trivial automorphism group (for $n \geq 78$ this is easy to do). For each vertex v of G attach one of the two graphs shown in Figure 11, which are taken from Figure 6. This constructs $2^{|G|} = 2^{\lfloor n/13 \rfloor}$ graphs on $13\lfloor n/13 \rfloor \leq n$ vertices. We note that for any one of these graphs we can easily recover

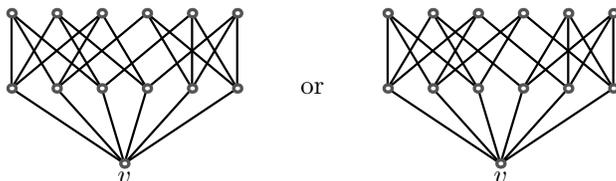


Figure 11: The two different widgets to add to vertices v in Theorem 6.

G and identify which attachment was done at v , so these graphs are all non-isomorphic. Further, by Theorem 3 changing our choice of widget at any single vertex does not change the spectrum and so all of the graphs are cospectral. \square

5 Concluding remarks

The method of swapping bipartite subgraphs was discovered by examining all cospectral pairs of graphs on at most eight vertices (all cospectral pairs on at most seven vertices has previously appeared in Tan [20]), which included several simple examples of this type. The set of all cospectral graphs on at most eight vertices can be easily constructed in SAGE (see Appendix A).

By looking at the small cases, we also discovered some new graphs which are regular, but are cospectral for the normalized Laplacian with graphs which are not regular. We have already seen that the four-cycle and the eight-cycle are members of cospectral pairs

(see Figures 1d and 4). Also, $K_{n,n}$ is cospectral with $K_{p,q}$ for $p+q = 2n$. Another example of a graph which is regular, but is cospectral with a graph which is not regular, is shown in Figure 12. These examples show that the normalized Laplacian cannot, in general, detect whether a graph is regular.

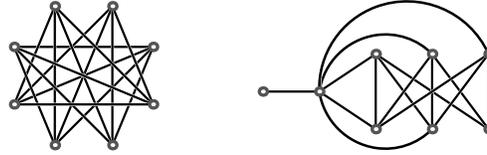


Figure 12: A regular non-bipartite graph which is cospectral for the normalized Laplacian with a graph which is not regular.

As noted in the introduction, the number of edges in a graph is not uniquely determined by the spectrum of the normalized Laplacian. As a consequence, it is possible for a graph to be cospectral with one of its subgraphs; examples of this are shown in Figure 13. Further examples of graphs which are cospectral with respect to the normalized Laplacian but have differing number of edges are shown in Figure 14 (there is currently no known method to generate cospectral graphs with differing number of edges other than using complete bipartite graphs). This also indicates the difficulty in counting the number of cospectral graphs. While for the adjacency and the combinatorial Laplacian, we could first subdivide the graphs according to the number of edges and work with this coarsening when finding cospectral pairs (see [15]), this is no longer possible for the normalized Laplacian.

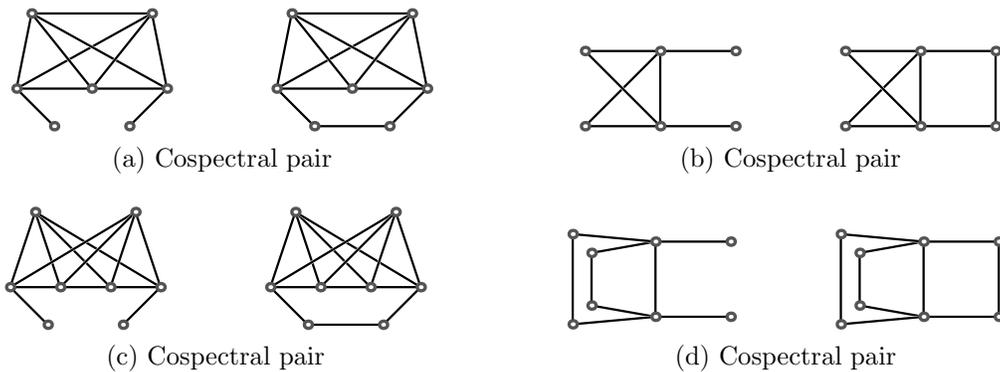


Figure 13: Examples of graphs which are cospectral with a subgraph for the normalized Laplacian.

It would be interesting to find techniques for constructing cospectral graphs for the normalized Laplacian which differ in the number of edges. Similarly, it would be interesting to investigate the cospectral graphs for the signless Laplacian and see if there are any constructions unique to forming cospectral graphs for that matrix (similar to what Theorem 1 is for the normalized Laplacian).

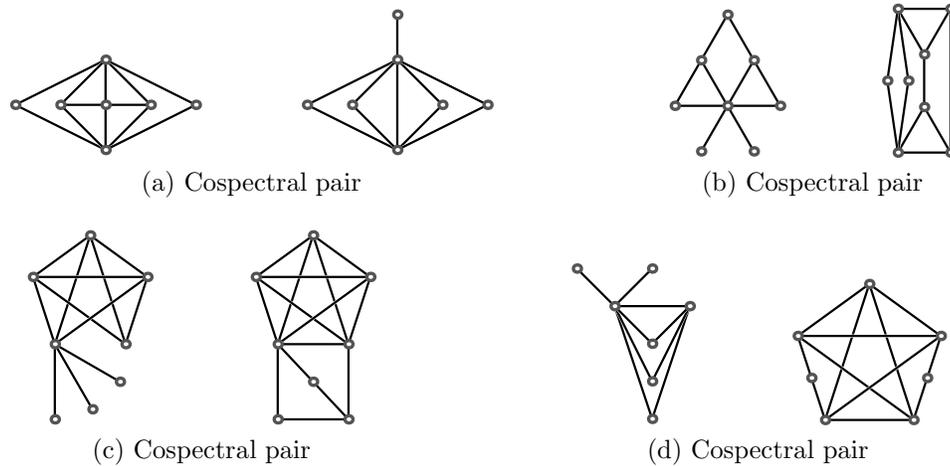


Figure 14: Examples of graphs which are cospectral for the normalized Laplacian but differ in the number of edges.

References

- [1] Andries E. Brouwer and Edward Spence, *Cospectral graphs on 12 vertices*, Electronic Journal of Combinatorics **16** (2009), #N20, 3pp.
- [2] Steve Butler, *Cospectral graphs for both the adjacency and normalized Laplacian matrices*, Linear and Multilinear Algebra **58** (2010), 387–390.
- [3] Michael Cavers, *The normalized Laplacian matrix and general Randic index of graphs*, Ph.D. Thesis, University of Regina, 2010.
- [4] Fan Chung, *Spectral Graph Theory*, CBMS Lecture Notes, AMS, Providence, RI, 1997.
- [5] Dragoš M. Cvetković, *Graphs and their spectra*, Univ. Beograd. Publ. Elektrotehn. Fak., Set. Mat. Fiz. **354-356** (1971), 1–50.
- [6] Dragoš M. Cvetković, Michael Doob and Horst Sachs, *Spectra of Graphs: Theory and Application*, Academic Press, New York, NY, 1979.
- [7] Dragoš M. Cvetković and Slobodan K. Simić, *Towards a spectral theory of graphs based on the Signless Laplacian I*, Institut Mathématique. Publications. Nouvelle Série **85(99)** (2009), 19-33.
- [8] Dragoš M. Cvetković and Slobodan K. Simić, *Towards a spectral theory of graphs based on the Signless Laplacian II*, Linear Algebra and its Applications **432** (2010), 2257-2272.
- [9] Dragoš M. Cvetković and Slobodan K. Simić, *Towards a spectral theory of graphs based on the Signless Laplacian III*, Applicable Analysis and Discrete Mathematics **4** (2010), 156-166.

- [10] Edwin R. van Dam, *Graphs with few eigenvalues: an interplay between combinatorics and algebra*, Ph.D. Thesis, Tilburg University, Center for Economic Research Dissertations Series, No. 20, 1996.
- [11] Edwin R. van Dam and Willem H. Haemers, *Which graphs are determined by their spectrum?*, *Linear Algebra and its Applications* **373** (2003), 241–272.
- [12] Edwin R. van Dam, Willem H. Haemers and Jack H. Koolen, *Cospectral graphs and the generalized adjacency matrix*, *Linear Algebra and its Applications* **423** (2007), 33–41.
- [13] Chris D. Godsil and Brendan McKay, *Products of graphs and their spectra*, *Combinatorial mathematics, IV (Proc. Fourth Australian Conf., Univ. Adelaide, Adelaide, 1975)*, *Lecture Notes in Math.*, Vol. 560, Springer, Berlin, 1976, 61–72.
- [14] Chris D. Godsil and Brendan McKay, *Constructing cospectral graphs*, *Aequationes Mathematicae* **25** (1982), 257–268.
- [15] Willem H. Haemers and Edward Spence, *Enumeration of cospectral graphs*, *European Journal of Combinatorics* **25** (2004), 199–211.
- [16] Charles R. Johnson and Morris Newman, *A note on cospectral graphs*, *Journal of Combinatorial Theory, Ser. B* **28** (1980), 96–103.
- [17] Russell Merriss, *Large families of Laplacian isospectral graphs*, *Linear and Multilinear Algebra* **43** (1997), 201–205.
- [18] Johan Jacob Seidel, *Graphs and two-graphs*, *Proceedings of the Fifth Southeastern Conference on Combinatorics, Graph Theory and Computing (Florida Atlantic Univ., Boca Raton, Fla., 1974)*, *Congressus Numerantium*, No. X, Utilitas Math., Winnipeg, Man., 1974, 125–143.
- [19] William A. Stein, et al., *Sage Mathematics Software (Version 4.6.1)*, The Sage Development Team, 2011, <http://www.sagemath.org>.
- [20] Jinsong Tan, *On isospectral graphs*, *Interdisciplinary Information Sciences* **4** (1998), 117–124.
- [21] Richard Wilson and Ping Zhu, *A study of graph spectra for comparing graphs and trees*, *Journal of Pattern Recognition*, **41** (2008), 2833–2841.

A SAGE code to generate cospectral graphs

The code listing below shows how to use Sage version 4.5.2 [19] to generate and check cospectral graphs with 8 vertices. In the listing below, we specifically generate cospectral graphs with respect to the normalized Laplacian. The lines starting with the number sign “#” are comment lines.

```
def DinverseA(g):
    "Calculate  $D^{-1}A$  for a graph g"
    A=g.adjacency_matrix().change_ring(QQ)
    for i in range(g.order()):
5       A.rescale_row(i, 1/len(A.nonzero_positions_in_row(i)))
    return A

# Calculate the graphs and store them in cospectral_list.
# We use the DinverseA function defined above to determine
10 # if two graphs are cospectral.
# This command takes a few minutes to complete.
cospectral_list=graphs.cospectral_graphs(8,
                                       graphs=lambda g: min(g.degree())>0)

15 # Give a list of all graph6 strings of these graphs.
graph6_list=[[g.graph6_string() for g in glist]
             for glist in cospectral_list]

# Show all of the graphs, each row being cospectral.
20 for glist in cospectral_list: show(glist)

# Get the first two cospectral graphs and check
# to make sure they are cospectral.
graph1, graph2 = cospectral_list[0]
25 graph1_poly=graph1.laplacian_matrix(normalized=True).charpoly()
graph2_poly=graph2.laplacian_matrix(normalized=True).charpoly()
graph1_poly==graph2_poly

# Construct fuzzy ball graphs corresponding to the partitions
30 # 4=3+1 and 4=2+2 and check that they are cospectral.
fb1=graphs.FuzzyBallGraph([3,1],0)
fb2=graphs.FuzzyBallGraph([2,2],0)
fb1_poly=fb1.laplacian_matrix(normalized=True).charpoly()
fb2_poly=fb2.laplacian_matrix(normalized=True).charpoly()
35 fb1_poly==fb2_poly
```