

# Spectral Graph Theory

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# Chapter 1

## Expander Graphs

*In this chapter, we assume all graphs are undirected, contain no loops, nor have any duplicate edges.*

Given a graph  $G$  containing some vertex  $v$ , we let  $N(v)$  denote the set of all vertices in  $G$  connected to  $v$  by an immediate edge, and we call this set the **neighbourhood** of  $v$  in  $G$ . One can study the properties of the neighbourhood function via the **adjacency operator**  $A : \mathbb{C}^V \rightarrow \mathbb{C}^V$  given by

$$(Af)(v) = \sum_{w \in N(v)} f(w)$$

With respect to the basis induced by the characteristic functions over the vertices of the graph,  $A$  is represented in matrix form by the adjacency matrix of its graph.

A graph isomorphism replaces an adjacency with a similar matrix, so the properties of the operator  $A$  invariant under basis changes are isomorphism invariant. We shall find that in particular, the eigenvalues of  $A$  give useful information about the graph  $G$ , especially if  $G$  is a regular graph. As  $G$  is undirected,  $A$  is a self-adjoint operator, and therefore is diagonalizable, with  $n$  real eigenvalues  $\lambda_1 \geq \dots \geq \lambda_n$ . We will write  $\lambda_i(G)$  if the graph is not specified. The study of the relationship between a graph  $G$  and the eigenvalues of the operator  $A$  form the core study of spectral graph theory.

**Example.** *The invariant subspaces of  $A$  correspond to the components of  $G$ . That is, if  $H$  is a connected component of  $G$ , then the subspace of functions*

vanishing outside of  $H$  is invariant under the action of  $A$ . On the other hand, if  $G$  is a connected graph, then all the invariant subspaces of  $A$  are trivial.

**Proposition 1.1.** *If  $\Delta(G)$  denotes the maximum degree in a graph  $G$ , and  $\delta(G)$  the minimum degree, then for any function  $f \in \mathbf{C}^G$  with  $L_0 \leq f \leq L_1$ ,*

$$\delta(G)L_0 \leq Af \leq \Delta(G)L_1$$

*Proof.* We find that for any vertex  $v$

$$\sum_{w \in N(v)} f(w) \leq \sum_{w \in N(v)} L_1 = |N(v)|L_1 \leq \Delta(G)L_1$$

and the reverse direction gives

$$\sum_{w \in N(v)} f(w) \geq \sum_{w \in N(v)} L_0 = |N(v)|L_0 \geq \delta(G)L_0$$

This gives the required inequality.  $\square$

**Corollary 1.2.** *For any eigenvalue  $\lambda$  of  $A$ ,  $|\lambda| \leq \max(\delta(G), \Delta(G))$ .*

*Proof.* The spectral radius of  $A$  is bounded by any of the matrix norms giving the space of operators a Banach algebra structure. In particular, the spectral radius is bounded by  $\|A\|_\infty$ , and the theorem above bounds the values of  $\|A\|_\infty$  as required.  $\square$

**Example.** *As a particular example of this, we find that for  $K$  regular graphs, we find  $-K \leq \lambda_n \leq \lambda_1 \leq K$ , and we actually have  $\lambda_1 = K$ , because  $A$  operates on constant functions by multiplication by  $K$ .*

**Proposition 1.3.** *On a  $K$  regular graph,  $\lambda_2 = K$  iff  $G$  is disconnected.*

*Proof.* If we break  $G$  into two connected components  $H_0$  and  $H_1$ , then the functions vanishing outside of  $H_0$  and the functions outside of  $H_1$  form two complementary invariant subspaces of  $G$ , and both of these subspaces contain eigenfunctions of eigenvalue 1. On the other hand, we show that if  $G$  is connected, then every eigenfunction  $f$  of eigenvalue  $K$  is constant. Let  $v^*$  maximize  $f$  over all vertices in  $v$ . Then

$$Kf(v^*) = \sum_{w \in N(v^*)} f(w) \leq Kf(v^*)$$

If  $f(w) \neq f(v^*)$  for some  $w \in N(v^*)$ , the inequality above is strict, which is impossible. Thus if  $v^*$  maximizes  $f$ , then all its neighbours maximize  $f$ . Since  $G$  is connected, we conclude that  $f$  is constant on  $G$ .  $\square$

**Lemma 1.4.** *If  $f$  is an eigenfunction of a  $K$  regular graph with eigenvalue  $\lambda \neq K$ , then  $\sum f(v) = 0$ .*

*Proof.* If  $\sum f(v) \neq 0$ , we may assume without loss of generality that  $\sum f(v) = 1$ . Then, if we take the sum of the adjacency of the operator over all vertices, we sum over each vertex  $K$  times, and so

$$K = K \sum_v f(v) = \sum_v \sum_{w \in N(v)} f(w) = \sum_v \lambda f(v) = \lambda$$

and this gives the required equality.  $\square$

**Example.** *If  $G$  is a  $K$  regular graph, then  $\lambda_n = -K$  if and only if  $G$  has a bipartite connected component. Because of our discussion, we might as well assume  $G$  is connected, because invariant subspaces contain all the eigenfunctions of an operator. If  $G = H_0 \cup H_1$  is bipartite, the function  $\chi_{H_0} - \chi_{H_1}$  has eigenvalue  $-K$ . On the other hand, if  $f$  is an eigenfunction of  $A$  with eigenvalue  $-K$ , we claim that  $f$  must cycle over two distinct values, which are negations of each other. Let  $v^*$  maximize  $f$ , and let  $v_*$  minimize the function. We know that  $f(v_*) < 0 < f(v^*)$ . Since*

$$-Kf(v^*) = \sum_{w \in N(v^*)} f(w) \geq Kf(v_*)$$

*We conclude that  $f(v_*) \leq -f(v^*)$ . The same argument applied to the neighbours of  $v_*$  gives  $-f(v_*) \leq f(v^*)$ . Putting these two inequalities together gives  $-f(v_*) = f(v^*)$ . If  $f(w) \neq f(v_*)$  for  $w \in N(v^*)$ , then the inequality above is strict, which is impossible. Similarly, we conclude that  $f(w) = f(v^*)$  for each  $w \in N(v_*)$ . Since  $v_*$  and  $v^*$  were arbitrary vertices minimizing and maximizing  $f$ , every vertex takes one of these two values, and adjacent vertices have opposite values, so these two values give a bipartite structure on the given graph.*

A  $K$  regular graph  $G$  is said to be an  $\varepsilon$  **expander** (one sided) if  $\lambda_2 \leq (1 - \varepsilon)K$ , and a *two sided*  $\varepsilon$  expander if one also has  $\lambda_n \geq -(1 - \varepsilon)K$ . Every connected graph is an  $\varepsilon$  expander for some  $\varepsilon$ , and a non bipartite graph is a two sided expander for some  $\varepsilon$ . Thus unless we want to perform analysis parameterized by  $\varepsilon$ , we are forced to look at ‘limits’ of graphs which control their eigenvalues. A sequence of  $K$  regular graphs is said to be an expander family if there is a  $\varepsilon$  such that eventually the graphs are all  $\varepsilon$  expanders.

**Theorem 1.5.** *Let  $G$  be a  $K$  regular graph on  $n$  vertices. Then*

- $\sum \lambda_i = 0$ .
- $\sum \lambda_i^2 = nK$ .
- $\max(|\lambda_2|, |\lambda_n|) \geq \sqrt{K} - o_K(1)$ .

where  $o_K(1)$  denotes a quantity tending to zero at a rate dependant on  $K$ .

*Proof.* In its adjacency matrix representation, the adjacency operator  $A$  has no loops, hence its diagonal vanishes and so it has trace zero. It follows that the sum of the eigenvalues of  $A$  is equal to zero. The sum of the squares of the eigenvalues is equal to the trace of  $A^2$ , which corresponds to a kind of ‘second order’ adjacency operator. Since we have no multiedges,

$$(A^2 \chi_v)(v) = \sum_{w \in N(v)} \sum_{u \in N(w)} \chi_v(u) = K$$

Hence, with respect to the canonical basis on the graph,  $A^2$  has diagonal entries  $K$ , and hence has trace  $Kn$ . Finally, this shows that

$$nK = \sum \lambda_i^2 = \sum |\lambda_i|^2 \leq 1 + (n-1) \max(|\lambda_2|^2, |\lambda_n|^2)$$

and therefore that

$$\max(|\lambda_2|, |\lambda_n|) = \sqrt{\max(|\lambda_2|^2, |\lambda_n|^2)} \geq \sqrt{\frac{nK-1}{n-1}}$$

and the last term is  $\sqrt{K} - \sqrt{K}O(1/n) + K^2O(1/n^2)$ . □

This result places an upper bound on the rate of a two sided expansion for large graphs. A more sophisticated result sharpens the inequality to obtain the improvement

$$\max(|\lambda_2|, |\lambda_n|) \geq 2\sqrt{K-1} - o_K(1)$$

Graphs with  $\max(|\lambda_2|, |\lambda_n|) \leq 2\sqrt{K-1}$  are known as Ramanujan graphs, and have connections to number theory.

**Example.** For each  $n$ , let  $G_n$  be the 2-regular graph whose vertex set is  $\mathbf{Z}_n$ , and such that the neighbours of  $k$  are  $k+1$  and  $k-1$ . Then the adjacency operator  $A_n$  is really just the combination of shifts  $L_1 + L_{-1}$  in disguise. Now if  $\chi_1, \dots, \chi_n$  are the basis of  $n$  characters on  $\mathbf{Z}_n$ , with  $\chi_m(1) = e^{2i\pi m/n}$ , then

$$\begin{aligned} A_n(\chi_m)(k) &= \chi_m(k+1) + \chi_m(k-1) \\ &= (\chi_m(1) + \chi_m(-1))\chi_m(k) \\ &= \cos(2\pi m/n)\chi_m(k) \end{aligned}$$

As  $n \rightarrow \infty$ , the space of eigenvalues of  $A_n$  becomes dense in  $[-1, 1]$ , so this sequence of graphs cannot be an expander family for any  $\varepsilon$ .

**Example.** The complete graph  $C_n$  on  $n$  vertices, which is an  $n-1$  regular graph, is an excellent example of an expander graph family. If  $f$  is an eigenfunction of  $A_n$  with eigenvalue  $\lambda_i \neq n-1$ , then  $\sum f(v) = 0$ , and so

$$\lambda_i f(v) = \sum_{w \neq v} f(w) = -f(v)$$

so  $\lambda_i = -1$ , and the sequence of eigenvalues for  $A_n$  is  $1, -1, \dots, -1$ . This shows that  $C_n$  is a one-sided  $1 + 1/(n-1)$  expander, and a  $1 - 1/(n-1)$  two sided expander. The goal of the theory of expander graphs is to find sparse graphs with similar properties to complete graphs, and so we desire sparse expander graphs.

**Example.** If  $G$  is a  $K$  regular graph on  $n$  edges, then the complement graph  $G^c$  is an  $n-K-1$  regular graph. If  $A$  is the adjacency operator corresponding to  $G$ , and  $B$  the operator corresponding to  $G^c$ , then  $(A+B)(f)(v) = \sum f(w) - f(v)$ . In particular, if  $f$  is an eigenvector for  $A$  with eigenvalue  $\lambda_i \neq K$ , then  $\sum f(w) = 0$ , and so  $Bf = -(1+A)f = -(1+\lambda_i)f$ . If we assume  $G$  is connected, then the sequence  $K \neq \lambda_2 \geq \dots \geq \lambda_n$  gives the eigenvalues  $-(1+\lambda_2) \leq \dots \leq -(1+\lambda_n)$ . Since none of the eigenfunctions  $f$  corresponding to the eigenvalues  $\lambda_2, \dots, \lambda_n$  are constant, we find that  $B$  has the additional eigenvalue  $n-K-1$  corresponding to the constant function.

**Example.** Let  $G$  be the complete bipartite graph between two sets of  $n$  vertices. Then  $\lambda_{2n} = -n$ , because  $G$  is bipartite. Note that for any function  $f$ ,  $Af$  is a function which is constant on each side of the bipartition, so for any eigenfunction  $f$  with a nonzero eigenvalue  $\lambda$ ,  $f$  is constant on each side of the bipartition. But this means that if  $f$  takes the value  $x$  on one side of the bipartition, and  $y$

on the other side, then  $\lambda x = ny$ , and  $\lambda y = nx$ . If  $x = 0$ , we find that  $y = 0$ . If  $x \neq 0$ , then  $\lambda = ny/x$ , and also  $ny^2/x = nx$ , so  $y^2 = x^2$  implying that  $y = \pm x$ . These give the two eigenfunctions corresponding to the constant function with eigenvalue  $n$  and the bipartite eigenfunction with eigenvalue  $-n$ . This tells us that the remaining eigenvalues  $\lambda_2, \dots, \lambda_{2n-1}$  are all equal to zero.

Viewing a graph as the discrete version of a  $K$  regular graph, and  $f$  is a function on the graph, we can define the **discrete gradient magnitude**

$$|\nabla f|(v) = \sqrt{\sum_{w \in N(v)} |f(w) - f(v)|^2}$$

The classical Poincaré inequality in Euclidean space says that  $\|f\|_2 \leq C \|\nabla f\|_2$ . This inequality is actually connected to the theory of expanders, because a graph  $G$  is a one sided  $\varepsilon$  expander if and only if

$$\|\nabla f\|_2^2 \geq 2K\varepsilon \|f\|_2^2$$

for any mean zero function  $f$ . We may assume that  $G$  is a connected graph to prove this theorem, for if  $G$  is not connected it is not an expander for any  $\varepsilon > 0$ . We can write

$$\begin{aligned} \sum_v \sum_{w \in N(v)} |f(w) - f(v)|^2 &= \sum_v \sum_{w \in N(v)} |f(w)|^2 + |f(v)|^2 - 2\Re[\overline{f(v)}f(w)] \\ &= 2K\|f\|_2^2 - 2\Re\left(\sum_v \overline{f(v)}(Af)(v)\right) \\ &= 2K\|f\|_2^2 - 2\Re\langle f, Af \rangle \end{aligned}$$

If  $f$  has mean zero, then we can write  $f = \sum f_j$  for the decomposition of  $f$  into the  $j$ 'th eigenspace. Since  $A$  is self adjoint, these eigenspaces are orthogonal, and so

$$\langle f, Af \rangle = \sum \lambda_j \|f_j\|_2^2$$

Since  $\sum_v f_j(v) = 0$  for all  $j > 1$ , we find  $0 = \sum_v f(v) = \mu_1 \sum_v f_1(v)$ . Since  $\sum_v f_1(v) \neq 0$ , we find  $\mu_1 = 0$ . This means that

$$\sum \lambda_i \|f_i\|_2^2 \leq \sum \lambda_2 \|f_i\|_2^2 = \lambda_2 \|f\|_2^2$$

and this shows that

$$\|\nabla f\|_2^2 \geq 2(K - \lambda_2) \|f\|_2^2$$

and  $K - \lambda_2$  gives the maximal value of  $\varepsilon$  for which  $G$  is a  $\varepsilon$  expander.



## 1.1 Expanders and Edge Expansion

We now make the intuition about expander graphs having good ‘expansion properties’ precise. Given two disjoint sets of vertices  $V$  and  $W$ , we let  $E(V, W)$  denote the set of edges between the two vertex sets. We find that

$$|E(V, W)| = \langle A\chi_V, \chi_W \rangle$$

Given a vertex set  $V$ , we let  $\delta(V)$  denote the set of edges leaving  $V$ . We define the **edge expansion ratio**

$$h(G) = \min_{|V| \leq n/2} \frac{|\delta(V)|}{|V|} = \min_{|V| \leq n/2} \frac{\langle A\chi_V, 1 - \chi_V \rangle}{|V|}$$

The choice that  $|V| \leq n/2$  is done to avoid trivial values where we let  $V$  be the set of all vertices in the graph, so that  $\delta(V) = \emptyset$ . The edge expansion ratio can be seen as a discrete isoperimetry bound for the graph, and as such it is often called the **Cheeger constant** of the graph.

**Proposition 1.6.**  *$h(G) \neq 0$  if and only if  $G$  is connected, and more generally, a family of  $K$  regular graphs  $G_i$  is an expander family if  $h(G_i)$  is lower bounded.*

*Proof.* We may assume each graph is connected. Fix  $\varepsilon > 0$  such that eventually  $\lambda_2 \leq (1 - \varepsilon)K$ . For any subset  $V$  of edges in  $G_i$ , the projection of  $\chi_V$  onto the first eigenspace is  $|V|/n$ , and so

$$\langle A\chi_V, \chi_V \rangle \leq \frac{1}{n}K|V|^2 + (1 - \varepsilon)K \left\| \chi_V - \frac{|V|}{n} \right\|^2$$

And

$$\begin{aligned} \left\| \chi_V - \frac{|V|}{n} \right\|^2 &= |V| \left( 1 - \frac{|V|}{n} \right)^2 + [n - |V|] \left( \frac{|V|}{n} \right)^2 \\ &= |V| - \frac{1}{n}|V|^2 \end{aligned}$$

Hence

$$\frac{1}{n}K|V|^2 + (1 - \varepsilon)K \left\| \chi_V - \frac{|V|}{n} \right\|^2 \leq (1 - \varepsilon)K|V| - \frac{\varepsilon K|V|^2}{n} \leq (1 - \varepsilon/2)K|V|$$

Since  $\langle A\chi_V, 1 \rangle = K|V|$ , this shows that

$$\begin{aligned} h(G) &= \min_{|V| \leq n/2} \frac{|\delta(V)|}{|V|} = \min_{|V| \leq n/2} \frac{\langle A\chi_V, 1 - \chi_V \rangle}{|V|} \\ &= \min_{|V| \leq n/2} \frac{\langle A\chi_V, 1 \rangle}{|V|} - \frac{\langle A\chi_V, \chi_V \rangle}{|V|} \\ &\geq \min_{|V| \leq n/2} K - (1 - \varepsilon/2)K = (\varepsilon/2)K \end{aligned}$$

Another way to see this is that  $\langle A\chi_V, \chi_V \rangle$  counts the number of edges between vertices in  $V$ , and since each vertex in  $V$  is adjacent to exactly  $K$  vertices, we conclude that the total number of edges leaving  $V$  is exactly  $K|V| - \langle A\chi_V, \chi_V \rangle$ , which we have bounded below by  $(\varepsilon/2)K|V|$ .

The other direction is harder. The difficulty is that the lower bound  $h(G) \geq c$  enables us to conclude that  $\langle A\chi_V, \chi_V \rangle \leq (K - c)|V|$  for all vertex sets with  $|V| \leq n/2$ , whereas proving that  $G$  is an expander requires us to understand  $\langle Af, f \rangle$  for all functions  $f$ , because we know

$$\lambda_2 = \sup_{\sum f(v)=0} \frac{\langle Af, f \rangle}{\|f\|_2^2}$$

so it suffices to show  $\langle Af, f \rangle \leq (1 - \varepsilon)K$  for all functions  $f$  with mean zero and with  $\|f\|_2 = 1$ , for some  $\varepsilon$  depending only on  $K$  and  $c$ . Since  $A$  is real, we may assume that  $f$  is real. We will prove that if  $f$  is non-negative, and is supported on a set of cardinality  $n/2$ , then  $\langle Af, f \rangle \leq (1 - c)K\|f\|_2^2$ . To see how this implies the main inequality, note that if we write  $f = f_+ - f_-$ , then

$$\langle Af, f \rangle \leq \langle Af_+, f_+ \rangle + \langle Af_-, f_- \rangle$$

and  $1 = \|f_+\|_2^2 + \|f_-\|_2^2$ . Either  $f_+$  or  $f_-$  is supported on a set of size less than or equal to  $n/2$ , and by symmetry we may assume this is  $f_-$ . Consider a small value  $\sigma$ , to be fixed later. If  $\|f_-\|_2^2 \geq \sigma^2$ , then applying the trivial bound  $\langle Af_+, f_+ \rangle \leq K\|f_+\|_2^2$

$$\langle Af_+, f_+ \rangle + \langle Af_-, f_- \rangle \leq K\|f_+\|_2^2 + (1 - c)K\|f_-\|_2^2 \leq K(1 - \sigma^2)\|f\|_2^2$$

TODO: FINISH BOUNDS LATER. □

**Example.** On the graph defined on  $\mathbf{Z}_n$  we considered before, for  $n > 2$ , the set of points  $\{1, \dots, n/2\}$  has two boundary edges, between 0 and 1, and between

$n/2$  and  $n/2 + 1$ . It follows that  $h(G_n) \leq 4/n$ , and this is an equality, because if  $|V| \leq n/2$ , then  $\delta(V)$  contains at least two edges, so  $\delta(V) \geq 2 \geq 4|V|/n$ . This is another way to think about why the graphs are not a family of expander graphs.

**Example.** Every 2 regular graphs breaks down into connected components, which form loops in the graph. It follows that if  $G$  is connected, it is isomorphic to  $\mathbb{Z}_n$ , and any family of such graphs which form an expander family must have bounded size.

A more precise relationship between the best value  $\varepsilon$  that makes  $G$  into a one sided expander and it's Cheeger constant. Namely,

$$\frac{\varepsilon K}{2} \leq h(G) \leq \sqrt{2\varepsilon K}$$

proved in the 1980s by Dodzuik and Alon-Milman.

TODO: EXERCISES

## 1.2 Random Walks

We now discuss how the theory of expanders connects to the theory of convergence rates of random walks on graphs. Given a graph  $G$  and an initial vertex  $v_0$  (which can be randomly chosen), the random walk  $v_0, v_1, v_2, \dots$  is chosen such that  $v_{i+1}$  is obtained from  $v_i$  by choosing a neighbour uniformly at random. We will let  $\mu_i$  be the function on the vertices of  $G$  such that  $\mu_i(v) = \mathbf{P}(v_i = v)$ . Then arguing by conditional probabilities, we find that  $\mu_{i+1} = (A\mu_i)/K$ . Provided our graph is connected and aperiodic, the corresponding Markov chain is ergodic, and the distributions  $\mu_i$  will converge pointwise to an eigenfunction for the adjacency operator of highest eigenvalue. In particular, over a  $K$  regular connected graph the distribution will eventually be indistinguishable from the uniform distribution.

We can measure how fast  $\mu_i$  converges to a constant distribution by quantifying  $\|\mu_i - 1/n\|_2$ . It is decreasing in  $i$ , because one calculates

$$\begin{aligned} \|A\mu_i/K - 1/n\|_2^2 - \|\mu_i - 1/n\|_2^2 &= \left[ \frac{\|A\mu_i\|_2^2}{K^2} - \frac{1}{n} \right] - \left[ \|\mu_i\|_2^2 - \frac{1}{n} \right] \\ &= \frac{\|A\mu_i\|_2^2}{K^2} - \|\mu_i\|_2^2 \end{aligned}$$

If  $\mu_i$  has a decomposition as  $\sum \mu_{ij}$  via its eigenspaces, then  $\mu_{i1} = 1/n$ , because  $\sum \mu_i(v) = 1$ , and so

$$\|A\mu_i\|_2^2 \geq K^2 \|\mu_{i1}\|_2^2 = K^2$$

Yet  $\|\mu_i\|_2^2 = \sum \mu_i(v)^2 \leq \sum \mu_i(v) = 1$ , so

$$\frac{\|A\mu_i\|_2^2}{K^2} - \|\mu_i\|_2^2 \geq 1 - 1 = 0$$

This is an equality only when  $\mu_i = 1/n$ , so the function is decreasing everywhere else. The expansion properties are intricately tied to the expansion properties of the graphs.

**Theorem 1.7.** *Fix  $\alpha > 1/2$ . A sequence of  $K$  regular graphs  $G_n$  with  $m_n$  vertices is a two-sided expander family if and only if there is  $C > 0$  independent of  $n$  such that for sufficiently large  $n$ ,  $\|\mu_i - m_n^{-1}\|_2 \leq m_n^{-\alpha}$  for all  $i \geq C \log m_n$ , and all choices of initial vertices  $v_0$ .*

Note that the theorem holds for any initial probability distribution by an easy application of the Minkowski inequality. Thus from a dynamical systems point of view, the uniform distribution is a very strong attractor in the space of all probability distributions. Essentially, this theorem says that two sided expanders are those graphs such that random walks become close to uniform in  $O(\log n)$  steps. On the other hand, the central limit theorem only implies that the walks  $Z_n$  become close to uniformly mixing only at time beyond  $n^2$ , as indicated by the central limit theorem (TODO: WHY?). This theorem is useful for generating near random distributions using little work by taking a random walk on some basic combinatorial structure.