

# A Five-Step Energy-Decomposition Proof of the Discrete Dirichlet Ground State Theorem

## Abstract

We give a completely rigorous and undergraduate-accessible proof of the classical “ground state” theorem for the Dirichlet graph Laplacian: on a connected interior, the smallest Dirichlet eigenvalue is simple and admits a strictly positive eigenfunction on the interior. The proof is organized to mirror a general methodology used in Dirichlet form theory: (1) energy decomposition, (2) Markov property, (3) irreducibility (connectivity of the interior), (4) compact resolvent (finite-dimensional), and (5) Perron–Frobenius/Krein–Rutman for a positive compact operator. Every step is proved concretely for finite graphs.

**About standardness.** We label each stated result as **(Standard)** when it is a well-known statement from linear algebra, spectral graph theory, discrete maximum principles, or Perron–Frobenius theory, and as **(New presentation)** when the content is primarily expository (e.g., the five-step organization or packaging of several standard facts). Proofs are included even for **(Standard)** results when they are short and illuminating.

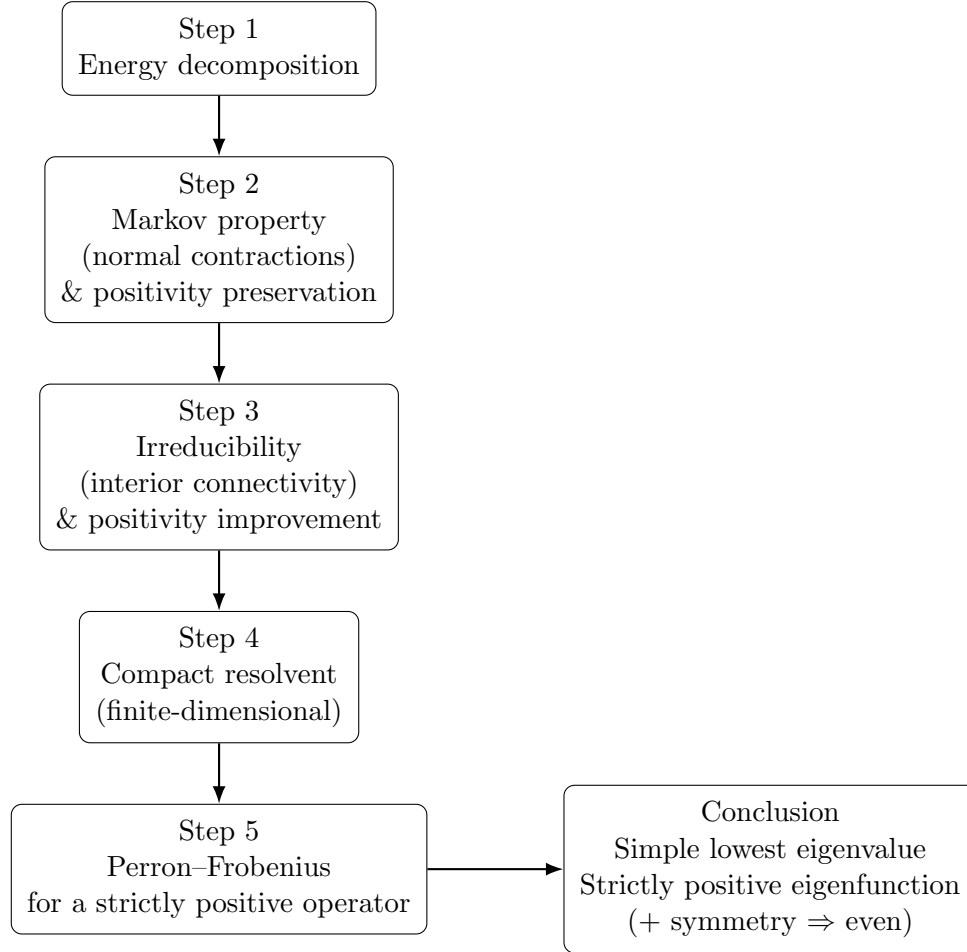
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# 1 The five-step spine and the main theorem

## 1.1 Flowchart



## 1.2 Main statement

**Definition 1** (Weighted graph with boundary and connected interior). Let  $G = (V, E, w)$  be a finite undirected graph with vertex set  $V$ , edge set  $E \subseteq \{\{x, y\} : x \neq y\}$ , and weights  $w_{xy} = w_{yx} > 0$  for  $\{x, y\} \in E$  (and  $w_{xy} = 0$  otherwise). Fix a nonempty boundary set  $B \subset V$ , and write  $I := V \setminus B$  for the interior. Assume:

- (i)  $G$  is connected (as an undirected graph),
- (ii)  $B \neq \emptyset$  and  $I \neq \emptyset$ ,
- (iii) the *interior induced graph*  $G_I := (I, E_I, w)$  is connected, where

$$E_I := \{\{x, y\} \in E : x \in I, y \in I\}.$$

*Remark 2* (Why interior connectivity is the right hypothesis). If  $G_I$  is not connected, then the Dirichlet operator  $L_B$  on  $\mathbb{R}^I$  decomposes into a direct sum over the connected components of  $G_I$ , and the global smallest eigenvalue need not be simple. In particular, the conclusion “ $\varphi_1 > 0$  on all of  $I$ ” can fail. Theorem 4 therefore assumes the standard analogue of “connected domain”: connectivity of the interior. A fully general componentwise statement is recorded in Remark 23 below.

**Definition 3** (Dirichlet Laplacian). For a function  $f : V \rightarrow \mathbb{R}$  with  $f|_B = 0$ , define for  $x \in I$

$$(L_B f)(x) := \sum_{y \in V} w_{xy} (f(x) - f(y)), \quad (1)$$

where  $f(y) = 0$  for  $y \in B$  (Dirichlet condition). Equivalently,  $L_B$  is the principal submatrix of the full Laplacian corresponding to  $I$ . We view  $L_B$  as a linear operator on  $\mathbb{R}^I$  by identifying  $f \in \mathbb{R}^I$  with its zero extension to  $V$ .

**Theorem 4** (Discrete Dirichlet ground state theorem (**Standard**)). *Let  $(G, w)$  be a finite connected weighted graph and  $B \subset V$  be a nonempty boundary set, with interior  $I = V \setminus B$  nonempty, and assume the induced interior graph  $G_I$  is connected. Then:*

- (a) *The smallest eigenvalue  $\lambda_1$  of  $L_B$  (acting on  $\mathbb{R}^I$ ) is strictly positive.*
- (b)  *$\lambda_1$  is simple (geometric and algebraic multiplicity one).*
- (c) *An eigenfunction  $\varphi_1$  for  $\lambda_1$  can be chosen strictly positive on  $I$ :*

$$\varphi_1(x) > 0 \quad \text{for all } x \in I.$$

- (d) *(Optional symmetry corollary) If  $R : V \rightarrow V$  is an involution ( $R^2 = \text{id}$ ) preserving  $E, w$  and preserving  $B$ , then  $\varphi_1$  can be chosen  $R$ -even:  $\varphi_1 \circ R = \varphi_1$  on  $I$ .*

*Remark 5* (What is “known” versus what is “methodology” (**New presentation**)). Theorem 4 is classical in spectral graph theory (see, e.g., [4, 3]); what is emphasized here is the organization into the five-step energy-decomposition spine and the explicit, reusable pattern of the proof.

## 2 Step 0: The Dirichlet energy form and the variational viewpoint

We work in the finite-dimensional Hilbert space  $\ell^2(I)$  with inner product  $\langle f, g \rangle := \sum_{x \in I} f(x)g(x)$ .

**Definition 6** (Dirichlet energy form). For  $f, g \in \mathbb{R}^I$ , let  $\tilde{f}, \tilde{g} : V \rightarrow \mathbb{R}$  be their zero extensions:  $\tilde{f}|_I = f$  and  $\tilde{f}|_B = 0$ . Define the symmetric bilinear form

$$\mathcal{E}(f, g) := \frac{1}{2} \sum_{\{x, y\} \in E} w_{xy} (\tilde{f}(x) - \tilde{f}(y)) (\tilde{g}(x) - \tilde{g}(y)), \quad (2)$$

and write  $\mathcal{E}(f) := \mathcal{E}(f, f)$ .

**Lemma 7** (Green’s identity: the generator of the form (**Standard**)). *For all  $f, g \in \mathbb{R}^I$ ,*

$$\mathcal{E}(f, g) = \langle L_B f, g \rangle.$$

*In particular,  $L_B$  is self-adjoint and positive semidefinite.*

*Proof.* Write  $f, g$  for their zero extensions on  $V$  to simplify notation. Expand:

$$\mathcal{E}(f, g) = \frac{1}{2} \sum_{\{x, y\} \in E} w_{xy} (f(x) - f(y))(g(x) - g(y)).$$

Regroup by vertices  $x \in I$  (note  $g|_B = 0$ ):

$$\mathcal{E}(f, g) = \sum_{x \in I} g(x) \sum_{y \in V} w_{xy} (f(x) - f(y)) = \sum_{x \in I} g(x) (L_B f)(x) = \langle L_B f, g \rangle.$$

Self-adjointness follows by symmetry of  $\mathcal{E}$ , and  $\mathcal{E}(f) = \langle L_B f, f \rangle \geq 0$  gives positivity.  $\square$

*Remark 8* (Background: spectral theorem for symmetric matrices (**Standard**)). We repeatedly use that real symmetric matrices have an orthonormal eigenbasis and real eigenvalues (the spectral theorem); see, e.g., [5, 6].

*Remark 9* (Variational characterization (optional background)). Since  $L_B$  is real symmetric, its eigenvalues are real and its smallest eigenvalue satisfies

$$\lambda_1 = \min_{f \neq 0} \frac{\langle L_B f, f \rangle}{\langle f, f \rangle} = \min_{f \neq 0} \frac{\mathcal{E}(f)}{\|f\|_2^2}.$$

We will not use calculus of variations for the positivity/simplicity conclusions; instead we follow the five-step spine.

### 3 Step 1: Energy decomposition

**Proposition 10** (Energy decomposition (**Standard**)). *For every  $f \in \mathbb{R}^I$ ,*

$$\mathcal{E}(f) = \frac{1}{2} \sum_{\{x, y\} \in E} w_{xy} (\tilde{f}(x) - \tilde{f}(y))^2. \quad (3)$$

*In particular,  $\mathcal{E}(f) = 0$  if and only if  $\tilde{f}(x) = \tilde{f}(y)$  for every edge  $\{x, y\} \in E$ .*

*Proof.* This is immediate from Definition 6 with  $g = f$ , and each summand is nonnegative.  $\square$

### 4 Step 2: Markov property (normal contractions) and positivity preservation

**Definition 11** (Normal contraction). A function  $\Phi : \mathbb{R} \rightarrow \mathbb{R}$  is a *normal contraction* if  $\Phi(0) = 0$  and

$$|\Phi(a) - \Phi(b)| \leq |a - b| \quad \text{for all } a, b \in \mathbb{R}.$$

**Lemma 12** (Markov property (**Standard**)). *For any normal contraction  $\Phi$  and any  $f \in \mathbb{R}^I$ ,*

$$\mathcal{E}(\Phi \circ f) \leq \mathcal{E}(f),$$

*where  $(\Phi \circ f)(x) := \Phi(f(x))$  on  $I$  and is extended by 0 on  $B$  (consistent with  $\Phi(0) = 0$ ).*

*Proof.* For each edge  $\{x, y\} \in E$ ,

$$(\Phi(\tilde{f}(x)) - \Phi(\tilde{f}(y)))^2 \leq (\tilde{f}(x) - \tilde{f}(y))^2$$

by the 1-Lipschitz property. Multiply by  $w_{xy}$ , sum over edges, and use (3).  $\square$

*Remark 13* (How Step 2 is used (**New presentation**)). Lemma 12 is a standard “Markovianity” property of the Dirichlet form (see, e.g., [9, 3] for general Dirichlet forms and for graph Laplacians). In the present finite-graph proof, we will not use Lemma 12 directly; instead we use the closely related and equally standard consequence that the resolvent  $(L_B + \mu \text{Id})^{-1}$  is positivity preserving (Lemma 16).

**Definition 14** (Positive and negative parts). For  $u \in \mathbb{R}^I$ , define  $u^+(x) := \max\{u(x), 0\}$  and  $u^-(x) := \max\{-u(x), 0\}$  so that  $u = u^+ - u^-$  and  $u^+ u^- = 0$  pointwise.

**Lemma 15** (Scalar inequality for the negative part (**Standard**)). For all  $a, b \in \mathbb{R}$ , writing  $t^- := \max\{-t, 0\}$ , one has

$$(a - b)(a^- - b^-) \leq -(a^- - b^-)^2. \quad (4)$$

*Proof.* We distinguish cases.

(i) If  $a \geq 0$  and  $b \geq 0$ , then  $a^- = b^- = 0$ , so both sides are 0.

(ii) If  $a \leq 0$  and  $b \leq 0$ , then  $a^- = -a$  and  $b^- = -b$ , so the left-hand side equals

$$(a - b)((-a) - (-b)) = (a - b)(-a + b) = -(a - b)^2 = -(a^- - b^-)^2.$$

(iii) If  $a \geq 0$  and  $b \leq 0$ , then  $a^- = 0$  and  $b^- = -b$ , so  $a^- - b^- = b$  and

$$(a - b)(a^- - b^-) = (a - b)b = ab - b^2 \leq -b^2 = -(a^- - b^-)^2,$$

since  $ab \leq 0$ . The case  $a \leq 0 \leq b$  is symmetric.  $\square$

**Lemma 16** (Resolvent exists and is positivity preserving (**Standard**)). Fix  $\mu > 0$ . For every  $f \in \mathbb{R}^I$  there is a unique  $u \in \mathbb{R}^I$  solving

$$(L_B + \mu \text{Id})u = f. \quad (5)$$

Moreover, if  $f \geq 0$  (pointwise on  $I$ ), then  $u \geq 0$ .

*Proof.* Existence and uniqueness follow since  $L_B + \mu \text{Id}$  is a real symmetric matrix with

$$\langle (L_B + \mu \text{Id})u, u \rangle = \mathcal{E}(u) + \mu \|u\|_2^2 \geq \mu \|u\|_2^2,$$

so it is positive definite and hence invertible.

Assume  $f \geq 0$  and let  $u$  solve (5). Take inner product with  $u^-$ :

$$\langle (L_B + \mu \text{Id})u, u^- \rangle = \langle f, u^- \rangle \geq 0.$$

Expand the left side using  $u = u^+ - u^-$  and Lemma 7:

$$\langle L_B u, u^- \rangle + \mu \langle u, u^- \rangle = \mathcal{E}(u, u^-) + \mu (\langle u^+, u^- \rangle - \|u^-\|_2^2) = \mathcal{E}(u, u^-) - \mu \|u^-\|_2^2$$

since  $\langle u^+, u^- \rangle = 0$ .

We claim  $\mathcal{E}(u, u^-) \leq 0$ . Write the edge expansion:

$$\mathcal{E}(u, u^-) = \frac{1}{2} \sum_{\{x, y\} \in E} w_{xy} (\tilde{u}(x) - \tilde{u}(y))(\widetilde{u^-}(x) - \widetilde{u^-}(y)).$$

By Lemma 15, for scalars  $a, b \in \mathbb{R}$  the inequality (4) holds. Applying (4) with  $a = \tilde{u}(x)$ ,  $b = \tilde{u}(y)$ , multiplying by  $w_{xy}$ , and summing gives

$$\mathcal{E}(u, u^-) \leq -\mathcal{E}(u^-, u^-) \leq 0.$$

Therefore,

$$0 \leq \langle (L_B + \mu \text{Id})u, u^- \rangle \leq -\mathcal{E}(u^-) - \mu \|u^-\|_2^2 \leq -\mu \|u^-\|_2^2,$$

which forces  $\|u^-\|_2 = 0$ , hence  $u^- \equiv 0$  and  $u \geq 0$ .  $\square$

## 5 A coercivity lemma (positivity of the first eigenvalue)

The following is the Dirichlet analogue of “constants are the only zero-energy functions, and the boundary kills constants.”

**Proposition 17** (Zero energy forces constancy on components (**Standard**)). *If  $f \in \mathbb{R}^I$  satisfies  $\mathcal{E}(f) = 0$ , then  $\tilde{f}$  is constant on each connected component of  $G$ .*

*Proof.* If  $\mathcal{E}(f) = 0$ , then by Proposition 10 every edge term satisfies  $\tilde{f}(x) = \tilde{f}(y)$ . Along any path, values propagate, so  $\tilde{f}$  is constant on each connected component.  $\square$

**Corollary 18** (Dirichlet coercivity;  $L_B$  is positive definite (**Standard**)). *Assume  $G$  is connected and  $B \neq \emptyset$  and  $I \neq \emptyset$ . If  $f \in \mathbb{R}^I$  satisfies  $\mathcal{E}(f) = 0$ , then  $f \equiv 0$ . Equivalently,  $L_B$  is positive definite and all its eigenvalues are strictly positive.*

*Proof.* By Proposition 17,  $\tilde{f}$  is constant on connected  $G$ . Since  $\tilde{f}|_B = 0$  and  $B \neq \emptyset$ , the constant must be 0, hence  $f \equiv 0$  on  $I$ . Now  $\langle L_B f, f \rangle = \mathcal{E}(f)$  (Lemma 7) implies  $\langle L_B f, f \rangle > 0$  for all  $f \neq 0$ , so  $L_B$  is positive definite.  $\square$

*Remark 19* (Where connectivity is used (**New presentation**)). Corollary 18 (and thus Theorem 4(a)) uses only that  $G$  is connected and that the boundary is nonempty: it ensures that the only globally constant zero-extension is the trivial one. By contrast, the *interior connectivity* hypothesis (Definition 1(iii)) is used only in Step 3 to obtain *positivity improvement* (Proposition 20) and hence simplicity/strict positivity of the ground state.

## 6 Step 3: Irreducibility (connected interior) and positivity improvement

### 6.1 A strong maximum principle adapted to the interior

**Proposition 20** (Strong maximum principle for the resolvent on a connected interior (**Standard**)). *Assume the induced interior graph  $G_I$  is connected. Fix  $\mu > 0$  and let  $u$  solve  $(L_B + \mu \text{Id})u = f$ . If  $f \geq 0$  and  $f \not\equiv 0$ , then  $u(x) > 0$  for every  $x \in I$ .*

*Proof.* By Lemma 16,  $u \geq 0$ . Suppose for contradiction that there exists  $x_0 \in I$  with  $u(x_0) = 0$ . Evaluate the equation at  $x_0$ :

$$f(x_0) = (L_B u)(x_0) + \mu u(x_0) = \sum_{y \in V} w_{x_0 y} (u(x_0) - u(y)) = - \sum_{y \in V} w_{x_0 y} u(y) \leq 0,$$

since all  $u(y) \geq 0$  and weights are nonnegative. But  $f(x_0) \geq 0$ , hence  $f(x_0) = 0$  and

$$\sum_{y \in V} w_{x_0 y} u(y) = 0.$$

Every term in the sum is nonnegative, and  $w_{x_0 y} > 0$  whenever  $\{x_0, y\} \in E$ , so we deduce

$$u(y) = 0 \quad \text{for every neighbor } y \text{ of } x_0.$$

Now restrict to neighbors  $y \in I$  (interior neighbors). The same argument shows that every interior neighbor of  $x_0$  also has value 0. Iterating along any path in the *interior graph*  $G_I$  shows that  $u \equiv 0$  on the entire connected interior  $I$ .

Finally,  $(L_B + \mu \text{Id})u = f$  then forces  $f \equiv 0$ , contradicting  $f \not\equiv 0$ . Therefore no such  $x_0$  exists and  $u(x) > 0$  for all  $x \in I$ .  $\square$

*Remark 21* (Background: discrete maximum principles **(Standard)**). The argument in Proposition 20 is the standard discrete maximum principle for graph Laplacians with Dirichlet boundary; see, for example, [4, Ch. 1] or [3, Ch. 3].

**Corollary 22** (The resolvent matrix is strictly positive **(Standard)**). *Assume  $G_I$  is connected. Let  $I = \{1, \dots, n\}$  index interior vertices. For  $\mu > 0$ , define*

$$R_\mu := (L_B + \mu \text{Id})^{-1} \in \mathbb{R}^{n \times n}.$$

*Then every entry of  $R_\mu$  is strictly positive:  $(R_\mu)_{ij} > 0$  for all  $i, j$ .*

*Proof.* The  $j$ th column of  $R_\mu$  is  $u = R_\mu e_j$ , where  $e_j \geq 0$  and  $e_j \not\equiv 0$ . By Proposition 20,  $u(i) > 0$  for all  $i$ , i.e.  $(R_\mu)_{ij} > 0$  for all  $i$ .  $\square$

*Remark 23* (Componentwise version without interior connectivity). If  $G_I$  has connected components  $I = \bigsqcup_\alpha I_\alpha$ , then the same proof shows: for  $f \geq 0$  and  $u = (L_B + \mu \text{Id})^{-1} f$ , one has  $u \geq 0$  and  $u > 0$  on each component  $I_\alpha$  where  $f \not\equiv 0$ . Equivalently, after reordering vertices,  $R_\mu$  is block diagonal with strictly positive blocks corresponding to the  $I_\alpha$ . This is the correct general form of “positivity improvement.”

## 7 Step 4: Compact resolvent (finite-dimensional)

**Lemma 24** (Compactness is automatic **(Standard)**). *In finite dimensions, every linear operator is bounded and maps bounded sets to relatively compact sets. In particular, the resolvent  $R_\mu = (L_B + \mu \text{Id})^{-1}$  is a compact operator on  $\ell^2(I)$ .*

*Proof.* In  $\mathbb{R}^n$ , bounded sets have compact closure (Heine–Borel; see, e.g., [8]). Since  $R_\mu$  is linear and continuous, it maps bounded sets to bounded sets, hence to relatively compact sets.  $\square$

## 8 Step 5: Perron–Frobenius and the ground state

We apply Perron–Frobenius to the strictly positive, compact operator  $R_\mu$ . Because  $L_B$  is symmetric,  $R_\mu$  is also symmetric, and eigenvalues are diagonalizable; thus geometric simplicity implies algebraic simplicity automatically in our application.

### 8.1 Perron–Frobenius for strictly positive matrices (finite-dimensional)

**Theorem 25** (Perron–Frobenius for strictly positive matrices (**Standard**)). *Let  $A \in \mathbb{R}^{n \times n}$  have strictly positive entries:  $A_{ij} > 0$  for all  $i, j$ . Then:*

- (i) *The spectral radius  $\rho(A)$  is an eigenvalue of  $A$ .*
- (ii) *There exists an eigenvector  $v \in \mathbb{R}^n$  with  $v_i > 0$  for all  $i$  and  $Av = \rho(A)v$ .*
- (iii) *The eigenspace for  $\rho(A)$  is one-dimensional (geometric simplicity).*
- (iv) *If  $w \geq 0$  and  $w \neq 0$ , then  $Aw$  has strictly positive entries.*

*Remark 26.* Part (iv) is immediate from  $A_{ij} > 0$ . Parts (i)–(iii) are proved in Appendix A. Standard references include [2, 1, 6].

### 8.2 From Perron–Frobenius to the Dirichlet ground state

**Proposition 27** (Ground state simplicity and positivity (**Standard**)). *Assume  $G_I$  is connected and fix  $\mu > 0$ . Let  $R_\mu = (L_B + \mu \text{Id})^{-1}$ . Then  $R_\mu$  has a unique (up to scaling) strictly positive eigenvector  $\varphi$  associated to its spectral radius  $\rho(R_\mu)$ . Moreover,  $\varphi$  is an eigenvector of  $L_B$  for the smallest eigenvalue*

$$\lambda_1 = \rho(R_\mu)^{-1} - \mu,$$

*and  $\lambda_1$  is simple.*

*Proof.* By Corollary 22,  $R_\mu$  has strictly positive entries, so Theorem 25 applies. Hence  $\rho(R_\mu)$  is an eigenvalue with a strictly positive eigenvector  $\varphi$ , and its eigenspace is one-dimensional.

If  $L_B \varphi = \lambda \varphi$ , then  $(L_B + \mu \text{Id})\varphi = (\lambda + \mu)\varphi$ , hence  $R_\mu \varphi = (\lambda + \mu)^{-1} \varphi$ . Thus eigenvalues of  $R_\mu$  are exactly  $(\lambda + \mu)^{-1}$  where  $\lambda$  ranges over eigenvalues of  $L_B$ . The largest eigenvalue of  $R_\mu$  is  $\rho(R_\mu)$ , so it corresponds to the smallest eigenvalue  $\lambda_1$  of  $L_B$  via  $\rho(R_\mu) = (\lambda_1 + \mu)^{-1}$ , i.e.  $\lambda_1 = \rho(R_\mu)^{-1} - \mu$ .

Since  $R_\mu$  is symmetric, it is diagonalizable. Therefore one-dimensionality of the  $\rho(R_\mu)$ -eigenspace implies that  $\rho(R_\mu)$  has algebraic multiplicity one. The eigenspace correspondence between  $L_B$  and  $R_\mu$  preserves dimensions, so  $\lambda_1$  is also simple (geometric and algebraic multiplicity one).  $\square$

*Proof of Theorem 4.* (a) By Corollary 18,  $L_B$  is positive definite, hence  $\lambda_1 > 0$ .

(b)–(c) By Proposition 27,  $\lambda_1$  is simple and has an eigenfunction  $\varphi_1 > 0$  on  $I$ .

(d) Suppose  $R : V \rightarrow V$  is an involution preserving  $E, w$  and  $B$ . Then  $R$  maps  $I$  to itself and induces a linear operator  $(Uf)(x) := f(Rx)$  on  $\mathbb{R}^I$ . A direct check from (1) shows  $UL_B = L_B U$ , hence also  $UR_\mu = R_\mu U$ . Let  $\varphi$  span the  $\rho(R_\mu)$ -eigenspace and satisfy  $\varphi > 0$  on  $I$ . Then  $U\varphi$  is also a  $\rho(R_\mu)$ -eigenvector. By simplicity,  $U\varphi = c\varphi$  for some scalar  $c$ . Since  $U^2 = \text{Id}$ , we have  $c^2 = 1$ , so  $c = \pm 1$ . But  $\varphi > 0$  and  $U\varphi > 0$ , so  $c = -1$  is impossible. Therefore  $c = +1$ , i.e.  $\varphi \circ R = \varphi$  on  $I$ .  $\square$



## 9 Worked example (optional for exposition)

*Remark 28* (Path graph). Take the path graph on vertices  $\{0, 1, \dots, m\}$  with boundary  $B = \{0, m\}$  and unit weights. Then  $I = \{1, \dots, m-1\}$  is connected, and  $L_B$  is the familiar tridiagonal matrix with 2 on the diagonal and  $-1$  on off-diagonals. Theorem 4 recovers the classical fact that the first Dirichlet mode is strictly positive and unique up to scale.

## A A proof of Perron–Frobenius for strictly positive matrices

This appendix proves Theorem 25(i)–(iii) in a form sufficient for the main text. Throughout,  $A \in \mathbb{R}^{n \times n}$  has  $A_{ij} > 0$  for all  $i, j$ .

### A.1 Existence of a positive eigenvector

Let

$$\Delta := \left\{ x \in \mathbb{R}^n : x_i \geq 0, \sum_{i=1}^n x_i = 1 \right\}$$

be the standard simplex (compact and convex). Define the continuous map  $T : \Delta \rightarrow \Delta$  by

$$T(x) := \frac{Ax}{\mathbf{1}^\top Ax}.$$

Since  $A$  has strictly positive entries,  $Ax$  has strictly positive coordinates for every  $x \in \Delta$ , and  $\mathbf{1}^\top Ax > 0$ . Thus  $T$  is well-defined and continuous, and indeed  $T(\Delta) \subseteq \Delta$ .

**Lemma 29** (Brouwer fixed point  $\Rightarrow$  positive eigenvector (**Standard**)). *There exists  $v \in \Delta$  and  $\lambda > 0$  such that  $Av = \lambda v$  and  $v_i > 0$  for all  $i$ .*

*Proof.* By Brouwer’s fixed point theorem (see, e.g., [7]),  $T$  has a fixed point  $v \in \Delta$ :  $T(v) = v$ . Then  $Av = (\mathbf{1}^\top Av) v$ , so  $Av = \lambda v$  with  $\lambda := \mathbf{1}^\top Av > 0$ . Since  $A > 0$  and  $v \neq 0$ , all coordinates of  $Av$  are positive, hence all coordinates of  $v$  are positive as well.  $\square$

### A.2 The eigenvalue equals the spectral radius

For a fixed vector  $v > 0$ , define the weighted sup norm

$$\|x\|_v := \max_{1 \leq i \leq n} \frac{|x_i|}{v_i}.$$

This is a genuine norm on  $\mathbb{R}^n$ .

**Lemma 30** (If  $Av = \lambda v$  with  $v > 0$ , then  $\rho(A) = \lambda$  (**Standard**)). *Let  $v > 0$  and  $\lambda > 0$  satisfy  $Av = \lambda v$ . Then  $\rho(A) = \lambda$ .*

*Proof.* First,  $\rho(A) \geq |\lambda| = \lambda$  since  $\lambda$  is an eigenvalue.

For the reverse inequality, let  $x \in \mathbb{R}^n$  and set  $t := \|x\|_v$ . Then  $|x_i| \leq tv_i$  for all  $i$ , i.e.  $|x| \leq tv$  entrywise. Since  $A$  has nonnegative entries, this implies

$$|Ax| \leq A|x| \leq A(tv) = tAv = t\lambda v$$

entrywise. Dividing by  $v_i$  and taking maxima gives  $\|Ax\|_v \leq \lambda \|x\|_v$ , hence  $\|A\|_{v \rightarrow v} \leq \lambda$  as an operator norm. A standard fact from linear algebra (see, e.g., [6, 5]) is that for any induced

operator norm  $\|\cdot\|$ , every eigenvalue  $\mu$  of  $A$  satisfies  $|\mu| \leq \|A\|$  (apply  $\|Ax\| \leq \|A\| \|x\|$  to an eigenvector). Consequently,

$$\rho(A) \leq \|A\|_{v \rightarrow v} \leq \lambda.$$

Therefore  $\rho(A) = \lambda$ . □

Together, Lemmas 29 and 30 give Theorem 25(i)–(ii).

### A.3 Geometric simplicity

**Lemma 31** (Uniqueness of the positive eigenvector (**Standard**)). *If  $Av = \rho(A)v$  and  $Aw = \rho(A)w$  with  $v > 0$  and  $w > 0$ , then  $w$  is a positive scalar multiple of  $v$ .*

*Proof.* Define

$$t_* := \inf\{t > 0 : w \leq tv \text{ entrywise}\}.$$

Since  $v > 0$  and  $w > 0$ , the set is nonempty and  $t_* \in (0, \infty)$ . By definition,  $w \leq t_*v$  and there exists an index  $i_0$  with  $w_{i_0} = t_*v_{i_0}$ .

If  $w \neq t_*v$ , then there exists  $j$  with  $w_j < t_*v_j$ . Because  $A > 0$ , strict inequality in one coordinate forces strict inequality after applying  $A$  in *every* coordinate:

$$(Aw)_i = \sum_j A_{ij}w_j < \sum_j A_{ij}(t_*v_j) = (t_*Av)_i \quad \text{for all } i,$$

hence  $Aw < t_*Av$  entrywise. Using  $Aw = \rho(A)w$  and  $Av = \rho(A)v$ , this becomes  $w < t_*v$  entrywise, contradicting the definition of  $t_*$  (we could then decrease  $t$  slightly and still have  $w \leq tv$ ). Therefore  $w = t_*v$ . □

**Lemma 32** (Any  $\rho(A)$ -eigenvector has constant sign (**Standard**)). *If  $Ay = \rho(A)y$  and  $y \neq 0$ , then either  $y \geq 0$  entrywise or  $y \leq 0$  entrywise.*

*Proof.* Suppose  $y$  has both positive and negative coordinates, and set  $x := |y| > 0$  entrywise. Then by positivity of  $A$ ,

$$Ax = A|y| \geq |Ay| = |\rho(A)y| = \rho(A)|y| = \rho(A)x$$

entrywise. Because  $y$  has mixed signs and  $A > 0$ , the triangle inequality is strict in every coordinate: for each  $i$ ,

$$(A|y|)_i = \sum_j A_{ij}|y_j| > \left| \sum_j A_{ij}y_j \right| = |(Ay)_i|,$$

so  $Ax > \rho(A)x$  entrywise.

Now apply  $A$  once more: since  $Ax - \rho(A)x \geq 0$  is nonzero and  $A > 0$ , we have

$$A(Ax - \rho(A)x) > 0 \quad \Rightarrow \quad A^2x > \rho(A)Ax \geq \rho(A) \cdot \rho(A)x = \rho(A)^2x$$

entrywise. In particular, letting

$$m := \min_{1 \leq i \leq n} \frac{(A^2x)_i}{x_i},$$

we have  $m > \rho(A)^2$ .

On the other hand, for any matrix  $M$  and any  $x > 0$ , the inequality  $Mx \geq mx$  implies  $\rho(M) \geq m$  by iterating  $Mx \geq mx$  and taking norms (a one-line argument identical to the standard lower Collatz–Wielandt bound). Applying this to  $M = A^2$  yields

$$\rho(A^2) \geq m > \rho(A)^2,$$

contradicting  $\rho(A^2) = \rho(A)^2$ . Therefore  $y$  cannot have mixed signs, proving the claim. □

**Lemma 33** (Geometric simplicity of  $\rho(A)$  (**Standard**)). *The eigenspace for  $\rho(A)$  is one-dimensional.*

*Proof.* By Lemma 29 and Lemma 30, there exists  $v > 0$  with  $Av = \rho(A)v$ . Let  $y$  satisfy  $Ay = \rho(A)y$ . By Lemma 32, either  $y \geq 0$  or  $y \leq 0$ . Replacing  $y$  by  $-y$  if necessary, we may assume  $y \geq 0$  and  $y \neq 0$ .

Because  $A > 0$  and  $y \geq 0$ , we have  $Ay > 0$ , hence  $y$  must in fact satisfy  $y > 0$  (otherwise a zero coordinate would force a zero coordinate in  $Ay = \rho(A)y$ , impossible since  $Ay > 0$ ). Thus  $y > 0$ , and Lemma 31 implies  $y$  is a scalar multiple of  $v$ . Therefore the eigenspace is one-dimensional.  $\square$

Lemmas 29, 30, and 33 yield Theorem 25(i)–(iii).

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