

## HARDY'S INEQUALITY ON FINITE NETWORKS

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ABSTRACT. The smallest eigenvalue of a weighted discrete Laplacian is closely related to a generalized Hardy's inequality on networks. We shall estimate the smallest eigenvalue by using a discrete Kuramochi potential with some numerical experiments.

### 1. PROBLEM SETTING

Let  $X$  be a finite set of nodes,  $Y$  be a finite set of arcs and  $K$  be the node-arc incidence matrix. Assume that the graph  $G := \{X, Y, K\}$  is connected and has no self-loop. For every two nodes  $a, b \in X$ , denote by  $\rho(a, b)$  the geodesic distance between  $a$  and  $b$ , i.e., the minimum number of arcs in the paths between  $a$  and  $b$ .

For a strictly positive real-valued function  $r$ ,  $N := \{G, r\}$  is called a network. Denote by  $L(X)$  the set of all real valued functions on  $X$ , by  $L^+(X)$  the set of all nonnegative  $u \in L(X)$ .

For  $u \in L(X)$ , the discrete derivative  $du$ , the discrete Laplacian  $\Delta u(x)$  and the Dirichlet sum  $D(u)$  of  $u$  on  $N$  are defined by

$$\begin{aligned} du(y) &:= -r(y)^{-1} \sum_{x \in X} K(x, y)u(x), \\ \Delta u(x) &:= \sum_{y \in Y} K(x, y)[du(y)], \\ D(u) &:= \sum_{y \in Y} r(y)[du(y)]^2. \end{aligned}$$

The mutual Dirichlet sum  $D(u, v)$  of  $u, v \in L(X)$  is defined by

$$D(u, v) := \sum_{y \in Y} r(y)[du(y)][dv(y)].$$

Let  $A_0$  be a nonempty subset of  $X$  such that  $X \setminus A_0$  is connected and let  $m \in L(X)$  satisfy  $m(x) = 0$  on  $A_0$  and  $m(x) > 0$  on  $X \setminus A_0$ .

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A generalized Hardy's inequality is to find the best possible constant  $C_m > 0$  such that

$$\sum_{x \in X} m(x)u(x)^2 \leq C_m D(u)$$

for all  $u \in L(X)$  such that  $u(x) = 0$  on  $A_0$ .

By special choices of  $N, A_0$  and  $m$ , we obtain Wirtinger's inequality and Hardy's inequality in [2] and [3]. We shall show that  $1/C_m$  is equal to the smallest eigenvalue of an eigenvalue problem. We shall also give an estimation of this value by using a discrete Kuramochi potential studied in [4] and [5].

## 2. MINIMUM EIGENVALUE

Let us put

$$L(X; A_0) := \{u \in L(X); u = 0 \text{ on } A_0\}.$$

For simplicity, let us put

$$\begin{aligned} ((u, v))_m &:= \sum_{x \in X} m(x)u(x)v(x), \\ \|u\|_m &:= [((u, u))_m]^{1/2}, \\ \chi_m(u) &:= \frac{D(u)}{\|u\|_m^2}. \end{aligned}$$

We shall consider the extremum problem ( $H_m$ ):

$$\begin{aligned} \text{Find } \lambda_m &:= \inf\{\chi_m(u); u \in L(X; A_0)\} \\ &= \inf\{D(u); u \in L(X; A_0), \|u\|_m = 1\}. \end{aligned}$$

**Proposition 2.1.** *There exists an optimal solution  $\tilde{u}$  of problem ( $H_m$ ), i.e.,  $\lambda_m = D(\tilde{u})$ ,  $\tilde{u} \in L(X; A_0)$  and  $\|\tilde{u}\|_m = 1$ .*

*Proof.* Let  $\{v_k\}$  be a sequence in  $L(X; A_0)$  such that  $\chi_m(v_k) \rightarrow \lambda_m$  as  $k \rightarrow \infty$ . Put  $u_k = v_k/\|v_k\|_m$ . Then  $\|u_k\|_m = 1$  and

$$\chi_m(u_k) = D(u_k) = D(v_k)/\|v_k\|_m^2 = \chi_m(v_k).$$

Since  $\{u_k(x)\}$  is bounded for each  $x \in X$ , we may assume that  $\{u_k\}$  converges pointwise to a function  $\tilde{u} \in L(X; A_0)$ . We have  $\|\tilde{u}\|_m = 1$  and

$$\lim_{k \rightarrow \infty} D(u_k) = D(\tilde{u}),$$

so that  $\chi_m(\tilde{u}) = \lambda_m$ . □

Denote by  $S(\lambda_m)$  be the set of all optimal solutions of problem ( $H_m$ ), i.e.,

$$S(\lambda_m) := \{u \in L(X; A_0); \chi_m(u) = \lambda_m\}.$$

Consider the following eigenvalue problem of finding a number  $\mu$  and a nonzero function  $u \in L(X; A_0)$  which satisfy

$$(Eig) \quad \Delta u(x) = -\mu m(x)u(x) \text{ on } X \setminus A_0.$$

Denote by  $E_m(\Delta)$  the set of all  $\mu$  satisfying (Eig) and by  $EV_m(\mu)$  the set of nonzero functions  $u$  satisfying (Eig) with  $\mu \in E_m(\Delta)$ .

For every  $u \in EV_m(\mu)$ , we have

$$\begin{aligned} D(u) &= -\sum_{x \in X} [\Delta u(x)]u(x) \\ &= \mu \sum_{x \in X} m(x)u(x)^2 = \mu \|u\|_m^2. \end{aligned}$$

Since  $D(u)$  is positive definite on the set  $L(X; A_0)$ , we see that  $E_m(\Delta)$  consists of positive real numbers.

By the above observation, we have

**Proposition 2.2.**  $\lambda_m = \min\{\mu; \mu \in E_m(\Delta)\}$ .

**Lemma 2.1.**  $S(\lambda_m) = EV_m(\lambda_m)$ .

Proof. By the above observation, it suffices to show  $S(\lambda_m) \subset EV_m(\lambda_m)$ . Let  $u \in S(\lambda_m)$ . Denote by  $\varepsilon_x \in L(X)$  the characteristic function of the set  $\{x\}$ . For any real number  $t$  and  $x \in X \setminus A_0$ , we have

$$\lambda_m = \chi_m(u) \leq \chi_m(u + t\varepsilon_x),$$

or

$$\lambda_m \|u + t\varepsilon_x\|_m^2 \leq D(u + t\varepsilon_x).$$

Noting the relation

$$\begin{aligned} D(u + t\varepsilon_x) &= D(u) + 2tD(\tilde{u}, \varepsilon_x) + t^2D(\varepsilon_x), \\ \|u + t\varepsilon_x\|_m^2 &= \|u\|_m^2 + 2t((u, \varepsilon_x))_m + t^2\|\varepsilon_x\|_m^2, \end{aligned}$$

we obtain

$$D(u, \varepsilon_x) = \lambda_m((u, \varepsilon_x))_m.$$

Since  $D(u, \varepsilon_x) = -\Delta u(x)$  and  $((u, \varepsilon_x))_m = m(x)u(x)$ , we conclude that  $u \in EV_m(\lambda_m)$ .  $\square$

**Lemma 2.2.** Assume that  $u \in S(\lambda_m)$ . Then  $|u| \in S(\lambda_m)$  and  $u(x_1)u(x_2) \geq 0$  for every  $x_1, x_2 \in X \setminus A_0$  with  $\rho(x_1, x_2) = 1$ .

Proof. Let  $v = |u|$ . Then  $v \in L(X; A_0)$  and  $D(v) \leq D(u)$  holds (cf.[9]). Since  $\|v\|_m = \|u\|_m$ , we have

$$\lambda_m \leq \chi_m(v) \leq \chi_m(u) = \lambda_m,$$

and hence  $v \in S(\lambda_m)$ . Suppose that there exist  $x_1, x_2 \in X \setminus A_0$  such that  $\rho(x_1, x_2) = 1$  and  $u(x_1)u(x_2) < 0$ . Let  $y' \in Y$  be an arc whose endpoints are  $x_1$  and  $x_2$ . Then

$$\begin{aligned} |dv(y')| &= r(y')^{-1}|v(x_1) - v(x_2)| \\ &< r(y')^{-1}|u(x_1) - u(x_2)| = |du(y')|, \end{aligned}$$

so that  $D(v) < D(u)$ . Thus  $\lambda_m = \chi_m(v) < \chi_m(u) = \lambda_m$ . This is a contradiction.  $\square$

**Corollary 2.1.** If  $u \in S(\lambda_m)$ , then either  $u = |u|$  or  $u = -|u|$ .

**Lemma 2.3.** If  $u \in S(\lambda_m)$  is non-negative, then  $u(x) > 0$  on  $X \setminus A_0$ .

Proof. Let  $u \in S(\lambda_m)$  be nonnegative. By Lemma 2.1,

$$\Delta u(x) = -\lambda_m m(x)u(x) \leq 0 \quad \text{on } X \setminus A_0.$$

Namely  $u$  is superharmonic on  $X \setminus A_0$ . By the minimum principle (cf. [9]), we have  $u(x) > 0$  on  $X \setminus A_0$ .  $\square$

**Corollary 2.2.** *If  $u \in S(\lambda_m)$ , then either  $\Delta u(x) < 0$  on  $X \setminus A_0$  or  $\Delta u(x) > 0$  on  $X \setminus A_0$ .*

**Lemma 2.4.** *The dimension of  $EV_m(\lambda_m)$  is one. Namely, if  $u_1, u_2 \in EV_m(\lambda_m)$ , then  $u_1$  and  $u_2$  are proportional.*

Proof. Assume that there exist  $u_1, u_2 \in EV_m(\lambda_m)$  such that they are not proportional. Choose numbers  $\alpha$  and  $\beta$  such that  $|\alpha| + |\beta| > 0$  and  $\alpha u_1(x_1) + \beta u_2(x_1) = 0$  for some  $x_1 \in X \setminus A_0$ . Let  $u = \alpha u_1 + \beta u_2$ . Then  $u \neq 0$ , since  $u_1$  and  $u_2$  are not proportional. We have

$$\begin{aligned} \Delta u(x) &= \alpha \Delta u_1(x) + \beta \Delta u_2(x) \\ &= -\lambda_m m(x)u_1(x) - \lambda_m m(x)u_2(x) \\ &= -\lambda_m m(x)u(x). \end{aligned}$$

Namely  $u \in EV_m(\lambda_m) = S(\lambda_m)$ . We have

$$\Delta u(x_1) = \lambda_m m(x)u(x_1) = 0.$$

This contradicts Corollary 2.2.  $\square$

Summing up these results, we obtain

**Theorem 2.1.** *There exists a unique  $\tilde{u} \in L(X; A_0)$  such that*

- (1)  $\lambda_m = D(\tilde{u})$  and  $\|\tilde{u}\|_m = 1$ ;
- (2)  $\tilde{u}(x) > 0$  on  $X \setminus A_0$ ;
- (3)  $\Delta \tilde{u}(x) = -\lambda_m m(x)\tilde{u}(x)$  on  $X \setminus A_0$ .

### 3. ESTIMATION OF $\lambda_m$

Let us put

$$D(N; A_0) := \{u \in L(X; A_0); D(u) < \infty\}.$$

Since  $N$  is a finite network, we see that  $D(N; A_0) = L(X; A_0)$ . Notice that  $D(N; A_0)$  is a Hilbert space with the inner product  $D(u, v)$  (cf. [9]).

The Kuramochi function  $\tilde{g}_x$  of  $N$  with pole at  $x \in X \setminus A_0$  is defined by the reproducing property:

$$u(x) = D(u, \tilde{g}_x) \quad \text{for all } u \in D(N; A_0)$$

(cf. [4]). For each nonempty subset  $B$  of  $X \setminus A_0$ , let us put

$$d(A_0, B) := \inf\{D(u); u \in D(N; A_0), u = 1 \text{ on } B\}.$$

We have

**Lemma 3.1.**  $\tilde{g}_x$  has the following properties:

- (1)  $\tilde{g}_x(z) = 0$  on  $A_0$ ;
- (2)  $0 \leq \tilde{g}_x \leq \tilde{g}_x(x)$  on  $X$ ;
- (3)  $\Delta \tilde{g}_x(z) = -\varepsilon_x(z)$  on  $X \setminus A_0$ .
- (4)  $d(A_0, \{x\}) = 1/\tilde{g}_x(x)$ .

Now we shall estimate the value of  $\lambda_m$ . Our idea is to use the discrete Kuramochi function studied in [4] and [5]. A similar idea can be founded in [8] to estimate Lyapunov's inequality.

The Kuramochi potential  $\tilde{G}m(x)$  of  $m$  is defined by

$$\tilde{G}m(x) := \sum_{z \in X} \tilde{g}_x(z)m(z).$$

**Lemma 3.2.** Let  $\tilde{u}$  be as in Theorem 2.1. Then

$$\tilde{u}(x) = \lambda_m \sum_{z \in X} m(z)[\tilde{u}(z)]\tilde{g}_x(z).$$

Proof. By the reproducing property of the Kuramochi function and Lemma 3.1, we have

$$\begin{aligned} \tilde{u}(x) &= D(\tilde{u}, \tilde{g}_x) \\ &= - \sum_{z \in X} [\Delta \tilde{u}(z)]\tilde{g}_x(z) \\ &= \lambda_m \sum_{z \in X} m(z)[\tilde{u}(z)]\tilde{g}_x(z). \end{aligned}$$

□

**Theorem 3.1.** The following estimation holds:

$$\min\{\tilde{G}m(x); x \in X \setminus A_0\} \leq \frac{1}{\lambda_m} \leq \max\{\tilde{G}m(x); x \in X \setminus A_0\}.$$

Proof. Let  $\tilde{u}$  be as in Theorem 2.1. There exists  $b \in X \setminus A_0$  such that  $\tilde{u}(b) = \max\{\tilde{u}(x); x \in X\}$ . Then we have by Lemma 3.2

$$\begin{aligned} \tilde{u}(b) &= \lambda_m \sum_{z \in X} m(z)[\tilde{u}(z)]\tilde{g}_b(z) \\ &\leq \lambda_m \tilde{u}(b) \sum_{z \in X} \tilde{g}_b(z)m(z) \\ &= \lambda_m \tilde{u}(b)\tilde{G}m(b) \\ &\leq \lambda_m \tilde{u}(b) \max\{\tilde{G}m(x); x \in X \setminus A_0\}. \end{aligned}$$

We can prove the right hand side inequality similarly. □

**Theorem 3.2.** Let  $m(X) := \sum_{x \in X} m(x)$ . Then the following estimation holds:

$$\min\{d(A_0, \{x\}); x \in X \setminus A_0\} \leq m(X)\lambda_m \leq \max\{d(A_0, \{x\}); x \in X \setminus A_0\}.$$

Proof. Let  $\tilde{u}$  be as in Theorem 2.1. There exists  $b \in X \setminus A_0$  such that  $\tilde{u}(b) = \max\{\tilde{u}(x); x \in X\}$ . Then we have by Lemma 3.2

$$\begin{aligned}\tilde{u}(b) &= \lambda_m \sum_{z \in X} m(z) [\tilde{u}(z)] \tilde{g}_b(z) \\ &\leq \lambda_m \tilde{u}(b) \max\{\tilde{g}_x(x); x \in X \setminus A_0\} m(X) \\ &= \lambda_m \tilde{u}(b) \max\{1/d(A_0, \{x\}); x \in X \setminus A_0\}.\end{aligned}$$

□

#### 4. CLASSICAL HARDY'S INEQUALITY

In this section, we consider the following special finite network  $N = \{X, Y, K, r\}$  defined by:

$$X = \{x_0, x_1, \dots, x_n\}, Y = \{y_1, y_2, \dots, y_n\}$$

$$K(x_i, y_i) = 1, K(x_{i-1}, y_i) = -1 \text{ for } i = 1, 2, \dots, n$$

and  $K(x, y) = 0$  for any other pair.

Notice that the graph  $\{X, Y, K\}$  is a subgraph of the one-dimensional lattice domain  $\mathbf{Z}$ . For simplicity, we set

$$u_k := u(x_k), r_k := r(y_k), w_k = r_k^{-1}(u_k - u_{k-1}).$$

Then  $\Delta u(x_k) = w_{k+1} - w_k$  for  $1 \leq k \leq n-1$ ,  $\Delta u(x_0) = w_1$  and  $\Delta u(x_n) = -w_n$ . Furthermore

$$D(u) = \sum_{k=1}^n r_k^{-1} (u_k - u_{k-1})^2 = \sum_{k=1}^n r_k w_k^2.$$

We shall prove

**Theorem 4.1.** *Let  $A_0 := \{x_0\}$  and put  $R_k = \sum_{j=1}^k r_j$ . Then*

$$\sum_{k=1}^{\infty} r_k \left( \frac{u_k}{R_k} \right)^2 \leq 4D(u)$$

for every  $u \in L(X; A_0)$ .

Proof. Let us put  $v_k := u_k - u_{k-1}$  and  $\alpha_k := u_k/R_k$ . Then

$$\begin{aligned}r_k \alpha_k^2 - 2\alpha_k v_k &= r_k \alpha_k^2 - 2\alpha_k (\alpha_k R_k - \alpha_{k-1} R_{k-1}) \\ &= (r_k - 2R_k) \alpha_k^2 + 2R_{k-1} \alpha_k \alpha_{k-1} \\ &\leq (r_k - 2R_k) \alpha_k^2 + R_{k-1} (\alpha_k^2 + \alpha_{k-1}^2) \\ &= -R_k \alpha_k^2 + R_{k-1} \alpha_{k-1}^2.\end{aligned}$$

Since  $u_0 = 0$ , we have

$$\sum_{k=1}^n (r_k \alpha_k^2 - 2\alpha_k v_k) \leq \sum_{k=1}^n (-R_k \alpha_k^2 + R_{k-1} \alpha_{k-1}^2) = -R_n \alpha_n^2 \leq 0.$$

Therefore we have

$$\begin{aligned} \sum_{k=1}^n r_k \alpha_k^2 &\leq 2 \sum_{k=1}^n \alpha_k v_k \\ &\leq 2 \left[ \sum_{k=1}^n r_k \alpha_k^2 \right]^{1/2} \left[ \sum_{k=1}^n r_k^{-1} v_k^2 \right]^{1/2}, \end{aligned}$$

so that

$$\sum_{k=1}^n r_k \alpha_k^2 \leq 4 \sum_{k=1}^n r_k^{-1} v_k^2 = 4D(u).$$

□

**Corollary 4.1.** *Let  $A_0 = \{x_0\}$  and  $m_k = m(x_k) := \frac{r_k}{R_k^2}$ . Then  $\lambda_m \geq 1/4$ .*

**Corollary 4.2.** *Assume that  $A_0 = \{x_0\}$  and  $r_k = 1$  for all  $k$ . Then*

$$\sum_{k=1}^n \left( \frac{u_k}{\rho(x_0, x_k)} \right)^2 \leq 4 \sum_{k=1}^n (u_k - u_{k-1})^2$$

for all  $u_k$  ( $k = 0, 1, \dots, n$ ) with  $u_0 = 0$ .

Notice that  $\rho(x_0, x_k) = R_k$  and  $\rho(x_0, x_k) = k$  in Corollary 4.2, this inequality can be found in [2], page 239. We may expect that Corollary 4.2 would also hold in the general case. However it is not true as shown by Table 4 in the next section.

Hereafter in this section we always assume that  $A_0 = \{x_0\}$  and  $m(x_k) := r_k/R_k^2$ . In order to obtain the value  $\lambda_m$ , we calculate the minimum eigenvalue of (*Eig*) numerically:

$$\begin{aligned} -2u_1 + u_2 &= \mu m_1 u_1 \\ -2u_k + u_{k+1} + u_{k-1} &= \mu m_k u_k \quad \text{for } 2 \leq k \leq n-1 \\ -u_n + u_{n-1} &= \mu m_n u_n \end{aligned}$$

In order to study  $\lambda_m$  as a function of the size  $n$  of  $N$ , we denote it by  $\lambda(n) := \lambda_m(n)$ . Some numerical experiments are given in the next section.

In the present case, the Kuramochi function is given by

$$\tilde{g}_{x_k}(x_j) = \begin{cases} R_j & \text{for } 0 \leq j \leq k \\ R_k & \text{for } k < j \leq n \end{cases}$$

We estimate  $\lambda_m$  by using the Kuramochi potential  $\tilde{G}m$ :

$$\tilde{G}m(x_k) = \sum_{j=1}^k \frac{r_j}{R_j} + R_k \sum_{j=k+1}^n \frac{r_j}{R_j^2}.$$

It is easily seen that

$$\begin{aligned}\mu^*(n) &:= \max\{\tilde{G}m(x); x \in X \setminus A_0\} = \sum_{k=1}^n \frac{r_k}{R_k} \\ \mu_*(n) &:= \min\{\tilde{G}m(x); x \in X \setminus A_0\} = \sum_{k=1}^n \frac{r_k}{R_k^2}.\end{aligned}$$

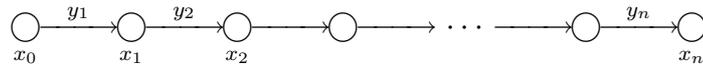
By Theorem 3.1, we have

$$\frac{1}{\mu^*(n)} \leq \lambda(n) \leq \frac{1}{\mu_*(n)}$$

Some numerical experiments for these quantities are also given in the next section.

## 5. NUMERICAL EXPERIMENTS

Let  $G = \{X, Y, K\}$  be the same graph as in Section 4. The graph can be drawn as follows:



We take  $A_0 = \{x_0\}$  and  $m(x_k) := r_k/R_k^2$  except in Table 4.

**Table 1:** The case where  $r_k = 1$  for all  $k$ .

$n$	$\lambda(n)$	$1/\mu^*(n)$	$1/\mu_*(n)$
10	0.502934	0.341417	0.645258
100	0.376383	0.192776	0.611627
1000	0.318182	0.133592	0.608297

**Table 2:** The case where  $r_k = 1/k$  for all  $k$ .

$n$	$1/\mu^*(n)$	$1/\mu_*(n)$
30	0.439971	0.625684
100	0.394713	0.604038
10,000	0.344817	0.583205

Calculus of the minimum eigenvalue:

$n$	$\lambda(n)$	Software
30	0.553865	<i>Mathematica</i>
100	0.518052	<i>Mathematica</i>
10,000	0.4564519	<i>Matlab</i>
100,000	0.4412748	<i>Matlab</i>

**Table 3:** The case where  $r_k = 2^{1-k}$  for all  $k$ .

$n$	$1/\mu^*(n)$	$1/\mu_*(n)$	Software
15	0.622407	0.729114	<i>Mathematica</i>
20	0.622396	0.728854	<i>Mathematica</i>
28	0.622396	0.728854	<i>Mathematica</i>
29	0.622396	0.728854	<i>Mathematica</i>

Calculus by Mathematica shows that

$$\begin{aligned} 1/\mu^*(n) &= 0.622396 \quad \text{for } n \geq 19 \\ 1/\mu_*(n) &= 0.728854 \quad \text{for } n \geq 19 \end{aligned}$$

Calculus of the minimum eigenvalue:

$n$	$\lambda(n)$	Software
5	0.708196	<i>Mathematica</i>
15	0.697629	<i>Mathematica</i>
17	0.697622	<i>Mathematica</i>
18	0.697625	<i>Mathematica</i> increases
20	0.69765	<i>Mathematica</i> increases
28	0.666465	<i>Mathematica</i> decreases
29	0	<i>Mathematica</i> absurd
29	0.697618	<i>Matlab</i>

Calculus by Mathematica shows that  $\lambda(n)$  becomes strange if  $n \geq 18$ .

Finally we change  $m(x)$  slightly and estimate  $\lambda_m(n)$  in this case.

**Table 4:** We choose  $m(x_k) = \frac{1}{R_k^2}$  and  $r_k = 2^{1-k}$ . Then we obtain:

$n$	$\lambda_m(n)$	$1/\mu^*(n)$	$1/\mu_*(n)$
30	0.0663717	0.0632777	0.116446
100	1.68955?	0.0196387	0.0383323

Calculus by Mathematica shows that  $\lambda_m(n)$  becomes strange if  $n \geq 51$ .

We remark that

$$\mu_*(n) = \tilde{G}m(x_1) = \sum_{k=1}^n \frac{1}{R_k^2} \rightarrow \infty$$

as  $n \rightarrow \infty$ , so that  $\lambda_m(n) \rightarrow 0$  as  $n \rightarrow \infty$ .

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