

# CPSC 490a Thesis Report

Xinwei (David) Yao  
Advisor: Daniel Spielman

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## Abstract

The thesis considers the hardness of the  $l_0$ -regularization problem of finding smooth extensions of functions on graphs, where one is given the function value at some initial vertices and an integer  $k$  and need to compute the function that minimizes the Laplacian quadratic form while coinciding with all but  $k$  of the initial values. The thesis focuses on its relation with the problem of finding the minimum bisection of a graph, which is known to be NP-hard, and gives a new reduction from MIN-BISECTION that yields further result on the relation between hardness of approximating the regularization problem and hardness of solving or approximating MIN-BISECTION for special graphs. Specifically,  $\frac{1}{n^4}$ -approximation of the regularization problem is NP-hard;  $\frac{1}{n^2}$ -approximation is at least as hard as MIN-BISECTION for bounded degree graphs; constant-factor approximation is at least as hard as constant-factor approximation of MIN-BISECTION for bounded degree graphs with large bisections.

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## 1 Introduction

Solving regressions on graphs is a very important problem for graph-based semi-supervised learning. In such a problem, one is given the value of a function on some of the vertices of the graph and needs to find an extension of the function to all vertices in a most continuous manner

respecting the edge weights. A common criteria for such continuity is to use the Laplacian quadratic form, as described below.

### 1.1 2-Laplacian minimization

Given a weighted undirected graph  $G = (V, E, l)$  and a function  $v_0 : T \rightarrow \mathbb{R}$  on  $T \subseteq V$ .  $|V| = n$  and  $|E| = m$ . Consider weights of the edges as a measure for the disparity between the two endpoints, so that less weight means greater similarity. Then the 2-Laplacian minimization problem asks for an extension  $v : V \rightarrow \mathbb{R}$  of the function  $v_0$  so that it solves the following:

$$\min_{\substack{v \in \mathbb{R}^n \\ v|_T = v_0}} v^\top L v,$$

where  $L$  is the Laplacian matrix of  $G$ .

Note that

$$v^\top L v = \sum_{(x,y) \in E} l(x,y)(v(x) - v(y))^2,$$

so one can think of the function  $v$  as assigning **voltages** to each vertex that minimize the overall **energy** on the graph.

The set  $T$  is also called the **boundary** of the problem since it gives the boundary condition, and the function values on vertices are also called **labels** in the language of Machine Learning.

### 1.2 Regularization

When solving the regression on data sets, one would also like to be able to perform regularization on the data to ignore noises in the initial labels  $v_0$  by removing the **outliers** from the boundary. The regularization that I consider is the  $l_0$  regularization of vertex values, which allows  $\|v|_T - v_0\|_0$ , the 0-norm in  $\mathbb{R}^{|T|}$ , to be at most  $k$ . In this way, the problem of  $l_0$ -regularization for  $l_2$  can be stated as finding the solution to the following:

$$\min_{\substack{v \in \mathbb{R}^n \\ \|v|_T - v_0\|_0 \leq k}} v^\top L v$$

Call this problem 2-REGULARIZATION.

### 1.3 Prior Work

[1] proves the following theorem by a Karp reduction from MIN-BISECTION.

**Theorem 1.1.** 2-REGULARIZATION is NP-Hard.

I will present a new polynomial reduction from MIN-BISECTION that implies not only Theorem 1.1 but also results on the approximability of the problem.

## 2 Reduction from Min-Bisection

Let  $G = (V, E)$  be any graph, and I will reduce MIN-BISECTION on  $G$  to 2-REGULARIZATION. Let  $n = |V|$  and  $m = |E|$ .

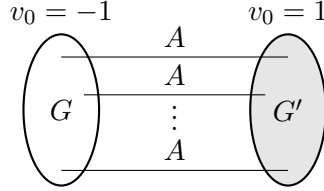


Figure 1: Reduction Gadget

## 2.1 Setup and notation

The graph  $H = (V_H, E_H, w)$  on which the regularization is performed will be two copies of  $G$  with a matching of the corresponding vertices in between.  $V_H = V \cup V'$ . The edges in the two copies have edge weight 1, and the  $n$  cross-edges have weight  $A$  for some large  $A$  to be determined later.

Call the two copies  $G$  and  $G'$ . I will consistently use lower-case letters to denote vertices in  $G$  and upper-case letters to denote sets of vertices in  $G$ , and use letters with the prime symbol to denote their corresponding copies in  $G'$ . Let  $\deg$  denote the degree of vertices in the original  $G$ .

Take any  $k \leq n$ , I will include all the vertices in the boundary. Let the initial label assignment  $v_0$  be  $-1$  on  $V$  and  $1$  on  $V'$ . Then regularization problem becomes changing  $k$  labels to achieve the lowest overall energy.

Define a function  $F : \mathcal{P}(V_H) \rightarrow \mathbb{R}$ . For any  $K \subseteq V_H$  be any set of vertices in  $H$ , let  $T = V_H \setminus K$  be its complement in  $H$ .  $F(K)$  is then defined to be the minimal energy achieved by relaxing the set  $K$  from the boundary:

$$F(K) = \min_{\substack{f \in \mathbb{R}^{2n} \\ v|_T = v_0|_T}} v^\top L_H v \quad (2.1)$$

where  $L_H$  is the Laplacian of the augmented graph  $H$ .

In this way, 2-REGULARIZATION becomes finding  $K^*$  and  $v^*$  so that  $|K^*| = k$  and

$$F(K^*) = \min_{\substack{K \subseteq V_H \\ |K|=k}} F(K) = F^*, \quad (2.2)$$

for some  $F^* \in \mathbb{R}$  that is the lowest possible energy. However,  $K^*$  may not be unique, so for my argument I will choose a  $K^*$  that is inside  $V$  as much as possible, that is:

$$|K^* \cap V| = \max_{\substack{|K|=k \\ F(K)=F^*}} |K \cap V| \quad (2.3)$$

$v^*$  is just the function  $v_0$  with change of labels in  $K^*$  that achieves  $F^*$ , that is:

$$\forall a \notin K^*, v^*(a) = v_0(a)$$

and

$$v^{*\top} L_H v^* = F(K^*) = F^*$$

The entire augmented graph  $H$  for regularization is shown in Figure 1. I will prove that when  $A$  is sufficiently large,  $K^*$  is a minimum-bisection of the original graph. The following fact I will use is from [2].

**Proposition 2.1.** *If  $v$  is the optimal function for relaxing set  $K$ , i.e. achieving the energy  $F(K)$ , then  $v$  is harmonic on  $H$  for the changed vertices. In other words,*

$$\forall a \notin K, v(a) = \frac{\sum_{(a,b) \in E_H} w_{a,b} v(b)}{\sum_{(a,b) \in E_H} w_{a,b}}$$

*In particular, the boundary prescribes the minimum and the maximum value of  $v$ :*

$$\forall a \in V_H, \min_{b \notin K} v_0(b) \leq v(a) \leq \max_{b \notin K} v_0(b)$$

## 2.2 Relaxation on one side only

In this subsection I will prove the following theorem.

**Theorem 2.2.** *For any Graph  $G = (V, E)$  and any integer  $0 \leq k \leq n$ ,  $K^* \subseteq V$  if  $A \geq 3d_{max}$  where  $d_{max}$  is the maximum degree of a node in  $G$ .*

I will first show the following Lemma.

**Lemma 2.3.** *For  $A \geq 3d_{max}$ , and some candidate  $K$ ,  $|K| = k \leq n$ , if  $\exists a \in K$  s.t.  $a' \in K$ , then  $\exists K'$  s.t.  $|K'| = |K|$ ,  $F(K') \leq F(K)$ , and  $|K' \cap V| > |K \cap V|$ .*

*Proof.* Let  $v$  be the function for  $K$  achieving  $F(K)$ . Since  $k \leq n$ , there are  $k - 2$  elements other than  $a$  and  $a'$  in  $K$ . But there are  $n - 1$  remaining pairs from the matching in  $H$ , then by Pigeonhole Principle we can find  $b \in V$  s.t.  $b, b' \notin K$ .

Let  $K' = K \setminus \{a'\} \cup \{b\}$ , and consider a test vector  $v'$  for  $K'$ , where

$$v'(x) = \begin{cases} 1, & x = a \\ v_0(a') = 1, & x = a' \\ 1, & x = b \\ v(x), & o.w. \end{cases}$$

By Equation 2.1, we have  $F(K') \leq v'^T L_H v'$ , so  $\Delta = F(K) - F(K') \geq v^T L_H v - v'^T L_H v'$ . Since  $v$  and  $v'$  are mostly the same, this difference only comes from edges that are attached to  $a$ ,  $a'$  and  $b$ .

The contribution from  $b$  is then

$$\begin{aligned} \Delta_b &= A(v(b') - v(b))^2 - 0 + \sum_{(b,c) \in E} (v(b) - v(c))^2 - (v'(b) - v'(c))^2 \\ &\geq 4A + \sum_{(b,c) \in E} -(v'(b) - v'(c))^2 \\ &= 4A + \sum_{(b,c) \in E, c \neq a} -(1 - v(c))^2, \text{ since } v'(a) = v'(b) = 1 \end{aligned}$$

By Proposition 2.1,  $-1 \leq v(c) \leq 1$ , so  $(1 - v(c)) \leq 2$ , and

$$\Delta_b \geq 4A - 4deg(b) \geq 4A - 4d_{max}$$

Similarly, the contribution from  $a$  and  $a'$  is

$$\begin{aligned}
\Delta_{aa'} &= A(v(a) - v(a'))^2 - 0 \\
&\quad + \sum_{(a,c) \in E} (v(a) - v(c))^2 - (v'(a) - v'(c))^2 + (v(a') - v(c'))^2 - (v'(a') - v'(c'))^2 \\
&\geq \sum_{(a,c) \in E} -(v'(a) - v'(c))^2 - (v'(a') - v'(c'))^2 \\
&\geq \sum_{(a,c) \in E, c \neq b} -(1 - v(c))^2 - (1 - v(c'))^2, \text{ since } v'(b) = v'(b') = v'(a) = v'(a') = 1 \\
&\geq -8\deg(a) \\
&\geq -8d_{max}
\end{aligned}$$

If  $A \geq 3d_{max}$ , combining the two terms we have  $\Delta \geq \Delta_b + \Delta_{aa'} \geq 4(A - 3d_{max}) \geq 0$ . Therefore  $F(K') \leq F(K)$  and by construction  $|K'| = k$  and  $|K' \cap V| > |K \cap V|$   $\square$

**Lemma 2.4.** *For some candidate  $K$ ,  $|K| = k \leq n$ . If  $\forall a \in K \cap V$ ,  $a' \notin K$ , then  $\exists K' \subseteq V$  s.t.  $|K'| = k$  and  $F(K') \leq F(K)$ .*

*Proof.* Suppose  $K \not\subseteq V$ , since otherwise we can set  $K' = K$ . Then  $K$  is partitioned into  $S$  and  $T'$  so that  $S \cup T' = K$ ,  $S \subseteq V$  and  $T' \subseteq V'$ . We also have that  $S \cap T = \emptyset$  where  $T$  is the copy of  $T'$  in  $G$ . Let  $v$  be the function for  $K$  that achieves  $F(K)$ .

Let  $K' = K \setminus T' \cup T \subseteq V$ , and consider a test vector  $v'$  for  $K'$ , where

$$v'(x) = \begin{cases} v_0(x) = 1, & x \in V' \\ v(x), & x \in S \\ -v(x'), & x \in T \\ v_0(x) = -1, & o.w. \end{cases}$$

By Equation 2.1, we then have  $F(K') \leq v'^\top L_H v'$ , so  $\Delta = F(K) - F(K') \geq v^\top L_H v - v'^\top L_H v'$ .

Notice that contributions of each cross edge stays the same, as  $(-1 - v(x'))^2 = (1 - v'(x))^2$  whenever  $x \in T$ . Contributions of edges within  $T$  for  $v'$  is also the same as contributions of edges within  $T'$  for  $v$ . Furthermore, contributions of edges from  $T$  to  $V \setminus K'$  for  $v'$  is the same as contributions of edges from  $T'$  to  $V' \setminus (S' \cup T')$ . Therefore the difference only comes from edges from  $T$  to  $S$  and from  $T'$  to  $S'$ .

$$\begin{aligned}
\Delta &\geq \sum_{\substack{(a,b) \in E \\ a \in S, b \in T}} (v(a') - v(b'))^2 + (v(a) - v(b))^2 - (v'(a) - v'(b))^2 - (v'(a') - v'(b'))^2 \\
&= \sum (1 - v(b'))^2 + (v(a) + 1)^2 - (v(a) + v(b'))^2 \\
&= \sum 2 + 2v(a) - 2v(b') - 2v(a)v(b') \\
&= \sum 2(1 + v(a))(1 - v(b'))
\end{aligned}$$

By Proposition 2.1, each term in the sum is non-negative, so  $\Delta \geq 0$ . Therefore  $\Delta \geq 0$ ,  $F(K') \leq F(K)$  and  $|K'| = k$  and  $K' \subseteq V$ .  $\square$

*Remark 2.5.* When  $K \subseteq V'$ , the  $K'$  and  $v'$  constructed in the proof is the symmetric reflection of  $v$  and  $K$ , and  $F(K) = F(K')$  since  $S = S' = \emptyset$ .

Now I can prove Theorem 2.2 from Lemma 2.3 and Lemma 2.4.

*Proof of Theorem 2.2.* By Lemma 2.3 and Equation 2.3, we conclude that  $\forall a \in K^* \cap V$ ,  $a' \notin K^*$ . Then we can apply Lemma 2.4 and conclude that  $K^* \subseteq V$ .  $\square$

### 2.3 Bounding the lowest energy value

**Theorem 2.6.** *For any graph  $G$  and any integer  $0 \leq k \leq n$ ,  $K \subseteq V$ . Let  $\delta = |\partial K|$  be the number of edges cut in  $G$ . Let  $T = V \setminus K$  be the complement of  $K$  in  $G$ . Then  $F(K)$  is between  $4A(n - k) + 4\delta - \frac{4d_{max}}{A+d_{max}}\delta$  and  $4A(n - k) + 4\delta$ .*

*Proof.* The upper bound is given by a test vector  $v'$  for  $K$  s.t.  $v'(K) = 1$ .

$$F(K) \leq v'^T L_H v' = 4A * |T| + 4\delta = 4A(n - k) + 4\delta$$

For any  $S \subseteq V$ ,  $U \subseteq V$ , write  $E(S, U)$  to denote the edges across the two sets, that is  $E(S, U) = \{(a, b) \in E : a \in S, b \in U\}$ .

For  $a \in K$ , let  $E(a, T) = \{(a, b) \in E : b \in T\}$  be the set of edges going from  $a$  to outside of  $K$ . Then we have

$$\delta = \sum_{a \in K} |E(a, T)| = |E(K, T)| \quad (2.4)$$

To get a lower bound for  $F(K)$ , we split it into contributions from three sets of edges: edges within  $K$ , edges from  $K$  to  $T$ , and cross edges between  $G$  and  $G'$ . Let  $v$  be the function for  $K$  that achieves  $F(K)$ .

$$\begin{aligned} v^T L_H v &= \sum_{(a,b) \in E(K,K)} (v(a) - v(b))^2 \\ &+ \sum_{(a,b) \in E(K,T)} (v(a) + 1)^2 + \sum_{a \in K} A(v(a) - 1)^2 \\ &+ 4A * |T| \\ &\geq 4A(n - k) + \sum_{a \in K} |E(a, T)| (v(a) + 1)^2 + A(v(a) - 1)^2 \\ &= 4A(n - k) + \sum_{a \in K} (|E(a, T)| + A)(v(a))^2 + \frac{2(|E(a, T)| - A)}{|E(a, T)| + A} v(a) + 1 \\ &\geq 4A(n - k) + \sum_{a \in K} (|E(a, T)| + A) \left(1 - \left(\frac{|E(a, T)| - A}{|E(a, T)| + A}\right)^2\right) \\ &= 4A(n - k) + \sum_{a \in K} \frac{4|E(a, T)|A}{|E(a, T)| + A} \\ &\geq 4A(n - k) + \sum_{a \in K} \frac{4A|E(a, T)|}{A + d_{max}} \\ &= 4A(n - k) + \frac{4A}{A + d_{max}} \sum_{a \in K} |E(a, T)| \\ &= 4A(n - k) + \left(4 - \frac{4d_{max}}{A + d_{max}}\right) \delta, \text{ by Equation 2.4} \end{aligned}$$

Therefore  $F(K)$  is between  $4A(n - k) + 4\delta - \frac{4d_{max}}{A+d_{max}}\delta$  and  $4A(n - k) + 4\delta$ .  $\square$

## 2.4 Min-Bisection

The relation with MIN-BISECTION can be shown by applying Theorem 2.2 and Theorem 2.6.

Let  $k = \frac{n}{2}$ . Since  $d_{max} \leq n - 1$  and  $\delta \leq (n - k)k = \frac{n^2}{4}$ , choose  $A = n^3 > 3d_{max}$ , then by Theorem 2.2, we have  $K^* \subseteq V$ . Furthermore,  $\frac{d_{max}\delta}{A+d_{max}} < \frac{1}{4}$ , so by Theorem 2.6,

$$F^* \in (2n^4 + 4\delta - 1, 2n^4 + 4\delta]$$

which strictly decreases with  $\delta$ , so  $K^*$  has to be a minimal bisection.

In summary, the reduction is as follows. Given any graph  $G = (V, E)$ , construct 2-REGULARIZATION problem on  $H = (V_H, E_H, w)$  takes polynomial time. Solving 2-REGULARIZATION for  $k = \frac{n}{2}$  gives the set  $K^*$ , which is a min-bisection of  $G$ .

## 3 Hardness Implications

### 3.1 Hardness of exact result

The polynomial reduction described in Section 2.4 implies Theorem 1.1.

### 3.2 Hardness of $\frac{1}{n^4}$ -approximation

**Theorem 3.1.**  $\frac{1}{(\frac{1}{2}+\epsilon)n^4}$ -approximation of 2-REGULARIZATION is NP-hard

*Proof.* Suppose there is a  $\frac{1}{(\frac{1}{2}+\epsilon)n^4}$  approximation, set  $A = \frac{n^3}{2}$ . Let  $K$  be the min-bisection and  $\delta = |\partial K|$ .

By Lemma 2.3 and Lemma 2.4, we can transform the output set of the approximation algorithm into a set  $K' \subseteq V$  which is at least as good an approximation, i.e.  $F(K') \leq (1 + \frac{1}{(\frac{1}{2}+\epsilon)n^4})F^*$ . Let  $\delta' = |\partial K'|$ . By Theorem 2.6, we have

$$n^4 + 4\delta' - \frac{4d_{max}}{n^3/2 + d_{max}}\delta' \leq (1 + \frac{1}{(\frac{1}{2} + \epsilon)n^4})(n^4 + 4\delta)$$

which simplifies to

$$\begin{aligned} \delta' - \delta &\leq \frac{d_{max}\delta'}{n^3/2 + d_{max}} + \frac{1}{2 + 4\epsilon} + \frac{\delta}{(\frac{1}{2} + \epsilon)n^4} \\ &< \frac{(n - 1)n^2/4}{n^3/2} + \frac{1}{2 + \epsilon} + \frac{n^2/4}{2n^4} \\ &< \frac{1}{2} + \frac{1}{2 + \epsilon} + \frac{1}{2n^2} \\ &< 1, \text{ when } n \text{ is sufficiently large} \end{aligned}$$

So for sufficiently large  $n$ ,  $\delta' = \delta$  and  $K'$  is the min-bisection of  $G$ . Since solving MIN-BISECTION is NP-hard,  $\frac{1}{(\frac{1}{2}+\epsilon)n^4}$ -approximation of 2-REGULARIZATION is NP-hard.  $\square$

### 3.3 Hardness of $\frac{1}{n^2}$ -approximation

Consider MIN-BISECTION on graphs with constant bounded degree  $d$ , and call this problem MIN-BISECTION $_d$ .

**Theorem 3.2.** *There is a polynomial reduction from MIN-BISECTION $_d$  to a  $\frac{1}{(d^2+\epsilon)n^2}$ -approximation of 2-REGULARIZATION*

*Proof.* Set  $A = d^2n$ . Since  $G$  has bounded degree  $d$ , with the same set up as in Theorem 3.1, we have  $\delta \leq \delta' \leq \frac{dn}{2}$ . Using Theorem 2.6, we get

$$\begin{aligned} \delta' - \delta &\leq \frac{d\delta'}{d^2n + d} + \frac{d^2n}{2(d^2 + \epsilon)n} + \frac{\delta}{(d^2 + \epsilon)n^2} \\ &< \frac{1}{2} + \frac{1}{2 + 2d^{-2}\epsilon} + \frac{1}{2dn} \\ &< 1, \text{ when } n \text{ is sufficiently large} \end{aligned}$$

Therefore for sufficiently large  $n$ ,  $\delta' = \delta$  and  $K'$  is the min-bisection of  $G$ .  $\square$

*Remark 3.3.* This implies that getting  $\frac{1}{(d^2+\epsilon)n^2}$ -approximation of 2-REGULARIZATION is at least as hard as solving MIN-BISECTION on graphs with bounded degree  $d$ .

### 3.4 Hardness of constant-factor approximation

Consider the problem MIN-BISECTION $_{d,\beta}$ . It is MIN-BISECTION $_d$  on graphs with not only constant bounded degree  $d$ , but also that any half of the vertices will have good expansion. More precisely, they are graphs  $G$  with  $\delta(G) \geq \beta nd$ , where  $\delta(G)$  is the number of edges cut in the minimum bisection.

**Theorem 3.4.** *There is a polynomial reduction from  $\epsilon$ -approximation of MIN-BISECTION $_{d,\beta}$  to a  $\frac{\beta\epsilon^2}{\alpha}$ -approximation of 2-REGULARIZATION, where  $\alpha = 2 + \epsilon + \frac{3}{2}\epsilon^2$ .*

*Proof.* Set  $A = \frac{2d}{\epsilon}$ . Use the same set up as in Theorem 3.1. We have  $\beta dn \leq \delta \leq \delta' \leq \frac{dn}{2}$ . Using Theorem 2.6, we get



$$\begin{aligned}
2nA + 4\delta' - \frac{4d}{A+d}\delta' &\leq \left(1 + \frac{\beta\epsilon^2}{\alpha}\right)(2nA + 4\delta) \\
\frac{4A}{A+d}\delta' &\leq 4\delta + \frac{\beta\epsilon^2}{\alpha}(2nA + 4\delta) \\
\frac{A}{A+d}\delta' &\leq \left(1 + \frac{\beta\epsilon^2}{\alpha}\right)\delta + \frac{\beta\epsilon nd}{\alpha} \\
\frac{A}{A+d} \frac{\delta'}{\delta} &\leq \left(1 + \frac{\beta\epsilon^2}{\alpha}\right) + \frac{\beta\epsilon nd}{\alpha\beta nd} \\
\frac{\delta'}{\delta} &\leq \left(1 + \frac{\beta\epsilon^2}{\alpha} + \frac{\epsilon}{\alpha}\right)\left(1 + \frac{d}{A}\right) \\
&\leq \left(1 + \frac{\beta\epsilon^2}{\alpha} + \frac{\epsilon}{\alpha}\right)\left(1 + \frac{\epsilon}{2}\right) \\
&\leq 1 + \frac{\epsilon}{2} + \frac{\beta\epsilon^2 + \beta\epsilon^3/2 + \epsilon + \epsilon^2/2}{\alpha} \\
&\leq 1 + \frac{\epsilon}{2} + \frac{\epsilon}{2} \frac{2\beta\epsilon + \beta\epsilon^2 + 2 + \epsilon^2}{\alpha} \\
&\leq 1 + \epsilon, \text{ since } \beta \leq \frac{1}{2}
\end{aligned}$$

Therefore  $\delta'$  is a  $\epsilon$ -approximation of  $\delta$ , the min-bisection of  $G$ . □

*Remark 3.5.* This implies that getting  $\frac{2\beta\epsilon^2}{9}$ -approximation of 2-REGULARIZATION is at least as hard as getting  $\epsilon$ -approximation for  $\text{MIN-BISECTION}_{d,\beta}$  for  $\epsilon \leq 1$

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