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# The Convexity Estimates for the Solutions of Two Elliptic Equations 

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#### Abstract

In this paper, for the solutions of two elliptic equations we find the auxiliary curvature functions which attain respective minimum on the boundary. These results are the generalization of the classical ones in Makar-Limanov [17] for the torsion equation and Acker et al. [1] for the first eigenfunction of the Laplacian in convex domains of dimension 2. Then we get the new proof of the specific convexity of the solutions of the above two elliptic equations. As a consequence, for the elliptic equation $v \Delta v=-\left(1+|\nabla v|^{2}\right)$ in a smooth, bounded and strictly convex domain $\Omega$ in $\mathbb{R}^{n}$ with homogeneous Dirichlet boundary value condition, we also get a sharply lower bound estimate of the Gaussian curvature for the solution surface by the curvature of the boundary of the domain.


Keywords Convex domain; Convexity estimates; First eigenvalue problem; Torsion problem.

Mathematics Subject Classification 35B30; 35J25; 35E10.

## 1. Introduction and Results

Let $\Omega$ be a bounded convex domain in $\mathbb{R}^{n}$. In this paper, we will consider the convexity estimates for the homogeneous Dirichlet problems of two elliptic partial differential equations. The first one is the torsion problem:

$$
\begin{cases}\Delta u=-2 & \text { in } \Omega,  \tag{1.1}\\ u=0 & \text { on } \partial \Omega .\end{cases}
$$

And the second one is the first eigenvalue problem of the Laplace operator:

$$
\begin{cases}\Delta u+\lambda u=0 & \text { in } \Omega,  \tag{1.2}\\ u=0 & \text { on } \partial \Omega .\end{cases}
$$

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In 1971, Makar-Limanov [17] considered the boundary value problem (1.1) in a bounded plane convex domain $\Omega$. He introduced the function

$$
P_{1}=2 u \operatorname{det} D^{2} u+2 u_{1} u_{2} u_{12}-u_{11} u_{2}^{2}-u_{22} u_{1}^{2}
$$

and proved that $P_{1}$ is a superharmonic function. Then he could obtain that $u^{\frac{1}{2}}$ is strictly concave.

In 1976, Brascamp and Lieb [4] established the log-concavity of the first eigenfunction of the Laplace operator for the problem (1.2) in convex domains. For the case of dimension two, Acker et al. [1] utilized the idea of Makar-Limanov [17] to find the following function

$$
P_{2}=\frac{1}{u}\left[u \operatorname{det} D^{2} u+2 u_{1} u_{2} u_{12}-u_{11} u_{2}^{2}-u_{22} u_{1}^{2}\right],
$$

which satisfies the following elliptic differential equality:

$$
\Delta P_{2}=0, \quad \bmod \left(\nabla P_{2}\right) \text { in } \Omega \backslash\{x \in \Omega \mid \Theta(x)=0\},
$$

where we have suppressed the terms containing the gradient of $P_{2}$ with locally bounded coefficients, and $\Theta(x)=4 v_{12}^{2}+\left(v_{11}-v_{22}\right)^{2}$ for $v=-\log u$. Then they obtained a new proof for the Brascamp-Lieb's result in two dimensional case.

Now we state a brief history on the convexity of elliptic PDE. In 1983, Korevaar [11] introduced a very useful technique to study the convexity of the solution for a class of elliptic equations. The new proofs of the log-concavity of the first eigenfunction of convex domains were given respectively by Korevaar [12] and Caffarelli and Spruck [6]. In different extent, Kawohl [9] and Kennington [10] improved Korevaar's methods, which enabled them to give a higher dimensional generalization of the result of Makar-Limanov [17]. In particular, Kennington pointed out that the concavity number $\frac{1}{2}$ of $u$ is sharp in the problem (1.1) in higher dimensional case. Singer et al. [18] and Caffarelli and Friedman [5] introduced a new deformation technique to deal with the convexity, and Caffarelli and Friedman [5] established the strict convexity of the solution for some equations in 2dimensional convex domains. Korevaar and Lewis [13] generalized the deformation method to higher dimensions, and obtained the strict concavity of $u^{\frac{1}{2}}$ in (1.1) in higher dimensional case. A survey of this subject is given by Kawohl [8] and Guan and Ma [7].

In this paper, we generalize the technique of Makar-Limanov [17] and Acker et al. [1] to the higher dimensional case. We also find new corresponding auxiliary functions, and modulo the gradient terms we prove that they are superharmonic under the strict convexity assumption of the solutions. So from the minimum principle we get the convexity estimates for the solutions of (1.1) and (1.2) via boundary data. Combining the deformation methods we can give the new proof of the specific convexity of the solutions of the above two elliptic equations, and we obtain the Gaussian curvature estimate for the graph of $v=-\sqrt{u}$ in the problem (1.1) using the curvature of the boundary of domain.

In order to state our results, we need the standard curvature formula of the level sets of a function (see Trudinger [20]). Since the level sets of the solutions in the
problems (1.1) and (1.2) are convex with respect to the normal direction $\nabla u$, we get

$$
\begin{equation*}
K(x)=(-1)^{n-1} \sum_{i, j=1}^{n} \frac{\partial \operatorname{det} D^{2} u}{\partial u_{i j}} u_{i} u_{j}|\nabla u|^{-(n+1)}, \tag{1.3}
\end{equation*}
$$

where $K(x)$ is the Gaussian curvature of the level sets of the solution $u$ at $x \in \Omega$.
Now we state our main results. (A function $v$ is called strictly convex here if $D^{2} v>0$. Similarly a domain $\Omega$ is called strictly convex if at every point of $\partial \Omega$ all principal curvatures are strictly positive.)

Theorem 1.1. Let $\Omega$ be a smooth, bounded convex domain in $\mathbb{R}^{n}, n \geq 2$, and $u$ the solution for the problem (1.1). If $v=-\sqrt{u}$ is a strictly convex function, then the function

$$
\psi_{1}=(-v)^{n+2} \operatorname{det} D^{2} v=(-2)^{-n} u \operatorname{det} D^{2} u+(-2)^{-n-1} \sum_{i, j=1}^{n} \frac{\partial \operatorname{det} D^{2} u}{\partial u_{i j}} u_{i} u_{j}
$$

satisfies the following elliptic differential inequality:

$$
\begin{equation*}
\Delta \psi_{1} \leq 0 \bmod \left(\nabla \psi_{1}\right) \text { in } \Omega, \tag{1.4}
\end{equation*}
$$

where we have suppressed the terms containing the gradient of $\psi_{1}$ with locally bounded coefficients. Moreover, the function $\psi_{1}$ attains its minimum on the boundary. So from (1.4), we have the following estimate for the solution of the problem (1.1)

$$
\begin{equation*}
\psi_{1}=(-v)^{n+2} \operatorname{det} D^{2} v \geq 2^{-(n+1)} \min _{\partial \Omega} K \min _{\partial \Omega}|\nabla u|^{n+1} \tag{1.5}
\end{equation*}
$$

where $K$ is the Gaussian curvature of $\partial \Omega$. Finally, the function $\psi_{1}$ attains its minimum in $\Omega$ if and only if $\Omega$ is an ellipsoid (ellipse).

For the eigenvalue problem (1.2), we give a similar result as Theorem 1.1, that is
Theorem 1.2. Let $\Omega$ be a smooth, bounded convex domain in $\mathbb{R}^{n}, n \geq 2$, and $u>0$ the first eigenfunction for the eigenvalue problem (1.2). If $v=-\log u$ is a strictly convex function, then the function

$$
\psi_{2}=e^{-(n+1) v} \operatorname{det} D^{2} v=(-1)^{n} u \operatorname{det} D^{2} u+(-1)^{n-1} \sum_{i, j=1}^{n} \frac{\partial \operatorname{det} D^{2} u}{\partial u_{i j}} u_{i} u_{j}
$$

satisfies the following elliptic differential inequality:

$$
\begin{equation*}
\Delta \psi_{2} \leq 0 \bmod \left(\nabla \psi_{2}\right) \text { in } \Omega, \tag{1.6}
\end{equation*}
$$

where we have suppressed the terms containing the gradient of $\psi_{2}$ with locally bounded coefficients. Moreover, the function $\psi_{2}$ attains its minimum on the boundary. So from (1.6), we have the following estimate for the solution of the problem (1.2)

$$
\begin{equation*}
\psi_{2}=e^{-(n+1) v} \operatorname{det} D^{2} v \geq \min _{\partial \Omega} K \min _{\partial \Omega}|\nabla u|^{n+1}, \tag{1.7}
\end{equation*}
$$

where $K$ is the Gaussian curvature of $\partial \Omega$.

From the results in Theorem 1.1, we further give the Gaussian curvature estimate for the graph of $v=-\sqrt{u}$ in the problem (1.1) via the boundary geometry of the domain.

Corollary 1.3. Let $\Omega$ be a smooth, bounded and strictly convex domain in $\mathbb{R}^{n}, \kappa_{\text {min }}$, $\kappa_{\text {max }}$ and $K_{\text {min }}$ the minimal principal curvature, maximal principal curvature and the minimal Gaussian curvature of the boundary $\partial \Omega$ respectively. If $u$ is the solution for (1.1) and $v=-\sqrt{u}$, then Gaussian curvature $K_{G}$ for the graph of $v$ satisfies the following sharp estimate

$$
\begin{equation*}
K_{G}=\frac{\operatorname{det} D^{2} v}{\left(1+|\nabla v|^{2}\right)^{\frac{n+2}{2}}} \geq \frac{K_{\min } \kappa_{\min }^{n+2}}{n^{\frac{n}{2}} \kappa_{\max }^{n+1}} \text { in } \Omega \tag{1.8}
\end{equation*}
$$

When we take $\Omega$ being the unit ball $B_{1}(0)$ of dimension $n$ centering at 0 , the equality in (1.8) holds at the origin 0 .

For the Gauss curvature estimates of the solution surface to elliptic partial differential equations, Corollary 1.3 is few example which generalize the two dimensional result by Ma [14] for the problem (1.1).

In a different situation, Andrews and Clutterbuck [2] obtained another type convexity estimates for the eigenvalue problem (1.2) and found a beautiful application on the proof of the fundamental gap conjecture.

This paper is organized as follows. In Section 2, we prove Theorem 1.1 through establishing a differential inequality for the given function. Then, we give some estimates on the solution in (1.1) and its gradient on the boundary. Combining these estimates with Theorem 1.1, we obtain the proof of Corollary 1.3. In Section 3, using the same process as the proof of Theorem 1.1, we prove Theorem 1.2.

## 2. Convexity Estimates for the Torsion Problem

In this section, through establishing a differential inequality for the given function and applying minimum principle, we first prove Theorem 1.1. In order to prove Corollary 1.3, we give two lemmas which involve the estimates of the solution $u$ and the modulus of its gradient $|\nabla u|$ on the boundary $\partial \Omega$. Then Corollary 1.3 is proved by applying Theorem 1.1.

Now, we begin to prove Theorem 1.1
Proof. Let $u$ be the solution for the problem (1.1) and $v=-\sqrt{u}$. Then $v$ is strictly convex from our assumption and satisfies the following problem

$$
\begin{cases}\Delta v=-\frac{1+|\nabla v|^{2}}{v} & \text { in } \Omega,  \tag{2.1}\\ v=0 & \text { on } \partial \Omega\end{cases}
$$

For

$$
\psi_{1}=(-v)^{n+2} \operatorname{det} D^{2} v=(-2)^{-n} u \operatorname{det} D^{2} u+(-2)^{-n-1} \sum_{i, j=1}^{n} \frac{\partial \operatorname{det} D^{2} u}{\partial u_{i j}} u_{i} u_{j}
$$

we shall show that

$$
\varphi=\log \psi_{1}=(n+2) \log (-v)+\log \operatorname{det} D^{2} v
$$

satisfies the following elliptic differential inequality:

$$
\begin{equation*}
\Delta \varphi \leq 0 \bmod (\nabla \varphi) \text { in } \Omega, \tag{2.2}
\end{equation*}
$$

which implies the inequality (1.4). Moreover, we obtain that the function $\psi_{1}$ attains its minimum on the boundary by the standard minimum principle. Therefore, from (1.3), we have

$$
\begin{aligned}
\psi_{1} & =(-v)^{n+2} \operatorname{det} D^{2} v \geq \min _{\partial \Omega} \psi_{1} \\
& =2^{-(n+1)} \min _{\partial \Omega}\left\{K|\nabla u|^{n+1}\right\} \\
& \geq 2^{-(n+1)} \min _{\partial \Omega} K \min _{\partial \Omega}|\nabla u|^{n+1},
\end{aligned}
$$

which is the estimate (1.5).
In order to prove (2.2) at an arbitrary point $x_{o}$, we can choose the coordinates at $x_{o}$ such that the matrix $\left(v_{i j}\left(x_{o}\right)\right)(1 \leq i, j \leq n)$ is diagonal. If we can establish (2.2) at $x_{o}$ under the above coordinates assumption, then go back to the original coordinates we find that (2.2) remain valid with new locally bounded coefficients on $\nabla \varphi$ in (2.2), depending smoothly on the independent variables. Thus it remains to establish (2.2) under the above assumption that the matrix $\left(v_{i j}\left(x_{o}\right)\right)(1 \leq i, j \leq n)$ is diagonal. From now on, all the calculation will be done at the fixed point $x_{o}$.

Because $v$ is strictly convex, the Hessian matrix $\left(v_{i j}\right)$ is positive definite. Let $\lambda_{i}=v_{i i}, \lambda=\left(\lambda_{1}, \lambda_{2}, \ldots, \lambda_{n}\right)$ and $\sigma_{1}(\lambda \mid i)=\sum_{k=1, k \neq i}^{n} \lambda_{k}, i=1,2, \ldots, n$. Let $\left(v^{i j}\right)$ be the inverse matrix of $\left(v_{i j}\right)$. Taking the first derivative of $\varphi$, we have

$$
\begin{equation*}
\varphi_{i}=\frac{(n+2) v_{i}}{v}+\sum_{k, l=1}^{n} v^{k l} v_{k l i} \tag{2.3}
\end{equation*}
$$

and it follows that

$$
\begin{equation*}
\sum_{k=1}^{n} v^{k k} v_{k k i}=-\frac{(n+2) v_{i}}{v}+\varphi_{i} . \tag{2.4}
\end{equation*}
$$

Taking the second derivative of $\varphi$, we get

$$
\begin{equation*}
\varphi_{i i}=\frac{(n+2)\left(v v_{i i}-v_{i}^{2}\right)}{v^{2}}+\sum_{k, l=1}^{n} v^{k l} v_{k l i i}-\sum_{k, l, p, q=1}^{n} v^{k q} v^{p l} v_{k l i} v_{p q i} . \tag{2.5}
\end{equation*}
$$

Then

$$
\begin{equation*}
\Delta \varphi=\frac{(n+2)\left(v \Delta v-|\nabla v|^{2}\right)}{v^{2}}+\sum_{k=1}^{n} v^{k k} \Delta\left(v_{k k}\right)-\sum_{k, l, i=1}^{n} v^{k k} v^{l l} v_{k l i}^{2} \triangleq A+B+C, \tag{2.6}
\end{equation*}
$$

where

$$
\begin{aligned}
A & =\frac{(n+2)\left(v \Delta v-|\nabla v|^{2}\right)}{v^{2}} \\
B & =\sum_{k=1}^{n} v^{k k} \Delta\left(v_{k k}\right) \\
C & =-\sum_{k, l, i=1}^{n} v^{k k} v^{l l} v_{k l i}^{2}
\end{aligned}
$$

We will deal with the three terms above respectively. For the term $A$, we have

$$
\begin{equation*}
A=\frac{(n+2) \Delta v}{v}-\frac{(n+2)|\nabla v|^{2}}{v^{2}} . \tag{2.7}
\end{equation*}
$$

For the term $B$, taking derivative of the equation in (2.1), it follows that

$$
\begin{aligned}
(\Delta v)_{k}= & \frac{\left(1+|\nabla v|^{2}\right) v_{k}}{v^{2}}-2 \frac{\sum_{i=1}^{n} v_{i} v_{i k}}{v}, \\
(\Delta v)_{k k}= & -2 \frac{\left(1+|\nabla v|^{2}\right) v_{k}^{2}}{v^{3}}+\frac{\left(1+|\nabla v|^{2}\right) v_{k k}+2 \sum_{i=1}^{n} v_{i} v_{i k} v_{k}}{v^{2}} \\
& +2 \frac{\sum_{i=1}^{n} v_{i} v_{i k} v_{k}}{v^{2}}-2 \frac{\sum_{i=1}^{n}\left(v_{i k}^{2}+v_{i} v_{i k k}\right)}{v} \\
= & -2 \frac{\left(1+|\nabla v|^{2}\right) v_{k}^{2}}{v^{3}}+\frac{\left(1+|\nabla v|^{2}\right) v_{k k}+4 v_{k}^{2} v_{k k}}{v^{2}}-2 \frac{v_{k k}^{2}+\sum_{i=1}^{n}\left(v_{i} v_{i k k}\right)}{v}
\end{aligned}
$$

Thus,

$$
\begin{align*}
B & =\sum_{k}^{n} v^{k k} \Delta\left(v_{k k}\right)=\sum_{k}^{n} v^{k k}(\Delta v)_{k k} \\
& =-2 \sum_{k=1}^{n} \frac{\left(1+|\nabla v|^{2}\right) v_{k}^{2}}{\lambda_{k} v^{3}}+\frac{n+(n+4)|\nabla v|^{2}}{v^{2}}-2 \frac{\Delta v+\sum_{i=1}^{n} v_{i}\left(-\frac{(n+2) v_{i}}{v}+\varphi_{i}\right)}{v} \\
& =2 \frac{\Delta v}{v^{2}} \sum_{k=1}^{n} \frac{v_{k}^{2}}{\lambda_{k}}-\frac{(n+2) \Delta v}{v}+\frac{(2 n+8)|\nabla v|^{2}}{v^{2}}-2 \sum_{i=1}^{n} \frac{v_{i} \varphi_{i}}{v}, \tag{2.8}
\end{align*}
$$

where we have used (2.4) and (2.1) in the last two equalities. Therefore, combining (2.7) with (2.8), we yield that

$$
\begin{align*}
A+B & =\frac{(n+6)|\nabla v|^{2}}{v^{2}}+2 \frac{\Delta v}{v^{2}} \sum_{k=1}^{n} \frac{v_{k}^{2}}{\lambda_{k}}-2 \sum_{i=1}^{n} \frac{v_{i} \varphi_{i}}{v} \\
& =\sum_{k=1}^{n}\left[(n+8)+2 \frac{\sigma_{1}(\lambda \mid k)}{\lambda_{k}}\right] \frac{v_{k}^{2}}{v^{2}}-2 \sum_{i=1}^{n} \frac{v_{i} \varphi_{i}}{v} \\
& =\sum_{k=1}^{n}\left[(n+8)+2 \frac{\sigma_{1}(\lambda \mid k)}{\lambda_{k}}\right] \frac{v_{k}^{2}}{v^{2}} \bmod (\nabla \varphi) . \tag{2.9}
\end{align*}
$$

Next, we treat the term $C$ and have

$$
\begin{aligned}
-C & =\sum_{k, l, i=1}^{n} \frac{v_{k l i}^{2}}{\lambda_{k} \lambda_{l}} \\
& =\left[\sum_{k=l=i=1}^{n}+\sum_{\substack{k, l, i=1 \\
k=l, k \neq i}}^{n}+\sum_{\substack{k, l, i=1 \\
k=i, k \neq l}}^{n}+\sum_{\substack{k, l, i=1 \\
l=i, k \neq l}}^{n}+\sum_{\substack{k, l, i=1 \\
k \neq l, l \neq i, k \neq i}}^{n}\right] \frac{v_{k l i}^{2}}{\lambda_{k} \lambda_{l}} \\
& =\sum_{i=1}^{n} \frac{v_{i i i}^{2}}{\lambda_{i}^{2}}+\sum_{\substack{k, l=1 \\
k \neq l}}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{l}}\right) v_{k k l}^{2}+\sum_{\substack{k, l, i=1 \\
k \neq l, l \neq i, k \neq i}}^{n} \frac{v_{k l i}^{2}}{\lambda_{k} \lambda_{l}} \\
& \geq \sum_{i=1}^{n} \frac{v_{i i l}^{2}}{\lambda_{i}^{2}}+\sum_{\substack{k_{l} l=1 \\
k \neq l}}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{l}}\right) v_{k k l}^{2} \\
& =\sum_{l=1}^{n}\left(\frac{v_{l l l}^{2}}{\lambda_{l}^{2}}+\sum_{k=1, k \neq l}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{l}}\right) v_{k k l}^{2}\right) \\
& \triangleq \sum_{l=1}^{n} C_{l},
\end{aligned}
$$

where

$$
C_{l}=\frac{v_{l l}^{2}}{\lambda_{l}^{2}}+\sum_{k=1, k \neq l}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{l}}\right) v_{k k l}^{2}, \quad l=1, \ldots, n .
$$

We claim that

$$
\begin{equation*}
C_{l} \geq\left[(n+8)+2 \frac{\sigma_{1}(\lambda \mid l)}{\lambda_{l}}\right] \frac{v_{l}^{2}}{v^{2}} \bmod (\nabla \varphi), \quad l=1,2, \ldots, n \tag{2.10}
\end{equation*}
$$

From this claim, we have

$$
\begin{equation*}
C \leq-\sum_{l=1}^{n} C_{l} \leq-\sum_{l=1}^{n}\left[(n+8)+2 \frac{\sigma_{1}(\lambda \mid l)}{\lambda_{l}}\right] \frac{v_{l}^{2}}{v^{2}} \bmod (\nabla \varphi) . \tag{2.11}
\end{equation*}
$$

Consequently, (2.6), (2.9) and (2.11) yields

$$
\Delta \varphi \leq 0 \quad \bmod (\nabla \varphi),
$$

which is exactly (2.2).
Now, we prove the claim for $l=1$ and others are the same completely. Taking derivative of the equation in (2.1) with respect to $x_{1}$, we get the equality

$$
\begin{aligned}
(\Delta v)_{1} & =\frac{\left(1+|\nabla v|^{2}\right) v_{1}}{v^{2}}-2 \frac{\sum_{i=1}^{n} v_{i} v_{i 1}}{v} \\
& =-\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right) \frac{v_{1}}{v},
\end{aligned}
$$

that is

$$
\begin{equation*}
v_{111}=-\sum_{k=2}^{n} v_{k k 1}-\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right) \frac{v_{1}}{v} . \tag{2.12}
\end{equation*}
$$

From (2.4), we have

$$
\begin{equation*}
\sum_{k=1}^{n} \frac{v_{k k 1}}{\lambda_{k}}=-\frac{(n+2) v_{1}}{v}+\varphi_{1} \tag{2.13}
\end{equation*}
$$

That (2.13) subtracts (2.12) multiplied by $\frac{1}{\lambda_{1}}$ yields that

$$
\begin{equation*}
\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}}-\frac{1}{\lambda_{1}}\right) v_{k k 1}=-\left(n-1-\frac{\sigma_{1}(\lambda \mid 1)}{\lambda_{1}}\right) \frac{v_{1}}{v}+\varphi_{1} \tag{2.14}
\end{equation*}
$$

Applying (2.12) to $C_{1}$, we have

$$
\begin{align*}
C_{1}= & \frac{1}{\lambda_{1}^{2}}\left[\sum_{k=2}^{n} v_{k k 1}+\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right) \frac{v_{1}}{v}\right]^{2}+\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{1}}\right) v_{k k 1}^{2} \\
= & \frac{1}{\lambda_{1}^{2}}\left(\sum_{k=2}^{n} v_{k k 1}\right)^{2}+\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{1}}\right) v_{k k 1}^{2} \\
& +\frac{2}{\lambda_{1}^{2}}\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right) \frac{v_{1}}{v}\left(\sum_{k=2}^{n} v_{k k 1}\right)+\frac{1}{\lambda_{1}^{2}}\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right)^{2}\left(\frac{v_{1}}{v}\right)^{2} . \tag{2.15}
\end{align*}
$$

Under the condition (2.14), we will compute the minimum of $C_{1}$. Let

$$
\begin{aligned}
f\left(y_{2}, y_{3}, \ldots, y_{n}\right)= & \frac{1}{\lambda_{1}^{2}}\left(\sum_{k=2}^{n} y_{k}\right)^{2}+\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{1}}\right) y_{k}^{2} \\
& +\frac{2}{\lambda_{1}^{2}}\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right) \frac{v_{1}}{v}\left(\sum_{k=2}^{n} y_{k}\right) \\
& +\frac{1}{\lambda_{1}^{2}}\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right)^{2}\left(\frac{v_{1}}{v}\right)^{2}
\end{aligned}
$$

and

$$
g\left(y_{2}, y_{3}, \ldots, y_{n}\right)=\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}}-\frac{1}{\lambda_{1}}\right) y_{k}+\left(n-1-\frac{\sigma_{1}(\lambda \mid 1)}{\lambda_{1}}\right) \frac{v_{1}}{v},
$$

where we have omitted the gradient term $\nabla \varphi$. We compute the least value of $f$ under the condition $g=0$.

Case 1: $\quad \lambda_{1}=\lambda_{2}=\cdots=\lambda_{n}$. In this case, $g \equiv 0$ and only need to compute the minimum of

$$
f=\frac{1}{\lambda_{1}^{2}}\left(\sum_{k=2}^{n} y_{k}\right)^{2}+\frac{3}{\lambda_{1}^{2}} \sum_{k=2}^{n} y_{k}^{2}+\frac{2(n+2)}{\lambda_{1}} \frac{v_{1}}{v}\left(\sum_{k=2}^{n} y_{k}\right)+(n+2)^{2}\left(\frac{v_{1}}{v}\right)^{2} .
$$

Because $f$ is a quadratic polynomial and is strictly convex, $f$ has a unique critical point and must take the least value at this point. We only need to compute critical value of $f$. Taking partial derivatives for $f$ with respect to $y_{k}, k=2, \ldots, n$, we obtain

$$
\frac{\partial f}{\partial y_{k}}=\frac{2}{\lambda_{1}^{2}}\left(\sum_{i=2}^{n} y_{i}\right)+\frac{6}{\lambda_{1}^{2}} y_{k}+\frac{2(n+2)}{\lambda_{1}} \frac{v_{1}}{v}
$$

Let $\frac{\partial f}{\partial y_{k}}=0, k=2, \ldots, n$. Solving this system, we get

$$
y_{2}=y_{3}=\cdots=y_{n}=-\frac{\lambda_{1} v_{1}}{v} .
$$

At this unique critical point, $f$ takes the least value $(3 n+6) \frac{v_{1}^{2}}{\frac{v^{2}}{2}}$. Therefore,

$$
f\left(y_{2}, \ldots, y_{n}\right) \geq(3 n+6) \frac{v_{1}^{2}}{v^{2}}=\left(n+8+\frac{2 \sigma_{1}(\lambda \mid 1)}{\lambda_{1}}\right) \frac{v_{1}^{2}}{v^{2}} .
$$

Case 2: There exists $2 \leq i \leq n$ such that $\lambda_{i} \neq \lambda_{1}$. In this case, we use the Lagrange method of multipliers to calculate the minimum. Let

$$
\begin{aligned}
& L\left(y_{2}, y_{3}, \ldots, y_{n} ; \mu\right) \\
& \quad=f\left(y_{2}, y_{3}, \ldots, y_{n}\right)+\mu g\left(y_{2}, y_{3}, \ldots, y_{n}\right) \\
& \quad=\frac{1}{\lambda_{1}^{2}}\left(\sum_{k=2}^{n} y_{k}\right)^{2}+\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{1}}\right) y_{k}^{2}+\frac{2}{\lambda_{1}^{2}}\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right) \frac{v_{1}}{v}\left(\sum_{k=2}^{n} y_{k}\right) \\
& \quad+\frac{1}{\lambda_{1}^{2}}\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right)^{2}\left(\frac{v_{1}}{v}\right)^{2}+\mu\left[\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}}-\frac{1}{\lambda_{1}}\right) y_{k}+\left(n-1-\frac{\sigma_{1}(\lambda \mid 1)}{\lambda_{1}}\right) \frac{v_{1}}{v}\right] .
\end{aligned}
$$

Taking partial derivatives for $L$, we have, for $k=2, \ldots, n$,

$$
\begin{aligned}
\frac{\partial L}{\partial y_{k}} & =2\left(\frac{1}{\lambda_{1}}+\frac{1}{\lambda_{k}}\right)^{2} y_{k}+\frac{2}{\lambda_{1}^{2}} \sum_{i=2, i \neq k}^{n} y_{i}+\frac{2}{\lambda_{1}^{2}}\left(3 \lambda_{1}+\sigma_{1}(\lambda \mid 1)\right) \frac{v_{1}}{v}+\left(\frac{1}{\lambda_{k}}-\frac{1}{\lambda_{1}}\right) \mu, \\
\frac{\partial L}{\partial \mu} & =\sum_{i=2}^{n}\left(\frac{1}{\lambda_{i}}-\frac{1}{\lambda_{1}}\right) y_{i}+\left(n-1-\frac{\sigma_{1}(\lambda \mid 1)}{\lambda_{1}}\right) \frac{v_{1}}{v} .
\end{aligned}
$$

Let

$$
\frac{\partial L}{\partial y_{k}}=0, \quad k=2, \ldots, n, \quad \text { and } \quad \frac{\partial L}{\partial \mu}=0
$$

we have

Because of the positive definiteness of the matrix

$$
V=\left(\begin{array}{cccc}
2\left(\frac{1}{\lambda_{1}}+\frac{1}{\lambda_{2}}\right)^{2} & \frac{2}{\lambda_{1}^{2}} & \ldots & \frac{2}{\lambda_{1}^{2}} \\
\frac{2}{\lambda_{1}^{2}} & 2\left(\frac{1}{\lambda_{1}}+\frac{1}{\lambda_{3}}\right)^{2} & \ldots & \frac{2}{\lambda_{1}^{2}} \\
\ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \\
\frac{2}{\lambda_{1}^{2}} & \frac{2}{\lambda_{1}^{2}} & \ldots & 2\left(\frac{1}{\lambda_{1}}+\frac{1}{\lambda_{n}}\right)^{2}
\end{array}\right) \text { and } \mathbf{w}=\left(\begin{array}{c}
\frac{1}{\lambda_{2}}-\frac{1}{\lambda_{1}} \\
\frac{1}{\lambda_{3}}-\frac{1}{\lambda_{1}} \\
\vdots \\
\frac{1}{\lambda_{n}}-\frac{1}{\lambda_{1}}
\end{array}\right) \neq \mathbf{0} \text {, }
$$

$\operatorname{det}(U)=\left(-\mathbf{w}^{T} V^{-1} \mathbf{w}\right) \operatorname{det} V<0$ and so the coefficient matrix $U$ is nonsingular. We can solve the linear system above and obtain the unique minimal value point

$$
\left(y_{2}, y_{3}, \ldots, y_{n}\right)=-\left(\frac{\lambda_{2} v_{1}}{v}, \frac{\lambda_{3} v_{1}}{v}, \ldots, \frac{\lambda_{n} v_{1}}{v}\right) .
$$

At this point,

$$
f=\left(n+8+\frac{2 \sigma_{1}(\lambda \mid 1)}{\lambda_{1}}\right) \frac{v_{1}^{2}}{v^{2}}
$$

must be the least value. Consequently,

$$
f\left(y_{2}, \ldots, y_{n}\right) \geq\left(n+8+\frac{2 \sigma_{1}(\lambda \mid 1)}{\lambda_{1}}\right) \frac{v_{1}^{2}}{v^{2}}
$$

under the condition $g\left(y_{2}, \ldots, y_{n}\right)=0$.
From the two cases above, we get

$$
C_{1} \geq\left(n+8+\frac{2 \sigma_{1}(\lambda \mid 1)}{\lambda_{1}}\right) \frac{v_{1}^{2}}{v^{2}} \bmod (\nabla \varphi)
$$

and complete the claim (2.10).
Finally, we prove that the function $\psi_{1}$ attains its minimum in $\Omega$ if and only if $\Omega$ is an ellipsoid (ellipse). In fact, from the above process of the proof, we can see
further that

$$
\begin{aligned}
-\Delta \varphi= & -(A+B+C) \\
= & \sum_{i=1}^{n} \frac{1}{\lambda_{i}^{2}}\left(v_{i i}+\frac{3 \lambda_{i} v_{i}}{v}\right)^{2}+\sum_{\substack{i, k=1 \\
i \neq k}}^{n}\left(\frac{1}{\lambda_{i}^{2}}+\frac{2}{\lambda_{i} \lambda_{k}}\right)\left(v_{i i k}+\frac{\lambda_{i} v_{k}}{v}\right)^{2} \\
& +\sum_{\substack{i, j, k=1 \\
i \neq j, j \neq k, k \neq i}}^{n} \frac{v_{i j k}^{2}}{\lambda_{j} \lambda_{k}} .
\end{aligned}
$$

If $\psi_{1}$ attains its minimum in $\Omega$, then $\psi_{1}$ is constant from the strong minimum principle. Therefore, $\Delta \varphi=\Delta \log \psi_{1}=0$ and

$$
v_{i j k}= \begin{cases}-\frac{3 \lambda_{i} v_{i}}{v}, & i=j=k  \tag{2.16}\\ -\frac{\lambda_{i} v_{k}}{v}, & i=j, i \neq k \\ 0, & i \neq j, j \neq k, k \neq i\end{cases}
$$

where $i, j, k=1,2, \ldots, n$. Because of $u=v^{2}$,

$$
u_{i j k}=2\left(v_{i} v_{j k}+v_{j} v_{k i}+v_{k} v_{i j}+v v_{i j k}\right)
$$

Note that the matrix $\left(v_{i j}\right)$ is diagonal by the choice of the coordinates, from (2.16), we have

$$
u_{i j k}=0, \quad i, j, k=1,2, \ldots, n
$$

that is, all the third derivatives of $u$ are vanish. Since $\Omega=\left\{x \in \mathbb{R}^{n} \mid u(x)>0\right\}$ is convex, $\Omega$ must be an ellipsoid. On the contrary, if $\Omega$ is an ellipsoid, then, up to a translation,

$$
\Omega=\left\{\left(x_{1}, x_{2}, \ldots, x_{n}\right) \in \mathbb{R}^{n} \left\lvert\, \frac{x_{1}^{2}}{a_{1}^{2}}+\frac{x_{2}^{2}}{a_{2}^{2}}+\cdots+\frac{x_{n}^{2}}{a_{n}^{2}}<1\right., a_{i}>0, i=1,2, \ldots, n\right\} .
$$

At this time, for the problem (1.1), the solution

$$
u=-b\left(\frac{x_{1}^{2}}{a_{1}^{2}}+\frac{x_{2}^{2}}{a_{2}^{2}}+\cdots+\frac{x_{n}^{2}}{a_{n}^{2}}-1\right)
$$

where $b=\left(\frac{1}{a_{1}^{2}}+\frac{1}{a_{2}^{2}}+\cdots+\frac{1}{a_{n}^{2}}\right)^{-1}$, and the function $\psi_{1} \equiv \frac{b^{n+1}}{a_{1}^{a_{2}^{2}} \cdots a_{n}^{2}}$ is constant. Naturally, $\psi_{1}$ attains its minimum in $\Omega$.

Lemma 2.1 ([19]). If u is a smooth positive solution of the problem (1.1), then the function $|\nabla u|^{2}+\frac{4 u}{n}$ attains its maximum on the boundary $\partial \Omega$ and satisfies

$$
\begin{equation*}
\max _{\Omega}\left\{|\nabla u|^{2}+\frac{4 u}{n}\right\} \leq \max _{\partial \Omega}|\nabla u|^{2} \tag{2.17}
\end{equation*}
$$

Lemma 2.2 ([3]). Let $\Omega$ be a smooth, bounded, and strictly convex domain in $\mathbb{R}^{n}$, $x \in \partial \Omega$, and $\kappa_{i}(x), i=1,2, \ldots, n-1$, all principal curvatures of $\partial \Omega$ at $x$. Let

$$
\begin{aligned}
\kappa_{m}(x) & =\min \left\{\kappa_{i}(x) \mid i=1,2, \ldots, n-1\right\}, \kappa_{M}(x)=\max \left\{\kappa_{i}(x) \mid i=1,2, \ldots, n-1\right\}, \\
\kappa_{\text {min }} & =\min \left\{\kappa_{m}(x) \mid x \in \partial \Omega\right\} \text { and } \kappa_{\max }=\max \left\{\kappa_{M}(x) \mid x \in \partial \Omega\right\} .
\end{aligned}
$$

If $u$ is a smooth positive solution of the problem (1.1), then, on the boundary $\partial \Omega$, the modulus of its gradient $|\nabla u|_{\partial \Omega}$ satisfies the following estimate

$$
\begin{equation*}
\frac{2}{n \kappa_{\max }} \leq|\nabla u|_{\partial \Omega} \leq \frac{2}{n \kappa_{\min }} . \tag{2.18}
\end{equation*}
$$

Proof. For the completion of the paper, we give the proof here. For any boundary point $x$, let $\Omega \subseteq \Omega_{0}$ and $\Omega_{1} \subseteq \Omega$ be two balls of radius $R=\kappa_{\min }^{-1}$ and $r=\kappa_{\max }^{-1}$ with the property that $x \in \bar{\Omega} \cap \bar{\Omega}_{j}, j=0,1$. Let $u_{\Omega_{j}}, j=0,1$, be the solution to the problem

$$
\begin{cases}\Delta u=-2 & \text { in } \Omega_{j} \\ u=0 & \text { on } \partial \Omega_{j}\end{cases}
$$

Since $u$ vanishes on $\partial \Omega$, it follows immediately that

$$
\left|\nabla u_{\Omega_{1}}(x)\right| \leq|\nabla u(x)| \leq\left|\nabla u_{\Omega_{0}}(x)\right| .
$$

An explicit calculation yields

$$
\left|\nabla u_{\Omega_{1}}(x)\right|=\frac{2 r}{n}, \quad\left|\nabla u_{\Omega_{0}}(x)\right|=\frac{2 R}{n}
$$

and so

$$
\frac{2 r}{n} \leq|\nabla u(x)| \leq \frac{2 R}{n} .
$$

Therefore,

$$
\frac{2}{n \kappa_{\max }} \leq|\nabla u|_{\partial \Omega} \leq \frac{2}{n \kappa_{\min }} .
$$

Next, we start the proof of Corollary 1.3.
Proof. Since $v=-\sqrt{u}$, we have $|\nabla v|^{2}=\frac{|\nabla u|^{2}}{4 u}$ and

$$
\operatorname{det} D^{2} v=(-2)^{-n} u^{-\frac{n}{2}} \operatorname{det} D^{2} u+(-2)^{-n-1} u^{-\frac{n+2}{2}} \sum_{i, j=1}^{n} \frac{\partial \operatorname{det} D^{2} u}{\partial u_{i j}} u_{i} u_{j} .
$$

So

$$
K_{G}=\frac{\operatorname{det} D^{2} v}{\left(1+|\nabla v|^{2}\right)^{\frac{n+2}{2}}}
$$

$$
\begin{align*}
& =\frac{2^{n+2}(-v)^{n+2} \operatorname{det} D^{2} v}{2^{n+2}(-v)^{n+2}\left(1+|\nabla v|^{2}\right)^{\frac{n+2}{2}}} \\
& =\frac{2^{n+2} \psi_{1}}{\left(4 u+|\nabla u|^{2}\right)^{\frac{n+2}{2}}} . \tag{2.19}
\end{align*}
$$

By the estimate (1.5) in Theorem 1.1 and (2.18) in Lemma 2.2, we have

$$
\begin{align*}
\min _{\bar{\Omega}} \psi_{1} & \geq 2^{-n-1}\left(\min _{\partial \Omega} K\right)\left(\frac{2}{n \kappa_{\max }}\right)^{n+1} \\
& =\frac{K_{\min }}{n^{n+1} \kappa_{\max }^{n+1}} \tag{2.20}
\end{align*}
$$

For the term $\left(4 u+|\nabla u|^{2}\right)^{\frac{n+2}{2}}$ in $\Omega$, by the (2.17) and (2.18), we have

$$
\begin{align*}
4 u+|\nabla u|^{2} & =n\left(\frac{|\nabla u|^{2}}{n}+\frac{4 u}{n}\right) \\
& \leq n \max _{\Omega}\left\{|\nabla u|^{2}+\frac{4 u}{n}\right\} \\
& \leq n \max _{\partial \Omega}|\nabla u|^{2} \\
& \leq \frac{4}{n \kappa_{\min }^{2}} . \tag{2.21}
\end{align*}
$$

Therefore, applying (2.20) and (2.21) to (2.19), we obtain

$$
\begin{equation*}
K_{G}(x) \geq \frac{2^{n+2} \frac{K_{\min }}{n^{n+1} \kappa_{\max }^{n+1}}}{\left(\frac{4}{n \kappa_{\min }^{2}}\right)^{\frac{n+2}{2}}}=\frac{K_{\min } \kappa_{\min }^{n+2}}{n^{\frac{n}{2}} \kappa_{\max }^{n+1}} \quad x \in \Omega, \tag{2.22}
\end{equation*}
$$

which is exactly (1.8). If $\Omega$ is the unit ball $B_{1}(0)=\left\{\left.x \in \mathbb{R}^{n}| | x\right|^{2}<1\right\}$, then $u=\frac{1-|x|^{2}}{n}$ is the solution to the problem (1.1), $v=-\sqrt{u}=-\sqrt{\frac{1-|x|^{2}}{n}}$ and $\kappa_{\max }=\kappa_{\min }=K_{\min }=$ 1. By the direct calculations, we get

$$
K_{G}(x)=\frac{\operatorname{det} D^{2} v(x)}{\left(1+|\nabla v(x)|^{2}\right)^{\frac{n+2}{2}}}=n^{-\frac{n}{2}}\left(1-|x|^{2}+\frac{|x|^{2}}{n}\right)^{-\frac{n+2}{2}}
$$

$K_{G}(0)=n^{-\frac{n}{2}}$, and the equality holds at the origin 0 in (2.22). We have completed the proof of Corollary 1.3.

Remark 2.3. When $n=2$, Corollary 1.3 is exactly Theorem 3.4 in [14].
Remark 2.4. From the Theorem 1.1, we can combine the deformation process to give a new proof of strict $\frac{1}{2}$-concavity of the solution $u$ for (1.1), i.e., $v=-\sqrt{u}$ is strictly convex when $\Omega$ is a smooth, bounded and strictly convex domain. The deformation process is well-known and can be found in Ma and Xu [16] or Ma et al. [15].

Suppose $0 \in \Omega$. At the initial time, let the domain be the standard unit ball $B_{1}(0)$ and

$$
\Omega_{t}=(1-t) B_{1}(0)+t \Omega, \quad 0 \leq t \leq 1,
$$

where the sum is the Minkowski vector sum. So the domain $\Omega_{t}$ is a family of smooth and strictly convex domains. We assume the function $v_{t}$ satisfies the problem (2.1) in the domain $\Omega_{t}$, that is

$$
\begin{cases}\Delta v_{t}=-\frac{1+\left|\nabla v_{t}\right|^{2}}{v_{t}} & \text { in } \Omega_{t}, \\ v_{t}=0 & \text { on } \partial \Omega_{t} .\end{cases}
$$

Now we prove the strict convexity of $v$. Because $\Omega_{t}, 0 \leq t \leq 1$, are all strictly convex, the Gauss curvature of $\partial \Omega_{t}$ have an uniformly positive lower bound. Suppose that $v$ is not strictly convex in $\Omega$. Then there exists the first time $0<t_{o} \leq 1$ such that the determinant det $D^{2} v_{t_{o}}$ for the Hessian of $v_{t_{o}}$ becomes zero at some point $x_{t_{o}} \in$ $\Omega_{t_{o}}$. Taking a sequence $\left\{t_{k}\right\}$ such that $0<t_{k}<t_{o}$ and $t_{k} \rightarrow t_{o}, k \rightarrow \infty$, then from the estimate (1.5) in Theorem 1.1 and the Hopf's Lemma, we get an uniformly positive lower bound for the function sequence $\left(-v_{t_{k}}\right)^{n+2} \operatorname{det} D^{2} v_{t_{k}}, k=1,2, \ldots$. By the Schauder estimates and taking limit for $k$, we get a positive lower bound on the function $\left(-v_{t_{o}}\right)^{n+2} \operatorname{det} D^{2} v_{t_{o}}$. This is a contradiction. Then we complete the proof of the strict convexity of $v$ on the strictly convex domain case.

For the general convex domain, we can first use the approximation with strictly convex domain to get the convexity of $v$ and then obtain its strict convexity through using the constant rank theorem of the Hessian of $v_{t}$ by Korevaar and Lewis [13].

## 3. Convexity Estimates for the Eigenvalue Problem

We now give the proof of Theorem 1.2.
Proof. The process of the proof is similar to that of Theorem 1.1.
Suppose that $u>0$ is the first eigenfunction of Laplace operator, i.e., $u$ is the solution for the eigenvalue problem (1.2) with $\lambda>0$ being the first eigenvalue. From the assumption, $v=-\log u$ is strictly convex and satisfies the following Dirichlet problem

$$
\begin{cases}\Delta v=\lambda+|\nabla v|^{2} & \text { in } \Omega  \tag{3.1}\\ v(x) \rightarrow+\infty & \text { as } x \rightarrow \partial \Omega\end{cases}
$$

For

$$
\psi_{2}=e^{-(n+1) v} \operatorname{det} D^{2} v=(-1)^{n} u \operatorname{det} D^{2} u+(-1)^{n-1} \sum_{i, j=1}^{n} \frac{\partial \operatorname{det} D^{2} u}{\partial u_{i j}} u_{i} u_{j},
$$

we shall show that

$$
\varphi=\log \psi_{2}=-(n+1) v+\log \operatorname{det} D^{2} v
$$

satisfies the following elliptic differential inequality:

$$
\begin{equation*}
\Delta \varphi \leq 0 \bmod (\nabla \varphi) \text { in } \Omega, \tag{3.2}
\end{equation*}
$$

which implies the inequality (1.6). Moreover, we obtain that the function $\psi_{2}$ attains its minimum on the boundary by the standard minimum principle. Therefore, by (1.3), we have

$$
\begin{aligned}
\psi_{2} & =e^{-(n+1) v} \operatorname{det} D^{2} v \\
& \geq \min _{\partial \Omega} \psi_{2}=\min _{\partial \Omega}\left\{K|\nabla u|^{n+1}\right\} \\
& \geq \min _{\partial \Omega} K \min _{\partial \Omega}|\nabla u|^{n+1},
\end{aligned}
$$

which is the estimate (1.7).
As in the proof of Theorem 1.1, in order to prove (3.2) at an arbitrary point $x_{o}$, we only need to establish (3.2) under the assumption that the matrix $\left(v_{i j}\left(x_{o}\right)\right)(1 \leq$ $i, j \leq n$ ) is diagonal. From now on, all the calculation will be done at the fixed point $x_{0}$.

Because $v$ is strictly convex, the Hessian matrix $\left(v_{i j}\right)$ is positive definite. Let $\lambda_{i}=$ $v_{i i}>0$ and $\left(v^{i j}\right)$ the inverse matrix of $\left(v_{i j}\right)$. Taking the first derivative of $\varphi$, we have

$$
\begin{equation*}
\varphi_{i}=-(n+1) v_{i}+\sum_{k, l=1}^{n} v^{k l} v_{k l i} \tag{3.3}
\end{equation*}
$$

and it follows that

$$
\begin{equation*}
\sum_{k=1}^{n} v^{k k} v_{k k i}=(n+1) v_{i}+\varphi_{i} \tag{3.4}
\end{equation*}
$$

Taking the second derivative of $\varphi$, we get

$$
\begin{equation*}
\varphi_{i i}=-(n+1) v_{i i}+\sum_{k, l=1}^{n} v^{k l} v_{k l i i}-\sum_{k, l, p, q=1}^{n} v^{k q} v^{p l} v_{k l i} v_{p q i} . \tag{3.5}
\end{equation*}
$$

Then

$$
\begin{equation*}
\Delta \varphi=-(n+1) \Delta v+\sum_{k}^{n} v^{k k} \Delta\left(v_{k k}\right)-\sum_{k, l, i=1}^{n} v^{k k} v^{l l} v_{k l i}^{2} \triangleq E+F+G, \tag{3.6}
\end{equation*}
$$

where

$$
\begin{aligned}
& E=-(n+1) \Delta v, \\
& F=\sum_{k=1}^{n} v^{k k} \Delta\left(v_{k k}\right), \\
& G=-\sum_{k, l, i=1}^{n} v^{k k} v^{l l} v_{k l i}^{2} .
\end{aligned}
$$

For the term $F$, taking derivative of the equation in (3.1), it follows that

$$
(\Delta v)_{k}=2 \sum_{i=1}^{n} v_{i} v_{i k},
$$

$$
\begin{aligned}
(\Delta v)_{k k} & =2 \sum_{i=1}^{n}\left(v_{i k}^{2}+v_{i} v_{i k k}\right) \\
& =2\left[v_{k k}^{2}+\sum_{i=1}^{n} v_{i} v_{i k k}\right] .
\end{aligned}
$$

Thus,

$$
\begin{align*}
F & =\sum_{k}^{n} v^{k k} \Delta\left(v_{k k}\right)=\sum_{k}^{n} v^{k k}(\Delta v)_{k k} \\
& =2 \Delta v+2 \sum_{i, k=1}^{n}\left(v_{i} v_{i k k} v^{k k}\right) \\
& =2 \Delta v+2(n+1)|\nabla v|^{2}+2 \sum_{i=1}^{n} v_{i} \varphi_{i} \tag{3.7}
\end{align*}
$$

where we have used (3.4) in the last equality. Therefore, using (3.7), we yield that

$$
\begin{align*}
E+F & =(1-n) \Delta v+2(n+1)|\nabla v|^{2}+2 \sum_{i=1}^{n} v_{i} \varphi_{i} \\
& =(1-n) \lambda+(n+3)|\nabla v|^{2}+2 \sum_{i=1}^{n} v_{i} \varphi_{i} \\
& <(n+3)|\nabla v|^{2} \bmod (\nabla \varphi) \tag{3.8}
\end{align*}
$$

where we have used the equation in (3.1) and $\lambda>0$. Next, we treat the term $G$ as $C$ in the proof of Theorem 1.1, we have

$$
-G \geq \sum_{l=1}^{n}\left(\frac{v_{l l}^{2}}{\lambda_{l}^{2}}+\sum_{k=1, k \neq l}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{l}}\right) v_{k k l}^{2}\right) \triangleq \sum_{l=1}^{n} G_{l},
$$

where

$$
G_{l}=\frac{v_{l l}^{2}}{\lambda_{l}^{2}}+\sum_{k=1, k \neq l}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{l}}\right) v_{k k l}^{2}, \quad l=1, \ldots, n
$$

We claim that

$$
\begin{equation*}
G_{l} \geq(n+3) v_{l}^{2} \bmod (\nabla \varphi), \quad l=1,2, \ldots, n \tag{3.9}
\end{equation*}
$$

From this claim, we have

$$
\begin{equation*}
G \leq-\sum_{l=1}^{n} G_{l} \leq-(n+3)|\nabla v|^{2} \bmod (\nabla \varphi) . \tag{3.10}
\end{equation*}
$$

Consequently, (3.6), (3.8) and (3.10) yield

$$
\Delta \varphi \leq 0 \quad \bmod (\nabla \varphi)
$$

which is exactly (3.2) and we have completed the proof.

Now, we prove the claim for $l=1$ and others are the same completely. Taking derivative of the equation in (3.1) with respect to $x_{1}$, we get the equality

$$
(\Delta v)_{1}=2 \sum_{i=1}^{n} v_{i} v_{i 1}=2 \lambda_{1} v_{1}
$$

that is

$$
\begin{equation*}
v_{111}=-\sum_{k=2}^{n} v_{k k 1}+2 \lambda_{1} v_{1} \tag{3.11}
\end{equation*}
$$

From (3.4), we have

$$
\begin{equation*}
\sum_{k=1}^{n} \frac{v_{k k 1}}{\lambda_{k}}=(n+1) v_{1}+\varphi_{1} . \tag{3.12}
\end{equation*}
$$

That (3.12) subtracts (3.11) multiplied by $\frac{1}{\lambda_{1}}$ yields that

$$
\begin{equation*}
\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}}-\frac{1}{\lambda_{1}}\right) v_{k k 1}=(n-1) v_{1}+\varphi_{1} \tag{3.13}
\end{equation*}
$$

and

$$
\begin{equation*}
v_{1}=\frac{1}{n-1} \sum_{k=2}^{n} \frac{\lambda_{1}-\lambda_{k}}{\lambda_{1} \lambda_{k}} v_{k k 1} \bmod (\nabla \varphi) \tag{3.14}
\end{equation*}
$$

Applying (3.11) to $G_{1}$, we have

$$
\begin{align*}
G_{1} & =\frac{1}{\lambda_{1}^{2}}\left(\sum_{k=2}^{n} v_{k k 1}-2 \lambda_{1} v_{1}\right)^{2}+\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{1}}\right) v_{k k 1}^{2} \\
& =\frac{1}{\lambda_{1}^{2}}\left(\sum_{k=2}^{n} v_{k k 1}\right)^{2}+\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{1}}\right) v_{k k 1}^{2}-\frac{4}{\lambda_{1}} v_{1}\left(\sum_{k=2}^{n} v_{k k 1}\right)+4 v_{1}^{2} . \tag{3.15}
\end{align*}
$$

From (3.14) and (3.15), we get that

$$
\begin{aligned}
G_{1} & -(n+3) v_{1}^{2} \\
= & \frac{1}{\lambda_{1}^{2}}\left(\sum_{k=2}^{n} v_{k k 1}\right)^{2}+\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{1}}\right) v_{k k 1}^{2} \\
& -\frac{4}{(n-1) \lambda_{1}^{2}}\left(\sum_{k=2}^{n} \frac{\lambda_{1}-\lambda_{k}}{\lambda_{k}} v_{k k 1}\right)\left(\sum_{l=2}^{n} v_{l l 1}\right)-\frac{1}{(n-1) \lambda_{1}^{2}}\left(\sum_{k=2}^{n} \frac{\lambda_{1}-\lambda_{k}}{\lambda_{k}} v_{k k 1}\right)^{2} \\
= & \frac{1}{\lambda_{1}^{2}}\left(\sum_{k=2}^{n} v_{k k 1}\right)^{2}+\sum_{k=2}^{n}\left(\frac{1}{\lambda_{k}^{2}}+\frac{2}{\lambda_{k} \lambda_{1}}\right) v_{k k 1}^{2} \\
& -\frac{1}{(n-1) \lambda_{1}^{2}}\left(\sum_{k=2}^{n} \frac{\lambda_{1}-\lambda_{k}}{\lambda_{k}} v_{k k 1}\right)\left(\sum_{l=2}^{n} \frac{\lambda_{1}+3 \lambda_{l}}{\lambda_{l}} v_{l l 1}\right)
\end{aligned}
$$

$$
\begin{align*}
= & \frac{1}{(n-1) \lambda_{1}^{2}}\left[\sum_{k=2}^{n}\left(4+(n-2)\left(1+\frac{\lambda_{1}}{\lambda_{k}}\right)^{2}\right) v_{k k 1}^{2}\right. \\
& \left.+2 \sum_{k, l=2, k<l}^{n}\left(n+3-\left(1+\frac{\lambda_{1}}{\lambda_{k}}\right)\left(1+\frac{\lambda_{1}}{\lambda_{l}}\right)\right) v_{k k 1} v_{l l 1}\right] \bmod (\nabla \varphi) . \tag{3.16}
\end{align*}
$$

By the Lemma 3.1, the matrix
is semi-positive definite for $a_{i}=1+\frac{\lambda_{1}}{\lambda_{i+1}}$ and $m=n-1$. Therefore, by (3.16), we obtain

$$
G_{1}-(n+3) v_{1}^{2} \geq 0 \quad \bmod (\nabla \varphi) .
$$

The claim (3.9) is true.
Now, we give the following lemma for the semi-positive definiteness of a class of matrices.

Lemma 3.1. Let $m \geq 1$ be an integer, $a_{i} \geq 1, i=1, \ldots, m$, and

$$
M=\left(\begin{array}{c}
4+(m-1) a_{1}^{2} m+4-a_{1} a_{2} \ldots m+4-a_{1} a_{m} \\
m+4-a_{1} a_{2} 4+(m-1) a_{2}^{2} \ldots m+4-a_{2} a_{m} \\
\ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \\
m+4-a_{m} a_{1} m+4-a_{m} a_{2} \ldots 4+(m-1) a_{m}^{2}
\end{array}\right) .
$$

Then the matrix $M$ is semi-positive definite.
Proof. The conclusion is trivial for $m=1$ and so we only consider $m \geq 2$. Let

$$
M_{i}^{(1)}=\left(\begin{array}{c}
m+4-a_{i} a_{1} \\
m+4-a_{i} a_{2} \\
\vdots \\
m+4-a_{i} a_{m}
\end{array}\right), \quad M_{i}^{(2)}=\left(\begin{array}{c}
0 \\
\vdots \\
0 \\
m\left(a_{i}^{2}-1\right) \\
0 \\
\vdots \\
0
\end{array}\right) \leftarrow i \text { th row. }
$$

Then,

$$
M=\left(M_{1}^{(1)}+M_{1}^{(2)}, M_{2}^{(1)}+M_{2}^{(2)}, \ldots, M_{m}^{(1)}+M_{m}^{(2)}\right)
$$

and we have

$$
\operatorname{det} M=\sum_{i_{1}, i_{2}, \ldots, i_{m}=1}^{2} \operatorname{det}\left(M_{1}^{\left(i_{1}\right)}, M_{2}^{\left(i_{2}\right)}, \ldots, M_{m}^{\left(i_{m}\right)}\right) .
$$

Because

$$
\left(M_{1}^{(1)}, M_{2}^{(1)}, \ldots, M_{m}^{(1)}\right)=(m+4)(1)_{m \times m}-\left(a_{i} a_{j}\right)_{m \times m}
$$

and the rank of both matrices $(m+4)(1)_{m \times m}$ and $\left(a_{i} a_{j}\right)_{m \times m}$ is 1 , we know that the rank of $\left(M_{1}^{(1)}, M_{2}^{(1)}, \ldots, M_{m}^{(1)}\right)$ is less than or equal to 2 . Therefore,

$$
\begin{align*}
\operatorname{det} M= & \sum_{i_{1}, i_{2}, \ldots, i_{m}=1}^{2} \operatorname{det}\left(M_{1}^{\left(i_{1}\right)}, M_{2}^{\left(i_{2}\right)}, \ldots, M_{m}^{\left(i_{m}\right)}\right) \\
= & \sum_{\substack{i, j=1 \\
i<j}}^{m} \operatorname{det}\left(M_{1}^{(2)}, \ldots, M_{i-1}^{(2)}, M_{i}^{(1)}, M_{i+1}^{(2)}, \ldots, M_{j-1}^{(2)}, M_{j}^{(1)}, M_{j+1}^{(2)}, \ldots, M_{m}^{(2)}\right) \\
& +\sum_{i=1}^{m} \operatorname{det}\left(M_{1}^{(2)}, \ldots, M_{i-1}^{(2)}, M_{i}^{(1)}, M_{i+1}^{(2)}, \ldots, M_{m}^{(2)}\right) \\
& +\operatorname{det}\left(M_{1}^{(2)}, \ldots, M_{m}^{(2)}\right) . \tag{3.17}
\end{align*}
$$

We will compute the three terms respectively in the last equality above.

$$
\begin{align*}
& I \triangleq \sum_{\substack{i, j=1 \\
i<j}}^{m} \operatorname{det}\left(M_{1}^{(2)}, \ldots, M_{i-1}^{(2)}, M_{i}^{(1)}, M_{i+1}^{(2)}, \ldots, M_{j-1}^{(2)}, M_{j}^{(1)}, M_{j+1}^{(2)}, \ldots, M_{m}^{(2)}\right) \\
& =\sum_{\substack{i, j=1 \\
i<j}}^{m}\left(\left(m+4-a_{i}^{2}\right)\left(m+4-a_{j}^{2}\right)-\left(m+4-a_{i} a_{j}\right)^{2}\right) \prod_{\substack{k=1 \\
k \neq i, k \neq j}}^{m} m\left(a_{k}^{2}-1\right) \\
& =-(m+4) m^{m-2} \sum_{\substack{i, j=1 \\
i<j}}^{m}\left(a_{i}-a_{j}\right)^{2} \prod_{\substack{k=1 \\
k \neq i, k \neq j}}^{m}\left(a_{k}^{2}-1\right),  \tag{3.18}\\
& I I \triangleq \sum_{i=1}^{m} \operatorname{det}\left(M_{1}^{(2)}, \ldots, M_{i-1}^{(2)}, M_{i}^{(1)}, M_{i+1}^{(2)}, \ldots, M_{m}^{(2)}\right) \\
& =\sum_{i=1}^{m}\left(m+4-a_{i}^{2}\right) \prod_{\substack{k=1 \\
k \neq i}}^{m} m\left(a_{k}^{2}-1\right) \\
& =m^{m-1} \sum_{i=1}^{m}\left(m+4-a_{i}^{2}\right) \prod_{\substack{k=1 \\
k \neq i}}^{m}\left(a_{k}^{2}-1\right), \tag{3.19}
\end{align*}
$$

and

$$
\begin{align*}
I I I & \triangleq \operatorname{det}\left(M_{1}^{(2)}, \ldots, M_{m}^{(2)}\right) \\
& =\prod_{k=1}^{m} m\left(a_{k}^{2}-1\right)=m^{m-1} \sum_{i=1}^{m}\left(a_{i}^{2}-1\right) \prod_{\substack{k=1 \\
k \neq i}}^{m}\left(a_{k}^{2}-1\right) . \tag{3.20}
\end{align*}
$$

Combining (3.19) with (3.20), we get

$$
\begin{align*}
I I+I I I & =m^{m-1} \sum_{i=1}^{m}\left(m+4-a_{i}^{2}\right) \prod_{\substack{k=1 \\
k \neq i}}^{m}\left(a_{k}^{2}-1\right)+m^{m-1} \sum_{i=1}^{m}\left(a_{i}^{2}-1\right) \prod_{\substack{k=1 \\
k \neq i}}^{m}\left(a_{k}^{2}-1\right) \\
& =m^{m-1} \sum_{i=1}^{m}(m+3) \prod_{\substack{k=1 \\
k \neq i}}^{m}\left(a_{k}^{2}-1\right) \\
& =\frac{(m+3) m^{m-1}}{m-1} \sum_{\substack{i, j=1 \\
i \neq j}}^{m}\left(a_{j}^{2}-1\right) \prod_{\substack{k=1 \\
k \neq i, k \neq j}}^{m}\left(a_{k}^{2}-1\right) \\
& =\frac{(m+3) m^{m-1}}{2(m-1)} \sum_{\substack{i, j=1 \\
i \neq j}}^{m}\left[\left(a_{i}^{2}-1\right)+\left(a_{j}^{2}-1\right)\right] \prod_{\substack{k=1 \\
k \neq i, k \neq j}}^{m}\left(a_{k}^{2}-1\right) \\
& =\frac{(m+3) m^{m-1}}{m-1} \sum_{\substack{i, j=1 \\
i<j}}^{m}\left(a_{i}^{2}+a_{j}^{2}-2\right) \prod_{\substack{k=1 \\
k \neq i, k \neq j}}^{m}\left(a_{k}^{2}-1\right) \tag{3.21}
\end{align*}
$$

Combining the above (3.21) with (3.18), we have

$$
\begin{aligned}
\operatorname{det}(M)= & I+I I+I I I \\
= & -(m+4) m^{m-2} \sum_{\substack{i, j=1 \\
i<j}}^{m}\left(a_{i}-a_{j}\right)^{2} \prod_{\substack{k=1 \\
k \neq i, k \neq j}}^{m}\left(a_{k}^{2}-1\right) \\
& +\frac{(m+3) m^{m-1}}{m-1} \sum_{\substack{i, j=1 \\
i<j}}^{m}\left(a_{i}^{2}+a_{j}^{2}-2\right) \prod_{\substack{k=1 \\
k \neq i, k \neq j}}^{m}\left(a_{k}^{2}-1\right) \\
= & \frac{m^{m-2}}{m-1} \sum_{\substack{i, j=1 \\
i<j}}^{m}\left[m(m+3)\left(a_{i}^{2}+a_{j}^{2}-2\right)\right. \\
& \left.-(m+4)(m-1)\left(a_{i}-a_{j}\right)^{2}\right] \prod_{\substack{k=1 \\
k \neq i, k \neq j}}^{m}\left(a_{k}^{2}-1\right) \\
= & \frac{m^{m-2}}{m-1} \sum_{\substack{i, j=1 \\
i<j}}^{m}\left[4\left(a_{i}-a_{j}\right)^{2}+2 m(m+3)\left(a_{i} a_{j}-1\right)\right] \prod_{\substack{k=1 \\
k \neq i, k \neq j}}^{m}\left(a_{k}^{2}-1\right) \\
\geq & 0
\end{aligned}
$$

by $a_{i} \geq 1, i=1,2, \ldots, m$.

For the $k$ th principal minor $[M]_{k}$ of $M, k=2, \ldots, m$, according to the same method above, we can get that

$$
\begin{aligned}
& {[M]_{k}=\operatorname{det}\left(\begin{array}{cccc}
4+(m-1) a_{1}^{2} & m+4-a_{1} a_{2} & \ldots & m+4-a_{1} a_{k} \\
m+4-a_{1} a_{2} & 4+(m-1) a_{2}^{2} & \ldots & m+4-a_{2} a_{k} \\
\ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \\
m+4-a_{k} a_{1} & m+4-a_{k} a_{2} & \ldots & 4+(m-1) a_{k}^{2}
\end{array}\right)} \\
& =-(m+4) m^{k-2} \sum_{\substack{i, j=1 \\
i<j}}^{k}\left(a_{i}-a_{j}\right)^{2} \prod_{\substack{l=1 \\
l \neq i, l \neq j}}^{k}\left(a_{l}^{2}-1\right) \\
& +m^{k-1} \sum_{i=1}^{k}\left(m+4-a_{i}^{2}\right) \prod_{\substack{l=1 \\
l \neq i}}^{k}\left(a_{l}^{2}-1\right)+\frac{m^{k}}{k} \sum_{i=1}^{k}\left(a_{i}^{2}-1\right) \prod_{\substack{l=1 \\
l \neq i}}^{k}\left(a_{l}^{2}-1\right) \\
& \geq-(m+4) m^{k-2} \sum_{\substack{i, j=1 \\
i<j}}^{k}\left(a_{i}-a_{j}\right)^{2} \prod_{\substack{l=1 \\
l \neq i, l \neq j}}^{k}\left(a_{l}^{2}-1\right)+(m+3) m^{k-1} \sum_{i=1}^{k} \prod_{\substack{l=1 \\
l \neq i}}^{k}\left(a_{l}^{2}-1\right) \\
& \geq-(m+4) m^{k-2} \sum_{\substack{i, j=1 \\
i<j}}^{k}\left(a_{i}-a_{j}\right)^{2} \prod_{\substack{l=1 \\
l \neq i, l \neq j}}^{k}\left(a_{l}^{2}-1\right) \\
& +\frac{(m+3) m^{k-1}}{m-1} \sum_{\substack{i, j=1 \\
i<j}}^{k}\left(a_{i}^{2}+a_{j}^{2}-2\right) \prod_{\substack{l=1 \\
l \neq i, 1 \neq j}}^{k}\left(a_{l}^{2}-1\right) \\
& =\frac{m^{k-2}}{m-1} \sum_{\substack{i, j=1 \\
i<j}}^{k}\left[4\left(a_{i}-a_{j}\right)^{2}+2 m(m+3)\left(a_{i} a_{j}-1\right)\right] \prod_{\substack{l=1 \\
l \neq 1, l \neq j}}^{k}\left(a_{l}^{2}-1\right) \\
& \geq 0,
\end{aligned}
$$

and $[M]_{1}=4+(m-1) a_{1}^{2}>0$. Up to now, we obtain that all principal minors of $M$ are nonnegative and so the matrix $M$ is semi-positive definite.

Remark 3.2. As the discussions in Remark 2.4, from the Theorem 1.2, we can combine the deformation process to give a new proof of log-concavity of the first eigenfunction $u$ for the problem (1.2), i.e., $v=-\log u$ is strictly convex when $\Omega$ is a smooth, bounded and convex domain.

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