UNIQUENESS OF CLOSED SELF-SIMILAR SOLUTIONS TO σ_k^{α} -CURVATURE FLOW

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ABSTRACT. By adapting the test functions introduced by Choi-Daskaspoulos [11] and Brendle-Choi-Daskaspoulos [9] and exploring properties of the k-th elementary symmetric functions σ_k intensively, we show that for any fixed k with $1 \le k \le n-1$, any strictly convex closed hypersurface in \mathbb{R}^{n+1} satisfying $\sigma_k^{\alpha} = \langle X, \nu \rangle$, with $\alpha \ge \frac{1}{k}$, must be a round sphere.

1. INTRODUCTION

Let $X: M \to \mathbb{R}^{n+1}$ be a smooth embedding of a closed, orientable hypersurface in \mathbb{R}^{n+1} with $n \ge 2$, satisfying

(1.1)
$$\sigma_k^{\alpha} = \langle X, \nu \rangle$$

where ν is the outward unit normal vector field of M, $\alpha > 0$, $1 \le k \le n$ and σ_k is the k-th elementary symmetric functions of principal curvatures of M.

This type of equation is important for the following curvature flow

(1.2)
$$\tilde{X}_t = -\sigma_k^{\alpha} \nu.$$

Actually, if X is a solution of (1.1), then

$$\tilde{X}(x,t) = ((k\alpha + 1)(T-t))^{\frac{1}{1+k\alpha}}X(x)$$

gives rise to the solution of (1.2) up to a tangential diffeomorphism [20]. So in the same spirit, we call the solutions of (1.1) self-similar solutions of (1.2).

For k = 1, G. Huisken proved the following famous result:

Theorem 1.1 (Huisken, [18]). If M is a closed hypersurface in \mathbb{R}^{n+1} , with nonnegative mean curvature σ_1 and satisfies the equation

$$\sigma_1 = \langle X, \nu \rangle,$$

then M must be a round sphere.

For k = n, very recently, Choi-Daskalopoulos [11], further, Brendle-Choi-Daskalopoulos [9] proved the following remarkable result:

Theorem 1.2 (Choi-Daskalopoulos [11], Brendle-Choi-Daskalopoulos [9]). Let M be a closed, strictly convex hypersurface in \mathbb{R}^{n+1} satisfying

$$\sigma_n^{\alpha} = \langle X, \nu \rangle.$$

If $\alpha > \frac{1}{n+2}$, then M must be a round sphere; if $\alpha = \frac{1}{n+2}$, then M is an ellipsoid.

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Remark 1.3. The results of convergence of σ_n^{α} -curvature flow could implies Theorem 1.2. In case $\alpha = \frac{1}{n}$, Theorem 1.2 was contained in the results of B. Chow in [12]. In case n = 2, Theorem 1.2 was proved by B. Andrews for $\alpha = 1$ in [3], by B. Andrews and X. Chen for $\frac{1}{2} \leq \alpha \leq 1$ in [6]. In case $\alpha = \frac{1}{n+2}$, Theorem 1.2 was proved by B. Andrews in [2]. The more properties of σ_n^{α} -curvature flow were studied by W. J. Firey [14], B. Chow [12], K. Tso [21], B. Andrews [3], P.-F. Guan and L. Ni [17], B. Andrews, P.-F. Guan and L. Ni [7], etc.

From Theorem 1.1 and Theorem 1.2, the following natural question arises:

Question. For any fixed k with $1 \le k \le n-1$, let M be a closed, strictly convex hypersurface in \mathbb{R}^{n+1} satisfying (1.1) with $\alpha \ge \frac{1}{k}$. Can we conclude that M must be a round sphere?

In this paper, we give an affirmative answer to the above question by proving the following result:

Theorem 1.4. For any fixed k with $1 \le k \le n-1$, let M be a closed, strictly convex hypersurface in \mathbb{R}^{n+1} satisfying

$$\sigma_k^{\alpha} = \langle X, \nu \rangle$$

with $\alpha \geq \frac{1}{k}$. Then M must be a round sphere.

Remark 1.5. Theorem 1.1 implies Theorem 1.4 for the case k = 1 and $\alpha = 1$. For $\alpha = \frac{1}{k}$, Theorem 1.4 was contained in the results of B. Chow [12, 13] and B. Andrews [1, 2, 4, 5]. For general k and α , there are some partial results under certain pinching condition of the principal curvatures of hypersurface, see [20], [8] and [15].

In fact, we prove the following two theorems:

Theorem A. For any fixed k with $1 \le k \le n$, let M be a closed, strictly convex hypersurface in \mathbb{R}^{n+1} satisfying

(1.3)
$$\sigma_k^{\alpha} + C = \langle X, \nu \rangle$$

with constants α and C. If either $1 \leq k \leq n-1$, $C \leq 0$, $\alpha \geq \frac{1}{k}$, or, k = n, C < 0, $\alpha \geq \frac{1}{n+2}$, then M must be a round sphere.

Remark 1.6. Choose C = 0, Theorem A reduces to Theorem 1.4. When $k = \alpha = 1$, Theorem A implies the uniqueness of closed λ -hypersurfaces introduced by Cheng-Wei [10].

Let $S_k(\lambda)$ denote the k-th power sum of the principal curvatures $\lambda_1, \dots, \lambda_n$, defined by $S_k(\lambda) = \sum_{i=1}^n \lambda_i^k$.

Theorem B. For any fixed k with $k \ge 1$, let M be a closed, strictly convex hypersurface in \mathbb{R}^{n+1} satisfying

(1.4)
$$S_k^{\alpha} + C = \langle X, \nu \rangle$$

with constants α and C. If $\alpha \geq \frac{1}{k}$ and $C \leq 0$, then M must be a round sphere.

In this paper we first consider the following general equation

(1.5)
$$F + C = \langle X, \nu \rangle,$$

where F = F(h) is a homogeneous smooth symmetric function of the second fundamental form $h = (h_{ij})$ of degree β and C is a constant. We also suppose F > 0

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and $\left(\frac{\partial F}{\partial h_{ij}}\right)$ is positive definite. In the spirit of Choi-Daskaspoulos [11] and Brendle-Choi-Daskaspoulos [9], we consider the quantities

(1.6)
$$Z = F \operatorname{tr} b - \frac{n(\beta - 1)}{2\beta} |X|^2,$$

(1.7)
$$\tilde{W} = F \lambda_{\min}^{-1} - \frac{\beta - 1}{2\beta} |X|^2,$$

where $b = (b^{ij})$ denotes the inverse of the second fundamental form $h = (h_{ij})$ and λ_{\min} is the smallest principal curvature of the hypersurface. In case $F = \sigma_k^{\alpha}$ or $F = S_k^{\alpha}$, by exploring properties of σ_k and S_k intensively, we find that the techniques in Choi-Daskaspoulos [11] and Brendle-Choi-Daskaspoulos [9] can be carried out effectively by using the strong maximum principle of $\mathcal{L} = \frac{\partial F}{\partial h_{ij}} \nabla_i \nabla_j$ for Z and by using the maximum principle for W (see Section 4 for definition of W).

The structure of this paper is as follows. In Section 2, we give some properties of the elementary symmetric functions σ_k and prove our key lemma (Lemma 2.7). In Section 3, we derive some fundamental formulas for the closed hypersurfaces which satisfies self-similar equation (1.5) with the general homogeneous symmetric function F. In Section 4, we do analysis at the maximum point of W. In Section 5 and 6, we present the proofs of Theorem A and Theorem B, respectively.

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2. Some properties of elementary symmetric functions and key lemma

We first collect some basic notations, definitions and properties of elementary symmetric functions, which are needed in our investigation of the σ_k^{α} self-similar solutions.

Let $\lambda = (\lambda_1, \dots, \lambda_n)$ denote the principal curvatures of M. Throughout this paper, we assume that $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$. Denote

$$\sigma_k(\lambda) = \sigma_k(\lambda(A)) = \sum_{1 \le i_1 < i_2 \cdots < i_k \le n} \lambda_{i_1} \lambda_{i_2} \cdots \lambda_{i_k}.$$

For convenience, we set $\sigma_0(\lambda) = 1$ and $\sigma_k(\lambda) = 0$ for k > n or k < 0. Let $\sigma_{k;i}(\lambda)$ denote the symmetric function $\sigma_k(\lambda)$ with $\lambda_i = 0$ and $\sigma_{k;ij}(\lambda)$, with $i \neq j$, denote the symmetric function $\sigma_k(\lambda)$ with $\lambda_i = \lambda_j = 0$. So $\frac{\partial \sigma_k(\lambda)}{\partial \lambda_i} = \sigma_{k-1;i}$, $\frac{\partial^2 \sigma_k(\lambda)}{\partial \lambda_i \partial \lambda_j} = \sigma_{k-2;ij}$. Remark that without causing ambiguity we omit λ in the notations of $\sigma_k(\lambda)$ for simplicity.

Definition 2.1. A hypersurface M is said to be *strictly convex* if $\lambda \in \Gamma_+ = \{\mu \in \mathbb{R}^n | \mu_1 > 0, \mu_2 > 0, \dots, \mu_n > 0\}$ for any point in M.

The following basic properties related to σ_k will be used directly.

Proposition 2.2 (See, for example, [19]). For $0 \le k \le n$ and $1 \le i \le n$, the following equalities hold:

$$\sigma_{k+1} = \sigma_{k+1;i} + \lambda_i \sigma_{k;i},$$

$$\sum_{i=1}^n \lambda_i \sigma_{k;i} = (k+1)\sigma_{k+1},$$

$$\sum_{i=1}^n \sigma_{k;i} = (n-k)\sigma_k,$$

$$\sum_{i=1}^n \lambda_i^2 \sigma_{k;i} = \sigma_1 \sigma_{k+1} - (k+2)\sigma_{k+2}.$$

Lemma 2.3. If $\lambda \in \Gamma_+$ and $i \neq j$, then

$$\frac{\sigma_{k-1;i}(\lambda_i - \lambda_1)^2 - \sigma_{k-1;j}(\lambda_j - \lambda_1)^2}{\lambda_i - \lambda_j} \ge 0.$$

Proof. Since $\sigma_{k-1;i} = \sigma_{k-1;ij} + \lambda_j \sigma_{k-2;ij}$, we have

$$\sigma_{k-1;i}(\lambda_i - \lambda_1)^2 - \sigma_{k-1;j}(\lambda_j - \lambda_1)^2$$

= $\sigma_{k-1;ij}(\lambda_i - \lambda_j)(\lambda_i + \lambda_j - 2\lambda_1) + \sigma_{k-2;ij}(\lambda_i - \lambda_j)(\lambda_i\lambda_j - \lambda_1^2)$
= $(\lambda_i - \lambda_j) \Big(\sigma_{k-1;ij}(\lambda_i + \lambda_j - 2\lambda_1) + \sigma_{k-2;ij}(\lambda_i\lambda_j - \lambda_1^2) \Big).$

Then

$$\frac{\sigma_{k-1;i}(\lambda_i - \lambda_1)^2 - \sigma_{k-1;j}(\lambda_j - \lambda_1)^2}{\lambda_i - \lambda_j}$$

= $\sigma_{k-1;ij}(\lambda_i + \lambda_j - 2\lambda_1) + \sigma_{k-2;ij}(\lambda_i\lambda_j - \lambda_1^2) \ge 0.$

Lemma 2.4. For $\lambda \in \Gamma_k = \{\mu \in \mathbb{R}^n | \sigma_1(\mu) > 0, \cdots, \sigma_k(\mu) > 0\}$, we have

$$\sigma_k \ge \frac{n}{k} \lambda_1 \sigma_{k-1;1}.$$

Proof. By using Proposition 2.2 and $\sigma_{k;i} \leq \sigma_{k;1}$, we have

$$k\sigma_k = \sum_i \lambda_i \sigma_{k-1;i} = \sum_i (\sigma_k - \sigma_{k;i}) \ge n(\sigma_k - \sigma_{k;1}) = n\lambda_1 \sigma_{k-1;1}.$$

We now turn to prove our key lemma of σ_k . First we show two lemmas. Let $D_m^{(k)}(\lambda) = (d_{ij}), i, j = 0, \cdots, m$, denote the following symmetric $(m+1) \times (m+1)$ -matrix

σ_k	$\sigma_{k;1}$	$\sigma_{k;2}$	• • •	$\sigma_{k;m}$)	1
$\sigma_{k;1}$	$\sigma_{k;1}$	$\sigma_{k;12}$	• • •	$\sigma_{k;1m}$	
$\sigma_{k;2}$	$\sigma_{k;21}$	$\sigma_{k;2}$	• • •	$\sigma_{k;2m}$	Ι.
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$\langle \sigma_{k;m}$	$\sigma_{k;m1}$	$\sigma_{k;m2}$		$\sigma_{k;m}$ /	

i.e., $d_{ij} = d_{ji}$ and

$$d_{ij} = \begin{cases} \sigma_k(\lambda), & \text{if } i = j = 0, \\ \sigma_{k;j}(\lambda), & \text{if } i = 0, \ 1 \le j \le m, \\ \sigma_{k;i}(\lambda), & \text{if } 1 \le i = j \le m, \\ \sigma_{k;ij}(\lambda), & \text{if } 1 \le i < j \le m. \end{cases}$$

Lemma 2.5. If $\lambda \in \Gamma_+$ and $n \geq 2$, then $D_n^{(k)}(\lambda)$ is semi-positive definite for $1 \leq k \leq n$.

Proof. First, since $\sigma_{n;i} = \sigma_{n;pq} = 0$ for $1 \leq i, p, q \leq n$, it is clear that $D_n^{(n)}$ is semi-positive definite.

For $1 \le k \le n-1$, the statement follows by induction on n. In fact, for n = 2, the semi-positive-definiteness is proved by directly computation. Now, assum that the statement is true for n-1. For $\lambda = (\lambda_1, ..., \lambda_n)$, the assumption implies the following matrices are semi-positive definite

$$D_{n-1;n}^{(k)}(\lambda) = \begin{pmatrix} \sigma_{k;n} & \sigma_{k;1n} & \sigma_{k;2n} & \cdots & \sigma_{k;n-1,n} \\ \sigma_{k;1n} & \sigma_{k;1n} & \sigma_{k;12} & \cdots & \sigma_{k;1,n-1,n} \\ \sigma_{k;2n} & \sigma_{k;21n} & \sigma_{k;2n} & \cdots & \sigma_{k;2,n-1,n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \sigma_{k;n-1,n} & \sigma_{k;n-1,1n} & \sigma_{k;n-1,2n} & \cdots & \sigma_{k;n-1,n} \end{pmatrix}$$

for $1 \le k \le n-1$. And, using

$$\sigma_k = \sigma_{k;n} + \lambda_n \sigma_{k-1;n}, \quad \sigma_{k,i} = \sigma_{k;in} + \lambda_n \sigma_{k-1;in} \ (1 \le i \le n-1),$$

we obtain

$$D_{n}^{(k)}(\lambda) = \lambda_{n} \begin{pmatrix} D_{n-1;n}^{(k-1)} & 0\\ 0 & 0 \end{pmatrix} + \begin{pmatrix} D_{n-1;n}^{(k)} & \eta\\ \eta^{T} & \sigma_{k;n} \end{pmatrix}$$

where $\eta^T = (\sigma_{k;n}, \sigma_{k;1n}, \sigma_{k;2n}, \cdots, \sigma_{k;n-1,n})$. For

$$\begin{pmatrix} D_{n-1;n}^{(k)} & \eta \\ \eta^T & \sigma_{k;n} \end{pmatrix} = \begin{pmatrix} \sigma_{k;n} & \sigma_{k;1n} & \sigma_{k;2n} & \cdots & \sigma_{k;n-1,n} & \sigma_{k;n} \\ \sigma_{k;1n} & \sigma_{k;1n} & \sigma_{k;12n} & \cdots & \sigma_{k;1,n-1,n}, & \sigma_{k;1n} \\ \sigma_{k;2n} & \sigma_{k;21n} & \sigma_{k;2n} & \cdots & \sigma_{k;2,n-1,n} & \sigma_{k;2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \sigma_{k;n-1,n} & \sigma_{k;n-1,1,n} & \sigma_{k;n-1,2,n} & \cdots & \sigma_{k;n-1,n} & \sigma_{k;n-1,n} \\ \sigma_{k;n} & \sigma_{k;n1} & \sigma_{k;n2} & \cdots & \sigma_{k;n,n-1} & \sigma_{k;n} \end{pmatrix}$$

by subtracting the first row from the last row and the first column from the last column, we find that $\begin{pmatrix} D_{n-1;n}^{(k)} & \eta \\ \eta^T & \sigma_{k;n} \end{pmatrix}$ is congruent to $\begin{pmatrix} D_{n-1;n}^{(k)} & 0 \\ 0 & 0 \end{pmatrix}$ which is semipositive definite. So $D_n^{(k)}(\lambda)$ is semi-positive definite. Thus, the proof is completed.

For $\lambda = (\lambda_1, ..., \lambda_n) \in \Gamma_+$, let $A^{(k)}(\lambda) = (a_{ij})_{n \times n}$ denote the following matrix

$$\begin{pmatrix} \frac{1}{\lambda_{1}}\sigma_{k-1;1} & \sigma_{k-2;12} & \sigma_{k-2;13} & \cdots & \sigma_{k-2;1n} \\ \sigma_{k-2;21} & \frac{1}{\lambda_{2}}\sigma_{k-1;2} & \sigma_{k-2;23} & \cdots & \sigma_{k-2;2n} \\ \sigma_{k-2;31} & \sigma_{k-2;32} & \frac{1}{\lambda_{3}}\sigma_{k-1;3} & \cdots & \sigma_{k-2;3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \sigma_{k-2;n1} & \sigma_{k-2;n2} & \sigma_{k-2;n3} & \cdots & \frac{1}{\lambda_{n}}\sigma_{k-1;n} \end{pmatrix}$$

i.e.,

$$a_{ij} = \begin{cases} \frac{1}{\lambda_i} \sigma_{k-1;i}(\lambda), & \text{for } i = j, \\ \sigma_{k-2;ij}(\lambda), & \text{for } i \neq j. \end{cases}$$

Lemma 2.6. Let $\xi^T = (\sigma_{k-1;1}, \sigma_{k-1;2}, ..., \sigma_{k-1;n})$. Then the matrix $\sigma_k A^{(k)} - \xi \xi^T$ is semi-positive definite.

Proof. Denote $\sigma_k A^{(k)} - \xi \xi^T = (w_{ij})_{n \times n}$. Thus

$$w_{ij} = \begin{cases} \frac{\sigma_{k-1;i}}{\lambda_i} \sigma_{k;i}, & \text{for } i = j, \\ \frac{1}{\lambda_i \lambda_j} (\sigma_k \sigma_{k;ij} - \sigma_{k;i} \sigma_{k;j}), & \text{for } i \neq j. \end{cases}$$

We divide the proof in three steps.

Step 1. Since the semi-positive-definiteness is preserved under congruent transformation, we multiply λ_i to the *i*-th row and the *i*-th column of $\sigma_k A^{(k)} - \xi \xi^T$ for $1 \leq i \leq n$. And, let $\tilde{A}^{(k)} = (\tilde{a}_{ij})_{n \times n}$ denote the new matrix which is defined by

$$\tilde{a}_{ij} = \begin{cases} \sigma_{k;i}(\sigma_k - \sigma_{k;i}), & \text{for } i = j, \\ \sigma_k \sigma_{k;ij} - \sigma_{k;i} \sigma_{k;j}, & \text{for } i \neq j. \end{cases}$$

We will discuss $\tilde{A}^{(k)}$ instead of $\sigma_k A^{(k)} - \xi \xi^T$ in the following. Step 2. $\tilde{A}^{(k)}$ is semi-positive definite if and only if its principal minors are all non-negative. Let $\tilde{A}_m^{(k)}$ denote the upper-left $m \times m$ sub-matrix of $\tilde{A}^{(k)}$. For the symmetry of the elemental functions, it suffices to show det $\tilde{A}_m^{(k)} \ge 0$. Step 3. det $\tilde{A}_m^{(k)}$ can be calculated as follows.

$$\det \tilde{A}_{m}^{(k)} = \det \begin{pmatrix} 1 & \sigma_{k;1} & \sigma_{k;2} & \cdots & \sigma_{k;m} \\ 0 & \sigma_{k}\sigma_{k;1} - \sigma_{k;1}^{2} & \sigma_{k}\sigma_{k;12} - \sigma_{k;1}\sigma_{k;2} & \cdots & \sigma_{k}\sigma_{k;1m} - \sigma_{k;1}\sigma_{k;m} \\ 0 & \sigma_{k}\sigma_{k;12} - \sigma_{k;1}\sigma_{k;2} & \sigma_{k}\sigma_{k;2} - \sigma_{k;2}^{2} & \cdots & \sigma_{k}\sigma_{k;2m} - \sigma_{k;2}\sigma_{k;m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & \sigma_{k}\sigma_{k;m1} - \sigma_{k;m}\sigma_{k;1} & \sigma_{k}\sigma_{k;m2} - \sigma_{k;m}\sigma_{k;2} & \cdots & \sigma_{k}\sigma_{k;m} - \sigma_{k;m}^{2} \\ \sigma_{k;1} & \sigma_{k}\sigma_{k;12} & \sigma_{k}\sigma_{k;12} & \cdots & \sigma_{k}\sigma_{k;1m} \\ \sigma_{k;2} & \sigma_{k}\sigma_{k;12} & \sigma_{k}\sigma_{k;2m} & \cdots & \sigma_{k}\sigma_{k;m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \sigma_{k;m} & \sigma_{k}\sigma_{k;m1} & \sigma_{k}\sigma_{k;m2} & \cdots & \sigma_{k}\sigma_{k;m} \end{pmatrix}$$

$$= \sigma_{k}^{-2} \det \begin{pmatrix} \sigma_{k}^{2} & \sigma_{k}\sigma_{k;1} & \sigma_{k}\sigma_{k;2m} & \cdots & \sigma_{k}\sigma_{k;m} \\ \sigma_{k}\sigma_{k;2} & \sigma_{k}\sigma_{k;12} & \sigma_{k}\sigma_{k;2m} & \cdots & \sigma_{k}\sigma_{k;m} \\ \sigma_{k}\sigma_{k;2} & \sigma_{k}\sigma_{k;12} & \sigma_{k}\sigma_{k;2m} & \cdots & \sigma_{k}\sigma_{k;m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \sigma_{k}\sigma_{k;m} & \sigma_{k}\sigma_{k;m1} & \sigma_{k}\sigma_{k;m2} & \cdots & \sigma_{k}\sigma_{k;m} \end{pmatrix}$$

$$= \sigma_{k}^{m-1} \det D_{m}^{(k)}.$$

By Lemma 2.5, we know det $D_m^{(k)} \ge 0$. So, det $\tilde{A}_m^{(k)} \ge 0$ which implies $\sigma_k A^{(k)} - \xi \xi^T$ is semi-positive definite.

With the help of the proceeding two lemmas, we finally obtain our key lemma.

Lemma 2.7. For $y = (y_1, y_2, ..., y_n) \in \mathbb{R}^n$, the following inequality holds

$$\sum_{i=1}^n \frac{\sigma_{k-1;i}}{\lambda_i \sigma_k} y_i^2 + \sum_{i \neq j} \frac{\sigma_{k-2;ij}}{\sigma_k} y_i y_j \ge (\sum_{i=1}^n \frac{\sigma_{k-1;i}}{\sigma_k} y_i)^2.$$

Proof. By Lemma 2.6, we know

$$y^T \left(\frac{1}{\sigma_k} A^{(k)} - \frac{1}{\sigma_k^2} \xi \xi^T\right) y \ge 0.$$

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3. Fundamental formulas of self-similar solution with general ${\cal F}$

Let $X : M^n \to \mathbb{R}^{n+1}$ be a closed convex hypersurface. Suppose that e_1, e_2, \cdots, e_n is an orthonormal frame on M. Let $h = (h_{ij})$ be the second fundamental form on M with respect to this given frame. And the principal curvatures are the eigenvalues of the second fundamental form h.

Let us first consider the following general equation

$$F(h) + C = \langle X, \nu \rangle,$$

where F is a homogeneous symmetric function of $h = (h_{ij})$ of degree β , C is a constant and ν is the outward normal vector field. And, let \mathcal{L} denote the operator $\mathcal{L} = \frac{\partial F}{\partial h_{ij}} \nabla_i \nabla_j$. We also suppose F > 0 and $(\frac{\partial F}{\partial h_{ij}})$ is positive definite. Inspired by [20], [11] and [9], we have the following proposition. The summation convention is used unless otherwise stated.

Proposition 3.1. Given a smooth function $F: M \to \mathbb{R}^{n+1}$ described as above, the following equations hold:

(1)
$$\mathcal{L}F = \langle X, \nabla F \rangle + \beta F - \frac{\partial F}{\partial h_{ij}} h_{jl} h_{li} (F + C),$$

(2)
$$\mathcal{L}h_{kl} = h_{klm} \langle X, e_m \rangle + h_{kl} - Ch_{km}h_{lm} - \frac{\partial^2 F}{\partial h_{ij}\partial h_{st}} h_{ijk}h_{stl} - \frac{\partial F}{\partial h_{ij}} h_{mj}h_{mi}h_{kl} + (\beta - 1)Fh_{km}h_{ml},$$

(3)
$$\mathcal{L}b^{kl} = \langle X, \nabla b^{kl} \rangle - b^{kl} + C\delta_{kl} + b^{kp}b^{ql}\frac{\partial^2 F}{\partial h_{ij}\partial h_{st}}h_{ijp}h_{stq}$$

$$+ b^{kl} \frac{\partial F}{\partial h_{ij}} h_{mj} h_{mi} - (\beta - 1) F \delta_{kl} + 2b^{ks} b^{pt} b^{lq} \frac{\partial F}{\partial h_{ij}} h_{sti} h_{pqj}$$

$$\begin{aligned} (4) \qquad \mathcal{L}(F\mathrm{tr}b) &= \langle X, \nabla(F\mathrm{tr}b) \rangle + (\beta - 1)F\mathrm{tr}b - n(\beta - 1)F^{2} \\ &+ C(nF - \mathrm{tr}b\frac{\partial F}{\partial h_{ij}}h_{jl}h_{li}) + 2\frac{\partial F}{\partial h_{ij}}\nabla_{i}F\nabla_{j}\mathrm{tr}b \\ &+ Fb^{kp}b^{qk}\frac{\partial^{2}F}{\partial h_{ij}\partial h_{st}}h_{ijp}h_{stq} + 2Fb^{ks}b^{pt}b^{kq}\frac{\partial F}{\partial h_{ij}}h_{sti}h_{pqj} \end{aligned}$$

$$(5) \qquad \mathcal{L}\frac{|X|^{2}}{2} = \sum_{i}\frac{\partial F}{\partial h_{ii}} - \beta F(F+C). \end{aligned}$$

Proof. (1) Differentiating (1.5) gives

(3.1)
$$\nabla_j F = h_{jl} \langle X, e_l \rangle$$

and

$$\nabla_i \nabla_j F = h_{jli} \langle X, e_l \rangle + h_{ij} - h_{jl} h_{il} \langle X, \nu \rangle$$
$$= h_{jli} \langle X, e_l \rangle + h_{ij} - h_{jl} h_{il} (F + C).$$

 $=h_{jli}\langle X,e_l\rangle+h_{ij}-h_{ji}$ Then, by $\frac{\partial F}{\partial h_{ij}}h_{ij}=\beta F$, we obtain

$$\mathcal{L}F = \nabla_l F \langle X, e_l \rangle + \beta F - \frac{\partial F}{\partial h_{ij}} h_{jl} h_{li} (F + C).$$

(2) By Codazzi equation and Ricci identity, we obtain

$$h_{klji} = h_{kjli} = h_{kjil} + h_{mj}R_{mkli} + h_{km}R_{mjli}.$$

Then, using Gauss equation we have

$$\mathcal{L}h_{kl} = \frac{\partial F}{\partial h_{ij}} (h_{kjil} + h_{mj}R_{mkli} + h_{km}R_{mjli})$$

$$= \nabla_l (\frac{\partial F}{\partial h_{ij}}h_{ijk}) - \frac{\partial^2 F}{\partial h_{ij}\partial h_{st}}h_{ijk}h_{stl} + \frac{\partial F}{\partial h_{ij}}h_{mj}(h_{ml}h_{ki} - h_{mi}h_{kl})$$

$$+ \frac{\partial F}{\partial h_{ij}}h_{km}(h_{ml}h_{ij} - h_{mi}h_{jl})$$

$$= \nabla_l \nabla_k F - \frac{\partial^2 F}{\partial h_{ij}\partial h_{st}}h_{ijk}h_{stl} - \frac{\partial F}{\partial h_{ij}}h_{mj}h_{mi}h_{kl} + \frac{\partial F}{\partial h_{ij}}h_{km}h_{ml}h_{ij}$$

$$= h_{klm}\langle X, e_m \rangle + h_{kl} - h_{km}h_{lm}(F + C) - \frac{\partial^2 F}{\partial h_{ij}\partial h_{st}}h_{ijk}h_{stl}$$

$$- \frac{\partial F}{\partial h_{ij}}h_{mj}h_{mi}h_{kl} + \beta Fh_{km}h_{ml}.$$

(3) Since $h_{km}b^{ml} = \delta_{kl}$, we have

(3.2)
$$\nabla_j b^{kl} = -b^{kp} b^{lq} \nabla_j h_{pq}.$$

And,

$$\begin{split} \nabla_i \nabla_j b^{kl} &= -\nabla_i (b^{kp} b^{lq} \nabla_j h_{pq}) \\ &= -b^{kp} b^{ql} \nabla_i \nabla_j h_{pq} + b^{ks} b^{pt} b^{lq} \nabla_i h_{st} \nabla_j h_{pq} + b^{kp} b^{ls} b^{qt} \nabla_i h_{st} \nabla_j h_{pq} \\ &= -b^{kp} b^{ql} \nabla_i \nabla_j h_{pq} + 2b^{ks} b^{pt} b^{lq} \nabla_i h_{st} \nabla_j h_{pq}. \end{split}$$

Then, we obtain

$$\begin{split} \mathcal{L}b^{kl} &= -b^{kp}b^{ql}\frac{\partial F}{\partial h_{ij}}\nabla_i\nabla_j h_{pq} + 2b^{ks}b^{pt}b^{lq}\frac{\partial F}{\partial h_{ij}}\nabla_i h_{st}\nabla_j h_{pq} \\ &- b^{kp}b^{ql}\frac{\partial F}{\partial h_{ij}}h_{pm}h_{mq}h_{ij} + 2b^{ks}b^{pt}b^{lq}\frac{\partial F}{\partial h_{ij}}h_{sti}h_{pqj} \\ &= \langle X, \nabla b^{kl} \rangle - b^{kl} + (F+C)\delta_{kl} + b^{kp}b^{ql}\frac{\partial^2 F}{\partial h_{ij}\partial h_{st}}h_{ijp}h_{stq} \\ &+ b^{kl}\frac{\partial F}{\partial h_{ij}}h_{mj}h_{mi} - \beta F\delta_{kl} + 2b^{ks}b^{pt}b^{lq}\frac{\partial F}{\partial h_{ij}}h_{sti}h_{pqj}. \end{split}$$

(4) From (3), we have

$$\begin{split} \frac{\partial F}{\partial h_{ij}} \nabla_i \nabla_j \mathrm{tr}b &= \langle X, \nabla \mathrm{tr}b \rangle - \mathrm{tr}b + n(F+C) + b^{kp} b^{qk} \frac{\partial^2 F}{\partial h_{ij} \partial h_{st}} h_{ijp} h_{stq} \\ &+ \mathrm{tr}b \frac{\partial F}{\partial h_{ij}} h_{mj} h_{mi} - n\beta F + 2 b^{ks} b^{pt} b^{kq} \frac{\partial F}{\partial h_{ij}} h_{sti} h_{pqj}. \end{split}$$

Furthermore,

$$\begin{split} \mathcal{L}(F\mathrm{tr}b) &= 2\frac{\partial F}{\partial h_{ij}} \nabla_i F \nabla_j \mathrm{tr}b + \mathrm{tr}b \frac{\partial F}{\partial h_{ij}} \nabla_i \nabla_j F + F \frac{\partial F}{\partial h_{ij}} \nabla_i \nabla_j \mathrm{tr}b \\ &= 2\frac{\partial F}{\partial h_{ij}} \nabla_i F \nabla_j \mathrm{tr}b + \mathrm{tr}b \langle X, \nabla F \rangle + \mathrm{tr}b \frac{\partial F}{\partial h_{ij}} h_{ij} \\ &- \mathrm{tr}b \frac{\partial F}{\partial h_{ij}} h_{jl} h_{li} (F + C) + F \langle X, \nabla \mathrm{tr}b \rangle - F \mathrm{tr}b + nF(F + C) \\ &+ F b^{kp} b^{qk} \frac{\partial^2 F}{\partial h_{ij} \partial h_{st}} h_{ijp} h_{stq} + F \mathrm{tr}B \frac{\partial F}{\partial h_{ij}} h_{mj} h_{mi} - nF \frac{\partial F}{\partial h_{ij}} h_{ij} \\ &+ 2F b^{ks} b^{pt} b^{kq} \frac{\partial F}{\partial h_{ij}} h_{sti} h_{pqj} \\ &= \langle X, \nabla (F \mathrm{tr}b) \rangle + (\beta - 1)F \mathrm{tr}B - n(\beta - 1)F^2 + C(nF - \mathrm{tr}B \frac{\partial F}{\partial h_{ij}} h_{jl} h_{li}) \\ &+ 2\frac{\partial F}{\partial h_{ij}} \nabla_i F \nabla_j \mathrm{tr}b + F b^{kp} b^{qk} \frac{\partial^2 F}{\partial h_{ij} \partial h_{st}} h_{ijp} h_{stq} + 2F b^{ks} b^{pt} b^{kq} \frac{\partial F}{\partial h_{ij}} h_{sti} h_{pqj}. \end{split}$$

(5) By direct computation and (1.5), we have

$$\mathcal{L}\frac{|X|^2}{2} = \frac{\partial F}{\partial h_{ij}} \nabla_i (\langle X, e_j \rangle)$$
$$= \sum_i \frac{\partial F}{\partial h_{ii}} - (F+C) \frac{\partial F}{\partial h_{ij}} h_{ij}.$$

To finish this section, we list the following well-known result (See for example [1] and [16]).

Lemma 3.2. If $W = (w_{ij})$ is a symmetric real matrix and $\lambda_m = \lambda_m(W)$ is one of its eigenvalues $(m = 1, \dots, n)$. If $F = F(W) = F(\lambda(W))$, then for any real symmetric matrix $B = (b_{ij})$, we have the following formulas:

(i)
$$\frac{\partial F}{\partial w_{ij}} b_{ij} = \frac{\partial F}{\partial \lambda_p} b_{pp},$$

(ii) $\frac{\partial^2 F}{\partial w_{ij} \partial w_{st}} b_{ij} b_{st} = \frac{\partial^2 F}{\partial \lambda_p \partial \lambda_q} b_{pp} b_{qq} + 2 \sum_{p < q} \frac{\frac{\partial F}{\partial \lambda_p} - \frac{\partial F}{\partial \lambda_q}}{\lambda_p - \lambda_q} b_{pq}^2.$

Remark 3.3. In the above lemma, $\frac{\frac{\partial F}{\partial \lambda_p} - \frac{\partial F}{\partial \lambda_q}}{\lambda_p - \lambda_q}$ is interpreted as a limit if $\lambda_p = \lambda_q$.

4. Analysis at the maximum points of ${\boldsymbol W}$

In the recent paper [9], S. Brendle, K. Choi and P. Daskalopoulos proved the following powerful lemma.

Lemma 4.1 ([9]). Let μ denote the multiplicity of λ_1 at a point x_0 , i.e., $\lambda_1(x_0) = \cdots = \lambda_{\mu}(x_0) < \lambda_{\mu+1}(x_0)$. Suppose that φ is a smooth function such that $\varphi \leq \lambda_1$ everywhere and $\varphi(x_0) = \lambda_1(x_0)$. Then, at x_0 , we have i) $h_{kli} = \nabla_i \varphi \delta_{kl}$ for $1 \leq k, l \leq \mu$. ii) $\nabla_i \nabla_i \varphi \leq h_{11ii} - 2 \sum_{l>\mu} (\lambda_l - \lambda_1)^{-1} h_{1li}^2$.

Let $\tilde{W} = \frac{F}{\lambda_1} - \frac{\beta - 1}{2\beta} |X|^2$ and let x_0 be an arbitrary point where \tilde{W} attains its maximum. In fact, we can choose a smooth function φ such that $W = \frac{F}{\varphi} - \frac{\beta - 1}{2\beta} |X|^2$ attains its maximum at x_0 . Thus, $\tilde{W}_{\text{max}} = W_{\text{max}}$. Now, we consider W at x_0 .

Lemma 4.2. At x_0 , W satisfies the following inequality

$$\begin{split} \mathcal{L}W &\geq \langle X, \nabla(\frac{F}{\varphi}) \rangle + 2 \frac{\partial F}{\partial h_{ij}} \nabla_i F \nabla_j \frac{1}{\varphi} + 2F \lambda_1^{-3} \frac{\partial F}{\partial \lambda_i} h_{11i}^2 \\ &+ F \lambda_1^{-2} \frac{\partial^2 F}{\partial h_{ij} \partial h_{st}} h_{ij1} h_{st1} + 2F \lambda_1^{-2} \frac{\partial F}{\partial \lambda_i} \sum_{l > \mu} (\lambda_l - \lambda_1)^{-1} h_{1li}^2 \\ &+ \frac{\beta - 1}{\beta} \frac{\partial F}{\partial \lambda_i} (\frac{\lambda_i}{\lambda_1} - 1) - C \frac{\partial F}{\partial \lambda_i} \lambda_i (\frac{\lambda_i}{\lambda_1} - 1). \end{split}$$

Proof. At x_0 , it follows from Lemma 4.1 and Proposition 3.1 that

$$\mathcal{L}\varphi \leq \mathcal{L}h_{11} - 2\frac{\partial F}{\partial \lambda_i} \sum_{l>\mu} (\lambda_l - \lambda_1)^{-1} h_{1li}^2$$

= $h_{11m} \langle X, e_m \rangle + \lambda_1 - \lambda_1^2 C - \lambda_1 \frac{\partial F}{\partial h_{ij}} h_{mj} h_{mi} + \lambda_1^2 (\beta - 1) F$
 $- \frac{\partial^2 F}{\partial h_{ij} \partial h_{st}} h_{ij1} h_{st1} - 2\frac{\partial F}{\partial \lambda_i} \sum_{l>\mu} (\lambda_l - \lambda_1)^{-1} h_{1li}^2.$

Furthermore, we have

$$\begin{split} \mathcal{L}\frac{F}{\varphi} &= 2\frac{\partial F}{\partial h_{ij}} \nabla_i F \nabla_j \frac{1}{\varphi} + \frac{1}{\varphi} \mathcal{L}F + F \mathcal{L}\frac{1}{\varphi} \\ &\geq 2\frac{\partial F}{\partial h_{ij}} \nabla_i F \nabla_j \frac{1}{\varphi} + \lambda_1^{-1} \nabla_l F \langle X, e_l \rangle + \lambda_1^{-1} \frac{\partial F}{\partial h_{ij}} h_{ij} - \lambda_1^{-1} \frac{\partial F}{\partial h_{ij}} h_{jl} h_{li} (F+C) \\ &+ 2F \lambda_1^{-3} \frac{\partial F}{\partial \lambda_i} h_{11i}^2 + F \nabla_m \frac{1}{\varphi} \langle X, e_m \rangle - F \lambda_1^{-1} + (1-\beta)F^2 + FC + F \lambda_1^{-1} \frac{\partial F}{\partial h_{ij}} h_{mj} h_{mi} \\ &+ F \lambda_1^{-2} \frac{\partial^2 F}{\partial h_{ij} \partial h_{st}} h_{ij1} h_{st1} + 2F \lambda_1^{-2} \frac{\partial F}{\partial \lambda_i} \sum_{l > \mu} (\lambda_l - \lambda_1)^{-1} h_{1li}^2 \\ &= 2\frac{\partial F}{\partial h_{ij}} \nabla_i F \nabla_j \frac{1}{\varphi} + 2F \lambda_1^{-3} \frac{\partial F}{\partial \lambda_i} h_{11i}^2 + \nabla_m \frac{F}{\varphi} \langle X, e_m \rangle \\ &+ (\beta - 1)F \lambda_1^{-1} + (1-\beta)F^2 + C(F - \lambda_1^{-1} \frac{\partial F}{\partial h_{ij}} h_{jl} h_{li}) \\ &+ F \lambda_1^{-2} \frac{\partial^2 F}{\partial h_{ij} \partial h_{st}} h_{ij1} h_{st1} + 2F \lambda_1^{-2} \frac{\partial F}{\partial \lambda_i} \sum_{l > \mu} (\lambda_l - \lambda_1)^{-1} h_{1li}^2. \end{split}$$

According to Proposition 3.1 and the homogeneity of F, we have

$$\begin{aligned} &-\frac{\beta-1}{\beta}\mathcal{L}\frac{|X|^2}{2} + (\beta-1)F\lambda_1^{-1} + (1-\beta)F^2 + C(F-\lambda_1^{-1}\frac{\partial F}{\partial h_{ij}}h_{jl}h_{li}) \\ &= \frac{\beta-1}{\beta}\frac{\partial F}{\partial \lambda_i}(\frac{\lambda_i}{\lambda_1}-1) - C\frac{\partial F}{\partial \lambda_i}\lambda_i(\frac{\lambda_i}{\lambda_1}-1), \end{aligned}$$

thus the proof is completed.

Let

$$J_1 = \frac{\beta - 1}{\beta} \frac{\partial F}{\partial \lambda_i} (\frac{\lambda_i}{\lambda_1} - 1) - C \frac{\partial F}{\partial \lambda_i} \lambda_i (\frac{\lambda_i}{\lambda_1} - 1).$$

Lemma 4.3. If $C \leq 0$ and $\beta > 1$, then $J_1 \geq 0$. And $J_1 = 0$ if and only if $\lambda_1 = \cdots = \lambda_n$.

Proof. The proof directly follows from that $\frac{\partial F}{\partial \lambda_i} > 0$ and $\frac{\lambda_i}{\lambda_1} \ge 1$.

Lemma 4.4. At x_0 , we have the following equalities

(1) $\langle X, \nabla(\frac{F}{\varphi}) \rangle = \frac{\beta - 1}{\beta} \sum_{i} \lambda_i^{-2} (\nabla_i F)^2,$

(2)
$$F\lambda_1^{-2}h_{11j} = (\lambda_1^{-1} - \frac{\beta - 1}{\beta}\lambda_j^{-1})\nabla_j F, \text{ for } 1 \le j \le n,$$

(3)
$$\nabla_m F = 0, \text{ for } 2 \le m \le \mu.$$

Proof. (1) Using $\nabla W = 0$ and (3.1), we have

$$\begin{split} \langle X, \nabla(\frac{F}{\varphi}) \rangle &= \langle X, \nabla W \rangle + \frac{\beta - 1}{\beta} \sum_{m} \langle X, e_{m} \rangle^{2} \\ &= \frac{\beta - 1}{\beta} \sum_{i} \lambda_{i}^{-2} (\nabla_{i} F)^{2}. \end{split}$$

(2) Using $\nabla_j W = 0$, Lemma 4.1 and (3.1), we have

$$0 = F\nabla_j \frac{1}{\varphi} + \frac{1}{\varphi} \nabla_j F - \frac{\beta - 1}{\beta} \lambda_j^{-1} \nabla_j F$$
$$= -F\lambda_1^{-2} h_{11j} + (\lambda_1^{-1} - \frac{\beta - 1}{\beta} \lambda_j^{-1}) \nabla_j F$$

(3) By Lemma 4.1, we have $h_{11m} = 0$ if $2 \le m \le \mu$. Then, (2) leads to (3). \Box

Lemma 4.5.

$$\begin{split} &\frac{\partial^2 F}{\partial h_{ij}\partial h_{st}}h_{ij1}h_{st1} + 2\frac{\partial F}{\partial \lambda_i}\sum_{l>\mu}(\lambda_l - \lambda_1)^{-1}h_{1li}^2\\ &= \frac{\partial^2 F}{\partial \lambda_i \partial \lambda_j}h_{ii1}h_{jj1} + 2\sum_{i>\mu}\frac{\partial F}{\partial \lambda_i}(\lambda_i - \lambda_1)^{-1}h_{11i}^2 + 2\sum_{i>\mu}\frac{\partial F}{\partial \lambda_i}(\lambda_i - \lambda_1)^{-1}h_{1ii}^2\\ &+ 2\sum_{i>j>\mu}\frac{\frac{\partial F}{\partial \lambda_i}(\lambda_i - \lambda_1)^2 - \frac{\partial F}{\partial \lambda_j}(\lambda_j - \lambda_1)^2}{(\lambda_i - \lambda_1)(\lambda_j - \lambda_1)(\lambda_i - \lambda_j)}h_{ij1}^2. \end{split}$$

Proof. Due to

$$\begin{split} \frac{\partial^2 F}{\partial h_{ij} \partial h_{st}} h_{ij1} h_{st1} &= \left(\frac{\partial^2 F}{\partial \lambda_i \partial \lambda_j} h_{ii1} h_{jj1} + 2 \sum_{i>j} (\lambda_i - \lambda_j)^{-1} (\frac{\partial F}{\partial \lambda_i} - \frac{\partial F}{\partial \lambda_j}) h_{ij1}^2 \right) \\ &= \frac{\partial^2 F}{\partial \lambda_i \partial \lambda_j} h_{ii1} h_{jj1} + 2 \sum_{i>\mu} (\lambda_i - \lambda_1)^{-1} (\frac{\partial F}{\partial \lambda_i} - \frac{\partial F}{\partial \lambda_1}) h_{11i}^2 \\ &+ 2 \sum_{i>j>\mu} (\lambda_i - \lambda_j)^{-1} (\frac{\partial F}{\partial \lambda_i} - \frac{\partial F}{\partial \lambda_j}) h_{ij1}^2 \end{split}$$

and

$$2\frac{\partial F}{\partial \lambda_i} \sum_{l>\mu} (\lambda_l - \lambda_1)^{-1} h_{1li}^2 = 2\frac{\partial F}{\partial \lambda_1} \sum_{l>\mu} (\lambda_l - \lambda_1)^{-1} h_{11l}^2 + 2\sum_{i>\mu} \frac{\partial F}{\partial \lambda_i} (\lambda_i - \lambda_1)^{-1} h_{1ii}^2 + 2\sum_{i>l>\mu} \frac{\partial F}{\partial \lambda_i} (\lambda_l - \lambda_1)^{-1} h_{1li}^2 + 2\sum_{l>i>\mu} \frac{\partial F}{\partial \lambda_i} (\lambda_l - \lambda_1)^{-1} h_{1li}^2,$$

the lemma follows by adding the above two equations.

Lemma 4.6. For $\beta \geq 1$, at x_0 , W satisfies the following inequality

$$\mathcal{L}W \ge J_1 + J_2 + J_3,$$

where

$$J_{1} = \frac{\beta - 1}{\beta} \frac{\partial F}{\partial \lambda_{i}} (\frac{\lambda_{i}}{\lambda_{1}} - 1) - C \frac{\partial F}{\partial \lambda_{i}} \lambda_{i} (\frac{\lambda_{i}}{\lambda_{1}} - 1),$$

$$J_{2} = 2F \lambda_{1}^{-2} \sum_{i > j > \mu} \frac{\frac{\partial F}{\partial \lambda_{i}} (\lambda_{i} - \lambda_{1})^{2} - \frac{\partial F}{\partial \lambda_{j}} (\lambda_{j} - \lambda_{1})^{2}}{(\lambda_{i} - \lambda_{1})(\lambda_{j} - \lambda_{1})(\lambda_{i} - \lambda_{j})} h_{ij1}^{2}$$

and

$$J_{3} = \frac{\beta - 1}{\beta} \lambda_{1}^{-1} \Big(\lambda_{1}^{-1} - \frac{2}{\beta} F^{-1} \frac{\partial F}{\partial \lambda_{1}} \Big) (\nabla_{1} F)^{2} + 2F \lambda_{1}^{-2} \frac{\partial F}{\partial \lambda_{i}} \sum_{i > \mu} (\lambda_{i} - \lambda_{1})^{-1} h_{1ii}^{2} + F \lambda_{1}^{-2} \frac{\partial^{2} F}{\partial \lambda_{i} \partial \lambda_{j}} h_{ii1} h_{jj1}.$$

Proof. By Lemma 4.4, we have

$$\begin{split} \langle X, \nabla(\frac{F}{\varphi}) \rangle &+ 2 \frac{\partial F}{\partial h_{ij}} \nabla_i F \nabla_j \frac{1}{\varphi} + 2F \lambda_1^{-3} \frac{\partial F}{\partial \lambda_i} h_{11i}^2 \\ &= \frac{\beta - 1}{\beta} \sum_i \lambda_i^{-2} (\nabla_i F)^2 - 2F^{-1} \frac{\partial F}{\partial \lambda_i} (\lambda_1^{-1} - \frac{\beta - 1}{\beta} \lambda_i^{-1}) (\nabla_i F)^2 + 2F \lambda_1^{-3} \frac{\partial F}{\partial \lambda_i} h_{11i}^2 \\ &= \frac{\beta - 1}{\beta} \lambda_1^{-1} \Big(\lambda_1^{-1} - \frac{2}{\beta} F^{-1} \frac{\partial F}{\partial \lambda_1} \Big) (\nabla_1 F)^2 + \sum_{i > \mu} \Big(\frac{\beta - 1}{\beta} \lambda_i^{-2} \\ &- \frac{2(\beta - 1)}{\beta} F^{-1} \frac{\partial F}{\partial \lambda_i} \lambda_1 \lambda_i^{-1} (\lambda_1^{-1} - \frac{\beta - 1}{\beta} \lambda_i^{-1}) \Big) (\nabla_i F)^2. \end{split}$$

Furthermore, by Lemma 4.4 and 4.5, we have

$$\begin{aligned} \mathcal{L}W &\geq \frac{\beta-1}{\beta}\lambda_1^{-1} \Big(\lambda_1^{-1} - \frac{2}{\beta}F^{-1}\frac{\partial F}{\partial\lambda_1}\Big)(\nabla_1 F)^2 \\ &+ \sum_{i>\mu} \Big\{\frac{\beta-1}{\beta}\lambda_i^{-2} + \frac{2}{\beta}F^{-1}\frac{\partial F}{\partial\lambda_i}\Big(\lambda_i^{-1} + \frac{1}{\beta}\lambda_1\lambda_i^{-2} + \frac{\lambda_1^2}{\beta\lambda_i^2(\lambda_i - \lambda_1)}\Big)\Big\}(\nabla_i F)^2 \\ &+ 2F\lambda_1^{-2}\sum_{i>j>\mu}\frac{\frac{\partial F}{\partial\lambda_i}(\lambda_i - \lambda_1)^2 - \frac{\partial F}{\partial\lambda_j}(\lambda_j - \lambda_1)^2}{(\lambda_i - \lambda_1)(\lambda_j - \lambda_1)(\lambda_i - \lambda_j)}h_{ij1}^2 + 2F\lambda_1^{-2}\frac{\partial F}{\partial\lambda_i}\sum_{i>\mu}(\lambda_i - \lambda_1)^{-1}h_{1ii}^2 \\ &+ F\lambda_1^{-2}\frac{\partial^2 F}{\partial\lambda_i\partial\lambda_j}h_{ii1}h_{jj1} + \frac{\beta-1}{\beta}\frac{\partial F}{\partial\lambda_i}(\frac{\lambda_i}{\lambda_1} - 1) - C\frac{\partial F}{\partial\lambda_i}\lambda_i(\frac{\lambda_i}{\lambda_1} - 1). \end{aligned}$$

Noticing the second term is nonnegative, we finish the proof.

Lemma 4.7. For $F = \sigma_k^{\alpha}$ and $C \leq 0$, if $\alpha > \frac{1}{k}$, then at the maximum point of \tilde{W} , $\lambda_1 = \cdots = \lambda_n$.

Proof. By Lemma 4.3, we know $J_1 \ge 0$ and the equality occurs if and only if $\lambda_1 = \cdots = \lambda_n$. And Lemma 2.3 implies $J_2 \ge 0$. Using Lemma 2.7, we have

$$J_{3} = \frac{\alpha(k\alpha - 1)}{k} \sigma_{k}^{2\alpha} \lambda_{1}^{-1} \left(\lambda_{1}^{-1} - \frac{2}{k} \frac{\sigma_{k-1;1}}{\sigma_{k}}\right) (\nabla_{1} \log \sigma_{k})^{2} + 2\alpha \sigma_{k}^{2\alpha - 1} \lambda_{1}^{-2} \sigma_{k-1;i} \sum_{i > \mu} (\lambda_{i} - \lambda_{1})^{-1} h_{1ii}^{2} + \alpha \sigma_{k}^{2\alpha - 1} \lambda_{1}^{-2} \sigma_{k-2;ij} h_{ii1} h_{jj1} + \alpha(\alpha - 1) \sigma_{k}^{2\alpha} \lambda_{1}^{-2} (\nabla_{1} \log \sigma_{k})^{2} \geq \alpha \sigma_{k}^{2\alpha} \lambda_{1}^{-1} \left(\frac{2k\alpha - k - 1}{k} \lambda_{1}^{-1} - 2 \frac{(k\alpha - 1)}{k^{2}} \frac{\sigma_{k-1;1}}{\sigma_{k}}\right) (\nabla_{1} \log \sigma_{k})^{2} + 2\alpha \sigma_{k}^{2\alpha - 1} \lambda_{1}^{-2} \sigma_{k-1;i} \sum_{i > \mu} (\lambda_{i} - \lambda_{1})^{-1} h_{1ii}^{2} + \alpha \sigma_{k}^{2\alpha} \lambda_{1}^{-2} (\nabla_{1} \log \sigma_{k})^{2} - \alpha \sigma_{k}^{2\alpha - 1} \lambda_{1}^{-2} \frac{\sigma_{k-1;i}}{\lambda_{i}} h_{ii1}^{2}.$$

Then using Lemma 4.4, we obtain

$$J_{3} \geq \alpha \sigma_{k}^{2\alpha} \lambda_{1}^{-1} \Big(\frac{2k\alpha - 1}{k} \lambda_{1}^{-1} - 2 \frac{(k\alpha - 1)}{k^{2}} \frac{\sigma_{k-1;1}}{\sigma_{k}} \Big) (\nabla_{1} \log \sigma_{k})^{2}$$
$$- \frac{\alpha}{k^{2}} \sigma_{k}^{2\alpha - 1} \frac{\sigma_{k-1;1}}{\sigma_{k}} \lambda_{1}^{-1} (\nabla_{1} \log \sigma_{k})^{2}$$
$$\geq \alpha \frac{2k\alpha - 1}{k^{2}} \sigma_{k}^{2\alpha} \lambda_{1}^{-2} \Big(k - \frac{\sigma_{k-1;1}\lambda_{1}}{\sigma_{k}} \Big) (\nabla_{1} \log \sigma_{k})^{2}$$
$$\geq \alpha \frac{(2k\alpha - 1)(n-1)}{nk} \sigma_{k}^{2\alpha} \lambda_{1}^{-2} (\nabla_{1} \log \sigma_{k})^{2},$$

thus $J_3 \geq 0$. For \mathcal{L} is an elliptic operator when $F = \sigma_k^{\alpha}$, at the maximum point of W, we have

$$0 \ge \mathcal{L}W \ge J_1 + J_2 + J_3 \ge 0.$$

Thus $J_1 = 0$, which implies $\lambda_1 = \cdots = \lambda_n$. Since \tilde{W} and W have the same maximum points, we finish the proof.

By similar discussion, for $F = S_k^{\alpha}$, we have the following lemma.

Lemma 4.8. For $F = S_k^{\alpha}$ and $C \leq 0$, if $k \geq 1$ and $\alpha > \frac{1}{k}$, then at the maximum point of \tilde{W} , $\lambda_1 = \cdots = \lambda_n$.

Proof. It is easy to check that $J_2 \ge 0$. We just show $J_3 \ge 0$ since the rest of the proof is similar to Lemma 4.7. Actually, for $F = S_k^{\alpha}$, we have

$$F\lambda_{1}^{-2} \frac{\partial^{2} F}{\partial \lambda_{i} \partial \lambda_{j}} h_{ii1} h_{jj1} = \alpha(\alpha - 1)\lambda_{1}^{-2} S_{k}^{2\alpha} (\nabla_{1} \log S_{k})^{2} + \alpha k(k - 1)\lambda_{1}^{-2} S_{k}^{2\alpha - 1} \lambda_{i}^{k - 2} h_{ii1}^{2}$$

$$\geq \alpha(\alpha - 1)\lambda_{1}^{-2} S_{k}^{2\alpha} (\nabla_{1} \log S_{k})^{2} + \frac{(k - 1)\alpha}{k} S_{k}^{2\alpha} \lambda_{1}^{-2} (\nabla_{1} \log S_{k})^{2}$$

$$= \frac{\alpha(k\alpha - 1)}{k} S_{k}^{2\alpha} \lambda_{1}^{-2} (\nabla_{1} \log S_{k})^{2},$$

where we use the Cauchy-Schwarz inequality for

$$(\nabla_1 \log S_k)^2 = k^2 \left(\frac{\lambda_i^{k-1}}{S_k} h_{ii1}\right)^2 \le k^2 \left(\sum_i \frac{\lambda_i^k}{S_k}\right) \left(\sum_i \frac{\lambda_i^{k-2}}{S_k} h_{ii1}^2\right) = k^2 \sum_i \frac{\lambda_i^{k-2}}{S_k} h_{ii1}^2.$$

Therefore,

$$J_{3} \geq \frac{\alpha(k\alpha - 1)}{k} S_{k}^{2\alpha} \lambda_{1}^{-1} (\lambda_{1}^{-1} - \frac{2\lambda_{1}^{k-1}}{S_{k}}) (\nabla_{1} \log S_{k})^{2} + \frac{\alpha(k\alpha - 1)}{k} S_{k}^{2\alpha} \lambda_{1}^{-2} (\nabla_{1} \log S_{k})^{2} = \frac{2\alpha(k\alpha - 1)}{k} S_{k}^{2\alpha} \lambda_{1}^{-2} (1 - \frac{\lambda_{1}^{k}}{S_{k}}) (\nabla_{1} \log S_{k})^{2} \geq 0.$$

5. Proof of Theorem A

In this section, by considering the quantity

$$Z = F \operatorname{tr} b - \frac{n(\beta - 1)}{2\beta} |X|^2,$$

we will prove Theorem A.

Lemma 5.1.

$$\mathcal{L}Z + R(\nabla Z) = L_1 + L_2 + L_3,$$

where $R(\nabla Z)$ denote the terms containing ∇Z ,

$$L_{1} = (\beta - 1)F \operatorname{tr} b - \frac{n(\beta - 1)}{\beta} \sum_{i} \frac{\partial F}{\partial \lambda_{i}} + C(n\beta F - \operatorname{tr} b \frac{\partial F}{\partial \lambda_{i}} \lambda_{i}^{2}),$$
$$L_{2} = \left(\frac{n(\beta - 1)}{\beta} \lambda_{i}^{-1} (2F^{-1} \frac{\partial F}{\partial \lambda_{i}} + \lambda_{i}^{-1}) - 2F^{-1} \frac{\partial F}{\partial \lambda_{i}} \operatorname{tr} b\right) (\nabla_{i} F)^{2}$$

and

$$L_{3} = 2F \frac{\partial F}{\partial \lambda_{i}} \lambda_{p}^{-2} \lambda_{q}^{-1} h_{pqi}^{2} + F \lambda_{p}^{-2} \frac{\partial^{2} F}{\partial \lambda_{i} \partial \lambda_{j}} h_{iip} h_{jjp} + F \lambda_{p}^{-2} \sum_{i \neq j} \left(\frac{\partial F}{\partial \lambda_{i}} - \frac{\partial F}{\partial \lambda_{j}} \right) (\lambda_{i} - \lambda_{j})^{-1} h_{ijp}^{2}.$$

Proof. By Proposition 3.1, we have

$$\begin{split} \mathcal{L}Z &= \langle X, \nabla(F\mathrm{tr}b) \rangle + (\beta - 1)F\mathrm{tr}b - \frac{n(\beta - 1)}{\beta} \sum_{i} \frac{\partial F}{\partial h_{ii}} \\ &+ C(n\beta F - \mathrm{tr}b \frac{\partial F}{\partial h_{ij}} h_{jl} h_{li}) + 2 \frac{\partial F}{\partial h_{ij}} \nabla_i F \nabla_j \mathrm{tr}b \\ &+ F b^{kp} b^{qk} \frac{\partial^2 F}{\partial h_{ij} \partial h_{st}} h_{ijp} h_{stq} + 2F b^{ks} b^{pt} b^{kq} \frac{\partial F}{\partial h_{ij}} h_{sti} h_{pqj}. \end{split}$$

From

$$\nabla_j Z = \mathrm{tr} b \nabla_j F + F \nabla_j \mathrm{tr} b - \frac{n(\beta - 1)}{\beta} \langle X, e_j \rangle,$$

we have

$$\nabla_{l}(F \mathrm{tr} b) \langle X, e_{l} \rangle = \nabla_{l} Z \langle X, e_{l} \rangle + \frac{n(\beta - 1)}{\beta} \sum_{l} \langle X, e_{l} \rangle^{2}$$
$$= \nabla_{l} Z \langle X, e_{l} \rangle + \frac{n(\beta - 1)}{\beta} \lambda_{l}^{-2} (\nabla_{l} F)^{2}$$

and

(5.1)

$$\nabla_i \operatorname{tr} b = F^{-1} (\nabla_i Z - \operatorname{tr} b \nabla_i F + \frac{n(\beta - 1)}{\beta} \lambda_i^{-1} \nabla_i F)$$

$$= F^{-1} \nabla_i Z + F^{-1} (-\operatorname{tr} b + \frac{n(\beta - 1)}{\beta} \lambda_i^{-1}) \nabla_i F.$$

Then, by Lemma 3.2, we have

$$\begin{aligned} \mathcal{L}Z + R(\nabla Z) &= (\beta - 1)F \mathrm{tr}b - \frac{n(\beta - 1)}{\beta} \sum_{i} \frac{\partial F}{\partial \lambda_{i}} + C(n\beta F - \mathrm{tr}b \frac{\partial F}{\partial \lambda_{i}} \lambda_{i}^{2}) \\ &+ \Big(\frac{n(\beta - 1)}{\beta} \lambda_{i}^{-1} (2F^{-1} \frac{\partial F}{\partial \lambda_{i}} + \lambda_{i}^{-1}) - 2F^{-1} \frac{\partial F}{\partial \lambda_{i}} \mathrm{tr}b \Big) (\nabla_{i}F)^{2} \\ &+ 2F \frac{\partial F}{\partial \lambda_{i}} \lambda_{p}^{-2} \lambda_{q}^{-1} h_{pqi}^{2} + F \lambda_{p}^{-2} \frac{\partial^{2} F}{\partial \lambda_{i} \partial \lambda_{j}} h_{iip} h_{jjp} \\ &+ F \lambda_{p}^{-2} \sum_{i \neq j} (\frac{\partial F}{\partial \lambda_{i}} - \frac{\partial F}{\partial \lambda_{j}}) (\lambda_{i} - \lambda_{j})^{-1} h_{ijp}^{2}. \end{aligned}$$

Lemma 5.2. For $F = \sigma_k^{\alpha}$, if $k\alpha \ge 1$ and $C \le 0$, then $L_1 \ge 0$. In particular, for $F = \sigma_n^{\alpha}$, if $C \le 0$ and $\alpha \ge 0$, then $L_1 \ge 0$.

Proof. For $F = \sigma_k^{\alpha}$ and $C \leq 0$, by Newton-Maclaurin inequality and $\operatorname{tr} b = \frac{\sigma_{n-1}}{\sigma_n}$, we have

$$L_{1} = (k\alpha - 1)\sigma_{k}^{\alpha}(\operatorname{tr} b - \frac{n(n-k+1)\sigma_{k-1}}{k\sigma_{k}}) - \alpha C\sigma_{k}^{\alpha} \Big(-nk + \operatorname{tr} b\Big(\sigma_{1} - (k+1)\frac{\sigma_{k+1}}{\sigma_{k}}\Big) \Big)$$

$$\geq (k\alpha - 1)\sigma_{k}^{\alpha} \Big(\frac{\sigma_{n-1}}{\sigma_{n}} - \frac{n(n-k+1)\sigma_{k-1}}{k\sigma_{k}}\Big) - \frac{k\alpha}{n} C\sigma_{k}^{\alpha} \Big(-n^{2} + \operatorname{tr} b\sigma_{1} \Big)$$

$$\geq 0.$$

Theorem 5.3. For $F = \sigma_k^{\alpha}$ and $C \leq 0$, if $\alpha > \frac{1}{k}$, then a strictly convex closed solution of (1.5) is a round sphere.

Proof. For $F = \sigma_k^{\alpha}$,

$$L_{2} = \left(\frac{n(k\alpha - 1)}{k\alpha}\lambda_{i}^{-1}\left(\frac{2\alpha\sigma_{k-1;i}}{\sigma_{k}} + \lambda_{i}^{-1}\right) - \frac{2\alpha\sigma_{k-1;i}}{\sigma_{k}}\operatorname{tr}b\right)(\nabla_{i}\sigma_{k}^{\alpha})^{2}$$
$$= \alpha\sigma_{k}^{2\alpha}\left(\frac{n(k\alpha - 1)}{k}\lambda_{i}^{-2}\left(\frac{2\alpha\sigma_{k-1;i}\lambda_{i}}{\sigma_{k}} + 1\right) - \frac{2\alpha^{2}\sigma_{k-1;i}}{\sigma_{k}}\operatorname{tr}b\right)(\nabla_{i}\log\sigma_{k})^{2}$$

and

$$\begin{split} L_{3} &= 2\alpha\sigma_{k}^{2\alpha-1}\sigma_{k-1;i}\lambda_{p}^{-2}\lambda_{q}^{-1}h_{pqi}^{2} + \alpha(\alpha-1)\sigma_{k}^{2\alpha-2}\lambda_{p}^{-2}(\nabla_{p}\sigma_{k})^{2} \\ &+ \alpha\sigma_{k}^{2\alpha-1}\lambda_{p}^{-2}\sigma_{k-2;ij}(h_{iip}h_{jjp} - h_{ijp}^{2}) \\ &= \alpha(\alpha-1)\sigma_{k}^{2\alpha}\lambda_{p}^{-2}(\nabla_{p}\log\sigma_{k})^{2} + \alpha\sigma_{k}^{2\alpha-1}\lambda_{p}^{-2}(\sigma_{k-1;i}\lambda_{i}^{-1}h_{iip}^{2} + \sigma_{k-2;ij}h_{iip}h_{jjp}) \\ &+ \alpha\sigma_{k}^{2\alpha-1}\lambda_{p}^{-2}\sum_{i\neq j}(2\sigma_{k-1;i}\lambda_{j}^{-1} - \sigma_{k-2;ij})h_{ijp}^{2} + \alpha\sigma_{k}^{2\alpha-1}\lambda_{p}^{-2}\sigma_{k-1;i}\lambda_{i}^{-1}h_{iip}^{2}. \end{split}$$

By using Lemma 2.7, $\sigma_{k-1;i}\lambda_j^{-1} \ge \sigma_{k-2;ij}$ and

$$k\sum_{i}\frac{\sigma_{k-1;i}}{\lambda_{i}\sigma_{k}}h_{iip}^{2} = \left(\sum_{i}\frac{\lambda_{i}\sigma_{k-1;i}}{\sigma_{k}}\right)\left(\sum_{i}\frac{\sigma_{k-1;i}}{\lambda_{i}\sigma_{k}}h_{iip}^{2}\right) \ge \left(\sum_{i}\frac{\sigma_{k-1;i}}{\sigma_{k}}h_{iip}\right)^{2}.$$

on the second, the third and the last terms, respectively, we obtain

$$L_{3} \geq \alpha(\alpha - 1)\sigma_{k}^{2\alpha}\lambda_{p}^{-2}(\nabla_{p}\log\sigma_{k})^{2} + \alpha\sigma_{k}^{2\alpha}\lambda_{p}^{-2}(\nabla_{p}\log\sigma_{k})^{2} + \frac{\alpha}{k}\sigma_{k}^{2\alpha}\lambda_{p}^{-2}(\nabla_{p}\log\sigma_{k})^{2}$$
$$= \frac{\alpha(k\alpha + 1)}{k}\sigma_{k}^{2\alpha}\lambda_{p}^{-2}(\nabla_{p}\log\sigma_{k})^{2}.$$

Then,

$$L_2 + L_3 \ge \alpha \sigma_k^{2\alpha} \Big(\lambda_i^{-2} \Big(\frac{n(k\alpha - 1)}{k} \frac{2\alpha \sigma_{k-1;i}\lambda_i}{\sigma_k} + \frac{(n+1)k\alpha - n + 1}{k}\Big) - \frac{2\alpha^2 \sigma_{k-1;i}}{\sigma_k} \operatorname{tr} b\Big) (\nabla_i \log \sigma_k)^2.$$

Assume that x_0 is a maximum point of \tilde{W} . Then it is follows from Lemma 4.7 that x_0 is an umbilic point. At x_0 , thus we have

$$\begin{split} \lambda_i^{-2} &(\frac{n(k\alpha-1)}{k} \frac{2\alpha\sigma_{k-1;i}\lambda_i}{\sigma_k} + \frac{(n+1)k\alpha - n + 1}{k}) - \frac{2\alpha^2\sigma_{k-1;i}}{\sigma_k} \mathrm{tr} b \\ &= \lambda_i^{-2} (\frac{n(k\alpha-1)}{k} \frac{2k\alpha}{n} + \frac{(n+1)k\alpha - n + 1}{k} - \frac{2\alpha^2k}{n}n) \\ &= \lambda_i^{-2} (n-1)(\alpha - \frac{1}{k}) > 0. \end{split}$$

Since $Z \leq n\tilde{W} \leq nW(x_0) = Z(x_0)$, Z attains its maximum at x_0 . Hence, there exists a neighborhood of x_0 , denoted by U, such that in U, $\mathcal{L}Z + R(\nabla Z) \geq 0$. By the strong maximum principle, we know $Z = Z(x_0)$ is constant in U, which implies \tilde{W} is also constant in U. Then the set of points where \tilde{W} attains its maximum is an open set. Hence \tilde{W} is constant on M, which implies that M is totally umbilic. \Box

In order to discuss $F=\sigma_k^\alpha$ further, we need the following lemma.

Lemma 5.4. Suppose $y_i \in \mathbb{R}$ and $t_i = \frac{\sigma_k}{\lambda_i \sigma_{k-1;i}}$ for $1 \le i \le n$. For any $1 \le m \le n$, the following inequality holds

$$\sum_{i} t_{i} y_{i}^{2} - 4\alpha y_{m} (\sum_{i} y_{i}) \ge \left(\frac{1}{k} (\frac{2\alpha}{t_{m}} - 1)^{2} - \frac{4\alpha^{2}}{t_{m}}\right) (\sum_{i} y_{i})^{2}.$$

Proof. If $\sum_i y_i = 0$, the inequality is trivial. If $\sum_i y_i \neq 0$, we may assume $\sum_i y_i = 1$. In fact, we will estimate the minimum of

$$f(y_1, \dots, y_n) = \sum_i t_i y_i^2 - 4\alpha y_m$$

under the condition $\sum_i y_i = 1$. Using Lagrangian multiplier technique, we solve the following equations for $\tilde{f} = f + \tau(\sum_i y_i - 1)$,

$$0 = \frac{\partial}{\partial y_i} \tilde{f} = 2t_i y_i - 4\alpha \delta_{im} + \tau,$$

$$0 = \frac{\partial}{\partial \tau} \tilde{f} = \sum_i y_i - 1.$$

And, using $\sum_i \lambda_i \sigma_{k-1;i} = k \sigma_k$, we have $y_i = \frac{2\alpha \delta_{im}}{ti} - \frac{1}{2t_i} \tau$ and $\tau = \frac{4\alpha}{kt_m} - \frac{2}{k}$. Thus, $y_i = \frac{1}{t_i} (2\alpha \delta_{im} - \frac{2\alpha}{kt_m} + \frac{1}{k})$. Because $t_i > 0$, we know

$$f_{\min} = \sum_{i} \frac{1}{t_{i}} (2\alpha \delta_{im} - \frac{2\alpha}{kt_{m}} + \frac{1}{k})^{2} - \frac{4\alpha}{t_{m}} (2\alpha - \frac{2\alpha}{kt_{m}} + \frac{1}{k})$$

$$= \sum_{i \neq m} \frac{1}{k^{2}t_{i}} (\frac{2\alpha}{t_{m}} - 1)^{2} + \frac{1}{t_{m}} (-2\alpha + \frac{2\alpha}{kt_{m}} - \frac{1}{k})(2\alpha + \frac{2\alpha}{kt_{m}} - \frac{1}{k})$$

$$= \sum_{i} \frac{1}{k^{2}t_{i}} (\frac{2\alpha}{t_{m}} - 1)^{2} - \frac{4\alpha^{2}}{t_{m}}$$

$$= \frac{1}{k} (\frac{2\alpha}{t_{m}} - 1)^{2} - \frac{4\alpha^{2}}{t_{m}}.$$

Now, we use (5.1) to estimate L_2 and L_3 in a different way.

Theorem 5.5. For $F = \sigma_k^{\alpha}$ and $C \leq 0$, if $2 \leq k \leq n-1$ and $\frac{1}{k} \leq \alpha \leq \frac{1}{2}$ or $k = \alpha = 1$, the strictly convex closed solution of (1.5) is a round sphere. For $F = \sigma_n^{\alpha}$ and C < 0, if $\frac{1}{n+2} \leq \alpha \leq \frac{1}{2}$, the strictly convex closed solution of (1.5) is a round sphere.

$$\begin{aligned} Proof. \text{ Using Lemma 2.7 and } \sigma_{k-1;i}\lambda_j^{-1} - \sigma_{k-2;ij} &> 0, \text{ we have} \\ \frac{1}{\alpha\sigma_k^{2\alpha}}L_3 &= (\alpha - 1)\lambda_p^{-2}(\nabla_p \log \sigma_k)^2 + \sigma_k^{-1}\lambda_p^{-2}(\sigma_{k-1;i}\lambda_i^{-1}h_{iip}^2 + \sigma_{k-2;ij}h_{iip}h_{jjp}) \\ &+ \frac{\sigma_{k-1;i}}{\sigma_k}\lambda_p^{-2}\lambda_i^{-1}h_{iip}^2 + \sigma_k^{-1}\sum_{\neq}\lambda_p^{-2}(2\sigma_{k-1;i}\lambda_j^{-1} - \sigma_{k-2;ij})h_{ijp}^2 \\ &+ 2\sigma_k^{-1}\sum_{i\neq j}\lambda_i^{-2}(\sigma_{k-1;i}\lambda_j^{-1} - \sigma_{k-2;ij})h_{iij}^2 + 2\sum_{i\neq j}\frac{\sigma_{k-1;i}}{\sigma_k}\lambda_j^{-3}h_{ijj}^2 \\ &\geq \alpha\lambda_p^{-2}(\nabla_p \log \sigma_k)^2 + \frac{\sigma_{k-1;i}}{\sigma_k}\lambda_p^{-2}\lambda_i^{-1}h_{iip}^2 + 2\sum_{i\neq j}\frac{\sigma_{k-1;i}}{\sigma_k}\lambda_j^{-3}h_{ijj}^2. \end{aligned}$$

Furthermore,

$$\frac{1}{\alpha \sigma_k^{2\alpha}} (L_2 + L_3) \ge \lambda_i^{-1} \Big(\frac{2\alpha n (k\alpha - 1)}{k} \frac{\sigma_{k-1;i}}{\sigma_k} + \frac{(n+1)k\alpha - n}{k} \lambda_i^{-1} \Big) (\nabla_i \log \sigma_k)^2 \\ - \frac{2\alpha^2 \sigma_{k-1;i}}{\sigma_k} \operatorname{trb}(\nabla_i \log \sigma_k)^2 + 2 \sum_{i \neq j} \frac{\sigma_{k-1;i}}{\sigma_k} \lambda_j^{-3} h_{ijj}^2 \\ + \frac{\sigma_{k-1;i}}{\sigma_k} \lambda_p^{-2} \lambda_i^{-1} h_{iip}^2.$$

It follows from (5.1) that

(5.2)
$$-\frac{n(k\alpha-1)}{k}\lambda_i^{-1}\frac{\sigma_{k-1;p}}{\sigma_k}h_{ppi} = R(\nabla Z) + \sum_p \lambda_p^{-1}(\lambda_p^{-1}h_{ppi} - \alpha\frac{\sigma_{k-1;q}}{\sigma_k}h_{qqi}).$$

By using (5.2), we can estimate the following two terms

$$-\frac{2\alpha^2 \sigma_{k-1;i}}{\sigma_k} \operatorname{tr} b(\nabla_i \log \sigma_k)^2 + 2 \sum_{i,j} \frac{\sigma_{k-1;i}}{\sigma_k} \lambda_j^{-3} h_{ijj}^2$$

= $2 \sum_i \frac{\sigma_{k-1;i}}{\sigma_k} \sum_j \lambda_j^{-1} (\lambda_j^{-2} h_{ijj}^2 - \alpha^2 (\nabla_i \log \sigma_k)^2)$
= $2 \sum_i \frac{\sigma_{k-1;i}}{\sigma_k} \sum_j \lambda_j^{-1} (\lambda_j^{-1} h_{ijj} - \alpha \nabla_i \log \sigma_k)^2 - \frac{4\alpha n (k\alpha - 1)}{k} \frac{\sigma_{k-1;i}}{\sigma_k} \lambda_i^{-1} (\nabla_i \log \sigma_k)^2 + R(\nabla Z).$

Therefore,

$$\frac{1}{\alpha \sigma_k^{2\alpha}} (L_2 + L_3) + R(\nabla Z)$$

$$\geq \lambda_i^{-1} \Big(-\frac{2\alpha((n-1)k\alpha - n)}{k} \frac{\sigma_{k-1;i}}{\sigma_k} + \frac{(n+1)k\alpha - n}{k} \lambda_i^{-1} \Big) (\nabla_i \log \sigma_k)^2 + 2\sum_i \frac{\sigma_{k-1;i}}{\sigma_k} \sum_{j \neq i} \lambda_j^{-1} (\lambda_j^{-1} h_{ijj} - \alpha \nabla_i \log \sigma_k)^2 + \lambda_i^{-2} \frac{\sigma_{k-1;p}}{\sigma_k} \lambda_p^{-1} h_{ppi}^2 - 4\alpha \frac{\sigma_{k-1;i}}{\sigma_k} \lambda_i^{-2} h_{iii} \nabla_i \log \sigma_k.$$

Let $t_i = \frac{\sigma_k}{\lambda_i \sigma_{k-1;i}}$ and using Lemma 5.4, we have

$$\lambda_i^{-2} \frac{\sigma_{k-1;p}}{\sigma_k} \lambda_p^{-1} h_{ppi}^2 - 4\alpha \frac{\sigma_{k-1;i}}{\sigma_k} \lambda_i^{-2} h_{iii} \nabla_i \log \sigma_k$$

= $\sum_i \lambda_i^{-2} \Big\{ \sum_p t_p \Big(\frac{\sigma_{k-1;p}}{\sigma_k} h_{ppi} \Big)^2 - 4\alpha \frac{\sigma_{k-1;i}}{\sigma_k} h_{iii} \Big(\sum_p \frac{\sigma_{k-1;p}}{\sigma_k} h_{ppi} \Big) \Big\}$
\ge $\sum_i \lambda_i^{-2} \Big(\frac{1}{k} (\frac{2\alpha}{t_i} - 1)^2 - \frac{4\alpha^2}{t_i} \Big) (\nabla_i \log \sigma_k)^2.$

Then, we have

$$\begin{aligned} &\frac{1}{\alpha \sigma_k^{2\alpha}} (L_2 + L_3) + R(\nabla Z) \\ &\ge \sum_i \lambda_i^{-2} \Big\{ \frac{2\alpha^2}{t_i} (\frac{2}{kt_i} - n - 1) + \alpha (\frac{2(n-2)}{kt_i} + n + 1) - \frac{n-1}{k} \Big\} (\nabla_i \log \sigma_k)^2 \\ &= \sum_i \lambda_i^{-2} \Big\{ (\frac{2\alpha}{t_i} - 1) ((\frac{2}{kt_i} - n - 1)\alpha + \frac{n-1}{k}) \Big\} (\nabla_i \log \sigma_k)^2. \end{aligned}$$

Since $t_i \ge 1$, if $k \ge 2$ and $\alpha \in [\frac{n-1}{k(n+1)-2}, \frac{1}{2}]$ or $k = \alpha = 1$, then $L_2 + L_3 \ge 0$. Since $L_1 \ge 0$, by the strong maximum principle, we know Z is constant. Hence, $L_1 = L_2 + L_3 = 0$. In case C < 0 or $\alpha > \frac{1}{k}$, $L_1 = 0$ implies M is totally umbilic; in other cases, $L_2 + L_3 = 0$ implies that the second fundamental form is parallel. Either of these implies the solution is a round sphere.

Proof of Theorem A. Combining Theorem 5.3 with Theorem 5.5, we complete the proof of Theorem A. \Box

6. Proof of Theorem B

For $F = S_k^{\alpha}$, by similar discussion, we have

Lemma 6.1. For $F = S_k^{\alpha}$ and $C \leq 0$, if $k \geq 1$ and $k\alpha \geq 1$, then $L_1 \geq 0$.

Proof. For $F = S_k^{\alpha}$ and $C \leq 0$, by $\frac{S_{k+1}}{S_k} \geq \frac{S_1}{n}$, we have

$$L_1 = (k\alpha - 1)S_k^{\alpha}(\operatorname{tr} b - \frac{nS_{k-1}}{S_k}) - k\alpha CS_k^{\alpha - 1}\left(\operatorname{tr} bS_{k+1} - nS_k\right) \ge 0.$$

Theorem 6.2. For $F = S_k^{\alpha}$ and $C \leq 0$, if $k \geq 1$ and $\alpha > \frac{1}{k}$, solution of (1.5) is a round sphere.

Proof. For $F = S_k^{\alpha}$,

$$L_2 = \alpha S_k^{2\alpha} \left(\frac{n(k\alpha - 1)}{k} \lambda_p^{-2} \left(\frac{2k\alpha}{S_k} \lambda_p^k + 1 \right) - 2k\alpha^2 \frac{\lambda_p^{k-1}}{S_k} \operatorname{tr} B \right) (\nabla_p \log S_k)^2$$

and

$$L_{3} \geq 2k\alpha S_{k}^{2\alpha-1}\lambda_{i}^{k-1}\lambda_{p}^{-2}\lambda_{q}^{-1}h_{pqi}^{2} + \alpha(\alpha-1)S_{k}^{2\alpha}\lambda_{p}^{-2}(\nabla_{p}\log S_{k})^{2}$$
$$+ \alpha k(k-1)S_{k}^{2\alpha-1}\lambda_{p}^{-2}\lambda_{i}^{k-2}h_{iip}^{2}$$
$$\geq \alpha k(k+1)S_{k}^{2\alpha-1}\lambda_{p}^{-2}\lambda_{i}^{k-2}h_{iip}^{2} + \alpha(\alpha-1)S_{k}^{2\alpha}\lambda_{p}^{-2}(\nabla_{p}\log S_{k})^{2}$$
$$\geq \frac{\alpha(k+1)}{k}S_{k}^{2\alpha}\lambda_{p}^{-2}(\nabla_{p}\log S_{k})^{2} + \alpha(\alpha-1)S_{k}^{2\alpha}\lambda_{p}^{-2}(\nabla_{p}\log S_{k})^{2},$$

where the last inequality is from Cauchy-Schwarz inequality for

$$(\nabla_p \log S_k)^2 = k^2 \left(\frac{\lambda_i^{k-1}}{S_k} h_{iip}\right)^2 \le k^2 \left(\sum_i \frac{\lambda_i^k}{S_k}\right) \left(\sum_i \frac{\lambda_i^{k-2}}{S_k} h_{iip}^2\right) = k^2 \sum_i \frac{\lambda_i^{k-2}}{S_k} h_{iip}^2.$$

Then,

$$L_{2} + L_{3} \ge \alpha S_{k}^{2\alpha} \Big\{ 2n\alpha(k\alpha - 1)\frac{\lambda_{p}^{k-2}}{S_{k}} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{-2} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{k-1} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_{p}^{k-1} - 2k\alpha^{2}\frac{\lambda_{p}^{k-1}\mathrm{tr}b}{S_{k}} \Big\} (\nabla_{p}\log S_{k})^{2} + 2k\alpha^{2$$

At an umbilic point, we have

$$2n\alpha(k\alpha - 1)\frac{\lambda_p^{k-2}}{S_k} + \frac{(n+1)k\alpha - n + 1}{k}\lambda_p^{-2} - 2k\alpha^2 \frac{\lambda_p^{k-1}\text{tr}b}{S_k}$$

= $\lambda_p^{-2}(2\alpha(k\alpha - 1) + \frac{(n+1)k\alpha - n + 1}{k} - 2k\alpha^2)$
= $\frac{(n-1)(k\alpha - 1)}{k}\lambda_p^{-2}$
> 0.

The rest of the proof is similar to Theorem 5.3.

Theorem 6.3. For $F = S_k^{\alpha}$ and $C \leq 0$, if $k \geq 1$ and $\alpha = \frac{1}{k}$, the solution of (1.5) is a round sphere.

Proof. In fact, for $F = S_k^{\alpha}$ and $\alpha = \frac{1}{k}$, we have

$$L_2 = -\frac{2}{k^2} S_k^{\frac{2-k}{k}} \operatorname{tr} b\lambda_i^{k-1} (\nabla_i \log S_k)^2$$

and

$$L_{3} \geq 2S_{k}^{\frac{2-k}{k}}\lambda_{i}^{k-1}\lambda_{p}^{-3}h_{ppi}^{2} + \frac{1-k}{k^{2}}\lambda_{p}^{-2}(\nabla_{p}\log S_{k})^{2} + (k-1)S_{k}^{\frac{2-k}{k}}\lambda_{p}^{-2}\lambda_{i}^{k-2}h_{iip}^{2} \\ \geq 2S_{k}^{\frac{2-k}{k}}\lambda_{i}^{k-1}\lambda_{p}^{-3}h_{ppi}^{2},$$

where the last inequality is from Cauchy-Schwarz inequality for

$$(\nabla_p \log S_k)^2 = k^2 \left(\frac{\lambda_i^{k-1}}{S_k} h_{iip}\right)^2 \le k^2 \left(\sum_i \frac{\lambda_i^k}{S_k}\right) \left(\sum_i \frac{\lambda_i^{k-2}}{S_k} h_{iip}^2\right) = k^2 \sum_i \frac{\lambda_i^{k-2}}{S_k} h_{iip}^2.$$

By (5.1) for $\alpha = \frac{1}{k}$, we have

(6.1)
$$R(\nabla Z) + \sum_{p} \lambda_p^{-1} (\lambda_p^{-1} h_{ppi} - \frac{1}{k} \nabla_i \log S_k) = 0.$$

Using (6.1), we have

$$L_{2} + L_{3} + R(\nabla Z) = 2S_{k}^{\frac{2-k}{k}} \lambda_{i}^{k-1} \sum_{p} \lambda_{p}^{-1} (\lambda_{p}^{-2} h_{ppi}^{2} - \frac{1}{k^{2}} (\nabla_{i} \log S_{k})^{2})$$
$$= 2S_{k}^{\frac{2-k}{k}} \lambda_{i}^{k-1} \sum_{p} \lambda_{p}^{-1} (\lambda_{p}^{-1} h_{ppi} - \frac{1}{k} \nabla_{i} \log S_{k})^{2}$$
$$\geq 0.$$

Since $L_1 \ge 0$, by the strong maximum principle, we know Z is constant. Hence, $L_1 = L_2 + L_3 = 0$. This implies that the solution is a sphere.

Proof of Theorem B. Combining Theorem 6.2 with Theorem 6.3, we complete the proof of Theorem B. \Box

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