

Projective Peridynamic Modeling of Hyperelastic Membranes with Contact: Supplementary Technical Document

April 7, 2023

This material serves as a supplementary technical document, so we suppose readers have read the original article. Some notation will not be repeated and share the same definition if it is not being mentioned again.

1 Extended Shape tensor and deformation tensor

We will first show that the extended shape tensor \mathbf{K}_i is non-singular, so that deformation gradient can be calculated as Eq.(15) in the article.

1.1 Non-singularity of Extended Shape Tensor

Silling and Lehoucq [3] demonstrate that if the deformation is sufficiently smooth, the peridynamic stress tensor converges to a Piola-Kirchhoff stress tensor, which is a function only of the local deformation gradient tensor, as in the classical theory. Originally, the 3×3 deformation gradient tensor \mathbf{F}_i has

$$\mathbf{F}_i = \left(\sum_j \omega_{ij} (\mathbf{y}_j - \mathbf{y}_i) (\mathbf{x}_j - \mathbf{x}_i)^T \right) \mathbf{K}_i^{-1}, \quad (1)$$

the 3×3 shape tensor \mathbf{K}_i calculated as

$$\mathbf{K}_i = \sum_j \omega_{ij} (\mathbf{x}_j - \mathbf{x}_i) (\mathbf{x}_j - \mathbf{x}_i)^T, \quad (2)$$

where ω is a **scalar non-negative state** acting as a weighting function, \mathbf{x}, \mathbf{y} represented the rest shape and currently deformed shape.

As for a co-dimensional membrane/plane model, it is possible that the rest shape remains to lay on a plane of \mathbb{R}^3 . Denoted the plane as

$$\mathbf{v} + \mathbf{X} \quad (3)$$

where \mathbf{v} is a transform and \mathbf{X} is hyperplane with two chosen orthonormal vector bases in \mathbb{R}^3 as $\mathbf{e}_1, \mathbf{e}_2$ (**distincted from the conventional orthonormal vector bases of \mathbb{R}^3** , e.g., $\mathbf{e}_1 = (1, 0, 0)^T$, which are not necessarily or able to be chosen for the hyperplane we discussed above), then we have $\mathbf{X} = \{q^1 \mathbf{e}_1 + q^2 \mathbf{e}_2 | \forall q^1, q^2 \in \mathbb{R}\}$, and $\mathbb{R}^3 = \mathbf{X} \oplus \mathbf{X}^\perp, \mathbf{X} \cap \mathbf{X}^\perp = \emptyset$, where \mathbf{X}^\perp is the orthogonal complement space of \mathbf{X} , and \oplus denotes the direct sum operator. It is simple to show that, with the concept of orthogonal complement space and direct sum, by denoting the orthonormal vector base of \mathbf{X}^\perp as \mathbf{e}_3 , we then have $\mathbf{X}^\perp = \{q^3 \mathbf{e}_3 | \forall q^3 \in \mathbb{R}, \mathbf{e}_3 = \pm \mathbf{e}_1 \times \mathbf{e}_2\}$, since only $\mathbf{e}_3 = \pm \mathbf{e}_1 \times \mathbf{e}_2$ provides orthogonality to \mathbf{e}_1 and \mathbf{e}_2 with unit length. To make a convention, we use the definition of $\mathbf{e}_3 = \mathbf{e}_1 \times \mathbf{e}_2$.

Extended shape tensor

$$\mathbf{K}_i = \sum_j \omega_{ij} (\mathbf{x}_j - \mathbf{x}_i) (\mathbf{x}_j - \mathbf{x}_i)^T + \mathbf{x}_i^\perp \otimes \mathbf{x}_i^\perp, \quad (4)$$

where $\mathbf{x}_i^\perp = \xi \text{norm} \left(\frac{\sum_J \theta_J \mathbf{n}_J}{\sum_J \theta_J} \right)$, J denotes triangle contained vertex i with index J , θ_J is the inner angle of triangle J at vertex i , \mathbf{n}_J is the normal of triangle J represented as norm $((\mathbf{x}_J^2 - \mathbf{x}_J^0) \times (\mathbf{x}_J^1 - \mathbf{x}_J^0))$, where the superscript $(\cdot)^{\{1,2,3\}}$ denotes three related vertexes in triangle and $\text{norm}(\cdot)$ denotes normalization operator.

We use $\{\mathbf{x}\}$ to denote all vertexes in the neighborhood of vertex i on rest shape, and $\{\mathbf{y}\}$ on the deformed shape. The adjacency relation is chosen as 1-neighborhood, which means that $\{\mathbf{x}\}$ collects all vertexes connected to vertex i with an edge, so that any vertex in $\{\mathbf{x}\}$ will be in a triangle which contain vertex i . (The generality will not get limited by such neighborhood chosen, since if a minimal neighborhood can guarantee the non-singularity of shape tensor, so does an expanded one). As $\{\mathbf{x}\}$ is a point set, $\{\mathbf{x} - \mathbf{x}_i\} := \{\mathbf{x} - \mathbf{x}_i | \forall \mathbf{x} \in \{\mathbf{x}\}\}$ provides the corresponding vector set. With the above illustration, if neighbors of vertex i lay on a plane, the Eq. (3) can be describe as $(\mathbf{x}_i - \mathbf{o}) + \mathbf{X}_i$, where \mathbf{X}_i is the spanning space of $\{\mathbf{x} - \mathbf{x}_i\}$ and \mathbf{o} the original point. Also we have \mathbf{X}_i is a hyperplane of \mathbb{R}^3 , i.e. $\forall \mathbf{u} \in \mathbf{X}_i = \{q^1 \mathbf{e}_1 + q^2 \mathbf{e}_2 | \forall q^1, q^2 \in \mathbb{R}\}$. An illustrative diagram on Fig. 1 will outline such case.

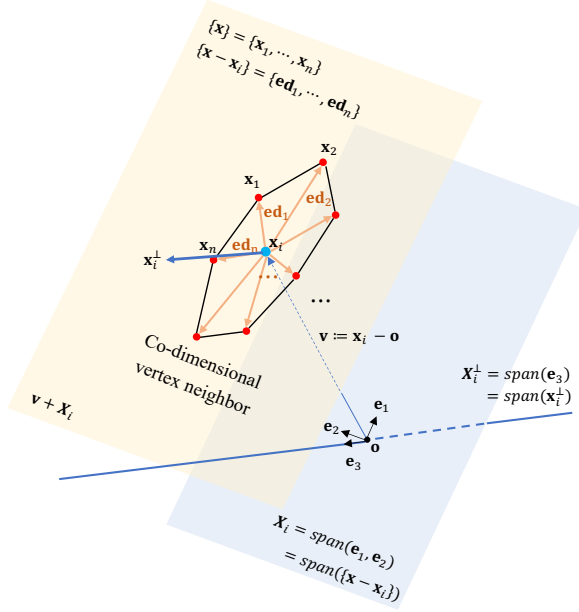


Figure 1: Illustrative diagram of the co-dimensional case.

We show that, in this case, the extended shape tensor is non-singular.

On the other hand, we also show that if $\{\mathbf{x} - \mathbf{x}_i\}$ span \mathbb{R}^3 , the extended shape tensor will also keep its non-singularity.

We use some mathematical convention if not mentioned, e.g., \otimes the Kronecker product and concepts on linear algebra.

Given Eq. (4), we proof the following two lemmas:

Lemma 1 $span(\{\mathbf{x} - \mathbf{x}_i\}) := \mathbf{X}_i = \{q^1 \mathbf{e}_1 + q^2 \mathbf{e}_2 | \forall q^1, q^2 \in \mathbb{R}\}$, then we have $span(\{\mathbf{x} - \mathbf{x}_i\} \cup \{\mathbf{x}_i^\perp\}) = \mathbb{R}^3$, as \mathbf{x}_i^\perp is given to be $\mathbf{x}_i^\perp = \xi \text{norm} \left(\frac{\sum_J \theta_J \mathbf{n}_J}{\sum_J \theta_J} \right)$.

Proof. Because of the adjacency relationship chosen above, we have that the collection set of vertexes of any triangle J which contain vertex i is the same as $\{\mathbf{x}\} \cup \{\mathbf{x}_i\}$ (See the diagram and the above discussion for an outline). Then we have $\mathbf{n}_J = \text{norm}((\mathbf{x}_j^2 - \mathbf{x}_j^0) \times (\mathbf{x}_j^1 - \mathbf{x}_j^0))$, the superscript denotes inner vertexes of triangle and they also in set $\{\mathbf{x}\} \cup \{\mathbf{x}_i\}$. If $\mathbf{x}_j^0 = \mathbf{x}_i$, then we will always have $(\mathbf{x}_j^2 - \mathbf{x}_j^0) \in \{\mathbf{x} - \mathbf{x}_i\}$, $(\mathbf{x}_j^1 - \mathbf{x}_j^0) \in \{\mathbf{x} - \mathbf{x}_i\}$. If \mathbf{x}_j^0 is not \mathbf{x}_i , then we will have, i.e., like $\mathbf{x}_j^2 = \mathbf{x}_i$, then $(\mathbf{x}_j^2 - \mathbf{x}_j^0) \in span(\{\mathbf{x} - \mathbf{x}_i\})$, $(\mathbf{x}_j^2 - \mathbf{x}_j^1) \in span(\{\mathbf{x} - \mathbf{x}_i\})$, and thus $\mathbf{x}_j^1 - \mathbf{x}_j^0 = (\mathbf{x}_j^1 - \mathbf{x}_j^2) - (\mathbf{x}_j^0 - \mathbf{x}_j^2) \in span(\{\mathbf{x} - \mathbf{x}_i\})$.

This gives the edges of triangle J , i.e., $\mathbf{ed}_{(\cdot)} := \mathbf{x}_j^{(\cdot)} - \mathbf{x}_j^0 \in span(\{\mathbf{x} - \mathbf{x}_i\})$, so that $\mathbf{ed}_1 = q_1^1 \mathbf{e}_1 + q_1^2 \mathbf{e}_2$, $\mathbf{ed}_2 = q_2^1 \mathbf{e}_1 + q_2^2 \mathbf{e}_2$. Then $\mathbf{n}_J = \text{norm}(\mathbf{ed}_1 \times \mathbf{ed}_2) = \text{norm}((q_1^1 \mathbf{e}_1 + q_1^2 \mathbf{e}_2) \times (q_2^1 \mathbf{e}_1 + q_2^2 \mathbf{e}_2)) = \text{norm}((q_1^1 q_2^2 - q_1^2 q_2^1) \mathbf{e}_3)$. As the term $(q_1^1 q_2^2 - q_1^2 q_2^1)$ is the determinant, i.e., double area of the planar triangle, if the triangle keeps non-degenerated, then this term will keep non-zero. If all triangle degenerated to a segment, then the co-dimensional material will remain to be a line, which is not the cases we design. Then we have $\mathbf{n}_J = \text{norm}((q_1^1 q_2^2 - q_1^2 q_2^1) \mathbf{e}_3) = \mathbf{e}_3$. Here our primitive are manipulated through an algorithm to keep the vertex and triangle norm remain outer.

Then following $\mathbf{x}_i^\perp = \xi \text{norm}(\sum_J \theta_J \mathbf{n}_J / \sum_J \theta_J)$, it is simply to verify that $\mathbf{x}_i^\perp = \xi \mathbf{e}_3$, so that we have $\text{span}(\{\mathbf{x} - \mathbf{x}_i\} \cup \{\mathbf{x}_i^\perp\}) = \mathbb{R}^3$. \square

Lemma 2 Spectral Representation.

Given a vector set (with n vectors) $\{\mathbf{u}_i\}$ spanning \mathbb{R}^3 , i.e., $\text{span}(\{\mathbf{u}_i\}) = \mathbb{R}^3 = \{q^1 \mathbf{e}_1 + q^2 \mathbf{e}_2 + q^3 \mathbf{e}_3 | \forall q^1, q^2, q^3 \in \mathbb{R}\}$, its generated second-order tensor $\mathbf{U} = \sum_{i=1}^n \omega_i \mathbf{u}_i \otimes \mathbf{u}_i$, $\omega_i > 0$ has **spectral representation** as $\mathbf{U} = \sum_{k=1}^3 \lambda_k \mathbf{n}_k \otimes \mathbf{n}_k$, where $\lambda_k \neq 0$ and $\{\mathbf{n}_1, \mathbf{n}_2, \mathbf{n}_3\}$ are orthonormal vector basis.

Proof. As $\sum_{i=1}^n \omega_i \mathbf{u}_i \otimes \mathbf{u}_i$ is a real-symmetric matrix, it has spectral representation (from spectral theorem) as $\sum_{i=1}^n \omega_i \mathbf{u}_i \otimes \mathbf{u}_i = \sum_{k=1}^3 \lambda_k \mathbf{n}_k \otimes \mathbf{n}_k$, where $\{\mathbf{n}_k\}$ are three orthonormal vectors. We have to show that each $\lambda_k \neq 0$. When $\{\mathbf{u}_i\}$ spanning \mathbb{R}^3 , we have $\mathbf{u}_i = (\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3) \mathbf{q}_i = \mathbf{E} \cdot \mathbf{q}_i$, where $\mathbf{E} = (\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3)$ are chosen orthonormal vector basis and \mathbf{q}_i are the coordinate of \mathbf{u}_i at such basis. We then have: $\sum_{i=1}^n \omega_i \mathbf{u}_i \otimes \mathbf{u}_i = \mathbf{E} (\sum_{i=1}^n \omega_i \mathbf{q}_i \otimes \mathbf{q}_i) \mathbf{E}^T$.

Now, consider if exists $\lambda_k = 0$. Then we have $(\sum_{k=1}^3 \lambda_k \mathbf{n}_k \otimes \mathbf{n}_k) \cdot \mathbf{n}_k = \mathbf{0}$, $\mathbf{n}_k \neq \mathbf{0}$ and $(\sum_{i=1}^n \omega_i \mathbf{u}_i \otimes \mathbf{u}_i) \cdot \mathbf{n}_k = \mathbf{0}$. Expresses \mathbf{n}_k as $\mathbf{n}_k = \mathbf{E} \cdot \mathbf{q}_k$ in vector basis \mathbf{E} and its coordinate \mathbf{q}_k . The above equation becomes:

$$\mathbf{E} \left(\sum_{i=1}^n \omega_i \mathbf{q}_i \otimes \mathbf{q}_i \right) \mathbf{E}^T \cdot \mathbf{E} \cdot \mathbf{q}_k = \mathbf{0}, \quad (5)$$

which gives: $\mathbf{E} (\sum_{i=1}^n \omega_i \mathbf{q}_i \otimes \mathbf{q}_i) \cdot \mathbf{q}_k = \mathbf{0}$. Because \mathbf{E} is a vector basis matrix of \mathbb{R}^3 , the only solution exists that $\mathbf{E} \cdot \mathbf{q} = \mathbf{0}$ as $\mathbf{q} = \mathbf{0}$, so that we must have $(\sum_{i=1}^n \omega_i \mathbf{q}_i \otimes \mathbf{q}_i) \cdot \mathbf{q}_k = \mathbf{0}$.

Now, we denote $\mathbf{N} = \sum_{i=1}^n \omega_i \mathbf{q}_i \otimes \mathbf{q}_i$ and show that $\mathbf{N} \cdot \mathbf{q}_k = \mathbf{0}$, if and only $\mathbf{q}_k = \mathbf{0}$ or $\forall \mathbf{q}_i, \mathbf{q}_i \cdot \mathbf{q}_k = 0$. Firstly, the sufficient condition is simple: if $\mathbf{q}_k = \mathbf{0}$, then it is obviously to show $\mathbf{N} \cdot \mathbf{q}_k = \mathbf{0}$ and if $\forall \mathbf{q}_i, \mathbf{q}_i \cdot \mathbf{q}_k = 0$, then $\mathbf{N} \cdot \mathbf{q}_k = \mathbf{0}$.

Now we want to show the necessary condition: if $\mathbf{N} \cdot \mathbf{q}_k = \mathbf{0}$, then $\mathbf{q}_k = \mathbf{0}$ or $\forall \mathbf{q}_i, \mathbf{q}_i \cdot \mathbf{q}_k = 0$. Assume $\mathbf{q}_k \neq \mathbf{0}$ and there exists some $\{\mathbf{q}_i\}$ s.t. $\mathbf{q}_i \cdot \mathbf{q}_k \neq 0$, then we at most choose three orthonormal vectors, s.t. $\mathbf{q}_1 \cdot \mathbf{q}_k \neq 0, \mathbf{q}_2 \cdot \mathbf{q}_k \neq 0, \mathbf{q}_3 \cdot \mathbf{q}_k \neq 0$ to represent \mathbf{q}_i . Denoted $\mathbf{q}_i = \alpha_i \mathbf{q}_1 + \beta_i \mathbf{q}_2 + \gamma_i \mathbf{q}_3$, we then have:

$$\begin{aligned} \mathbf{N} \cdot \mathbf{q}_k &= \left(\sum_{i=1}^n \omega_i (\alpha_i \mathbf{q}_1 + \beta_i \mathbf{q}_2 + \gamma_i \mathbf{q}_3) \otimes (\alpha_i \mathbf{q}_1 + \beta_i \mathbf{q}_2 + \gamma_i \mathbf{q}_3) \right) \cdot \mathbf{q}_k \\ &= \sum_{i=1}^n \omega_i \alpha_i (\mathbf{q}_i \cdot \mathbf{q}_k) \mathbf{q}_1 + \sum_{i=1}^n \omega_i \beta_i (\mathbf{q}_i \cdot \mathbf{q}_k) \mathbf{q}_2 + \sum_{i=1}^n \omega_i \gamma_i (\mathbf{q}_i \cdot \mathbf{q}_k) \mathbf{q}_3 = \mathbf{0}, \end{aligned} \quad (6)$$

which gives three parameter equations of $\mathbf{q}_1, \mathbf{q}_2, \mathbf{q}_3$ correspondingly to be zero, since $\mathbf{q}_1, \mathbf{q}_2, \mathbf{q}_3$ are chosen orthonormal.

We look at the equations (also remind $\omega_i > 0$):

$$\begin{cases} \sum_{i=1}^n \omega_i \alpha_i (\mathbf{q}_i \cdot \mathbf{q}_k) = 0 \\ \sum_{i=1}^n \omega_i \beta_i (\mathbf{q}_i \cdot \mathbf{q}_k) = 0 \\ \sum_{i=1}^n \omega_i \gamma_i (\mathbf{q}_i \cdot \mathbf{q}_k) = 0. \end{cases} \quad (7)$$

Case 1. If $\{\mathbf{q}_i\}$ lays on $\text{span}(\{\mathbf{q}_1\})$ (this chosen of one basis does not limit general cases), then $\alpha_i \neq 0, \beta_i = 0, \gamma_i = 0$. As $\sum_{i=1}^n \omega_i \alpha_i (\mathbf{q}_i \cdot \mathbf{q}_k) = \sum_{i=1}^n \omega_i \alpha_i^2 (\mathbf{q}_1 \cdot \mathbf{q}_k) = 0$, which gives $\mathbf{q}_1 \cdot \mathbf{q}_k = 0$. Then we have $\mathbf{q}_i \cdot \mathbf{q}_k = 0$.

Case 2. If $\{\mathbf{q}_i\}$ lays on $\text{span}(\{\mathbf{q}_1, \mathbf{q}_2\})$ (this chosen of two bases does not limit general cases), then $\gamma_i = 0$. Then we will have :

$$\begin{cases} \sum_{i=1}^n \omega_i \alpha_i^2 (\mathbf{q}_1 \cdot \mathbf{q}_k) + \sum_{i=1}^n \omega_i \alpha_i \beta_i (\mathbf{q}_2 \cdot \mathbf{q}_k) = 0 \\ \sum_{i=1}^n \omega_i \alpha_i \beta_i (\mathbf{q}_1 \cdot \mathbf{q}_k) + \sum_{i=1}^n \omega_i \beta_i^2 (\mathbf{q}_2 \cdot \mathbf{q}_k) = 0. \end{cases} \quad (8)$$

If $\sum_{i=1}^n \omega_i \alpha_i \beta_i = 0$, then we have $\mathbf{q}_1 \cdot \mathbf{q}_k = 0, \mathbf{q}_2 \cdot \mathbf{q}_k = 0$, since $\{\mathbf{q}_i\}$ lays on $\text{span}(\{\mathbf{q}_1, \mathbf{q}_2\})$ gives that $\exists \mathbf{q}_i, \mathbf{q}_j$, s.t. $\alpha_i, \beta_j \neq 0$ and thus $\sum_i \omega_i \alpha_i^2 > 0, \sum_i \omega_i \beta_i^2 > 0$. If $\sum_i \omega_i \alpha_i \beta_i \neq 0$, the above equation can be deduced as $((\sum_i \omega_i \alpha_i \beta_i)^2 - (\sum_i \omega_i \alpha_i^2)(\sum_i \omega_i \beta_i^2)) \cdot (\mathbf{q}_2 \cdot \mathbf{q}_k) = 0$. As the first term is a classical equation

which gives $((\sum_i \omega_i \alpha_i \beta_i)^2 - (\sum_i \omega_i \alpha_i^2)(\sum_i \omega_i \beta_i^2)) = 0$ if and only if $\forall i, \alpha_i = \beta_i$, then we have $\{\mathbf{q}_i\}$ lays on space spanning by a single vector $\mathbf{q}_1 + \mathbf{q}_2$, which is discussed on Case 1. Then we have the equation has to stand for other cases with arbitrary $\omega_i, \alpha_i, \beta_i$ gives $\mathbf{q}_1 \cdot \mathbf{q}_k = 0, \mathbf{q}_2 \cdot \mathbf{q}_k = 0$. Then we have $\mathbf{q}_i \cdot \mathbf{q}_k = 0$.

Case 3. If \mathbf{q}_i lays on $\text{span}(\{\mathbf{q}_1, \mathbf{q}_2, \mathbf{q}_3\})$, the above equation set with three equations has unique solution, since \mathbf{q}_k has three unknowns, then the number of equations is the same as the number of unknowns. It is simple to verify that this unique solution is $\mathbf{q}_k = \mathbf{0}$. Then we have $\mathbf{q}_i \cdot \mathbf{q}_k = 0$.

Now we have already verified that $\mathbf{N} \cdot \mathbf{q}_k = \mathbf{0}$ if and only if $\mathbf{q}_k = \mathbf{0}$ or $\forall \mathbf{q}_i, \mathbf{q}_i \cdot \mathbf{q}_k = 0$. As $\{\mathbf{u}_i\}$ span \mathbb{R}^3 , then its coordinate $\{\mathbf{q}_i\}$ on vector basis \mathbf{E} also span \mathbb{R}^3 , which means that $\forall \mathbf{q}_i, \mathbf{q}_i \cdot \mathbf{q}_k = 0$ gives $\mathbf{q}_k = \mathbf{0}$.

Above Eq. (5) we assume that there exist $\lambda_k = 0$. By showing that $\mathbf{q}_k = \mathbf{0}$ uniquely, which means that the solution of $(\sum_i^n \omega_i \mathbf{u}_i \otimes \mathbf{u}_i) \mathbf{n} = \mathbf{0}$ only have zero trivial solution $\mathbf{n} = \mathbf{0}$, so that $\mathbf{U} = \sum_{k=1}^3 \lambda_k \mathbf{n}_k \otimes \mathbf{n}_k$ is of full rank. Then the assumption of existing $\lambda_k = 0$ is not stand. So we have $\lambda_k \neq 0, \forall k = 1, 2, 3$. \square

Theorem 1 For a 3×3 second-order tensor of $\mathbf{U} = \lambda_1 \mathbf{n}_1 \otimes \mathbf{n}_1 + \lambda_2 \mathbf{n}_2 \otimes \mathbf{n}_2 + \lambda_3 \mathbf{n}_3 \otimes \mathbf{n}_3$, where $\mathbf{n}_1, \mathbf{n}_2, \mathbf{n}_3$ are orthonormal vector basis and $\lambda_k \neq 0, k = 1, 2, 3$. Then the tensor is non-singular, with inverse $\mathbf{U}^{-1} = \lambda_1^{-1} \mathbf{n}_1 \otimes \mathbf{n}_1 + \lambda_2^{-1} \mathbf{n}_2 \otimes \mathbf{n}_2 + \lambda_3^{-1} \mathbf{n}_3 \otimes \mathbf{n}_3$.

Proof. Yet simple to verify $\mathbf{U}\mathbf{U}^{-1} = \mathbf{U}^{-1}\mathbf{U} = \mathbf{I}$, where \mathbf{I} is the 3×3 identity matrix. \square

From Lem. (1), Lem. (2) and Theo. (1), it is shown that as if the vertex neighborhood remains to be of full dimension, i.e., either by extending the neighborhood with \mathbf{x}_i^\perp or the neighborhood originally keeping full dimension (e.g., of some complex configuration embeded in \mathbb{R}^3), then our extended shape tensor will always keep its non-singularity (from Lem. (2) and Theo. (1)), so that its inverse \mathbf{K}_i^{-1} exists. The full-dimensional (extended) neighborhood has been a sufficient condition. So even if original shape tensor has already kept non-singularity, adding the $\mathbf{x}_i^\perp \otimes \mathbf{x}_i^\perp$ part will not change this.

In the next section, we show that, such extension will also guarantee rest and affine transform stability.

1.2 Rest and Rigid-body deformation stability for the Extended Deformation Gradient

Extended deformation gradient

$$\mathbf{F}_i = \left(\sum_j \omega_{ij} (\mathbf{y}_j - \mathbf{y}_i) (\mathbf{x}_j - \mathbf{x}_i)^T + \mathbf{y}_i^\perp \otimes \mathbf{x}_i^\perp \right) \mathbf{K}_i^{-1}. \quad (9)$$

We shown that, if the deformation configuration $\{\mathbf{y}\} = \{\mathbf{x}\}$, which should give $\mathbf{F}_i = \mathbf{I}$, and the affine transform configuration $\{\mathbf{y}\} = \mathbf{R}\{\mathbf{x}\} + \mathbf{v}$, where \mathbf{R} is a rotational matrix and \mathbf{v} a movement, which should gives $\mathbf{F}_i = \mathbf{R}$.

When $\{\mathbf{y}\} = \{\mathbf{x}\}$, $\mathbf{D}_i := \left(\sum_j \omega_{ij} (\mathbf{y}_j - \mathbf{y}_i) (\mathbf{x}_j - \mathbf{x}_i)^T + \mathbf{y}_i^\perp \otimes \mathbf{x}_i^\perp \right)$ is exactly the same as \mathbf{K}_i , so that we will simply get $\mathbf{F}_i = \mathbf{D}_i \mathbf{K}_i^{-1} = \mathbf{K}_i \mathbf{K}_i^{-1} = \mathbf{I}$.

When $\{\mathbf{y}\} = \mathbf{R}\{\mathbf{x}\} + \mathbf{v}$, $\mathbf{y}_j - \mathbf{y}_i = \mathbf{R}(\mathbf{x}_j - \mathbf{x}_i)$. Now we look at the extended term $\mathbf{y}_i^\perp \otimes \mathbf{x}_i^\perp$. As

$$\mathbf{x}_i^\perp = \xi_{\text{norm}} \left(\frac{\sum_J \theta_J \mathbf{n}_J}{\sum_J \theta_J} \right), \mathbf{y}_i^\perp = \xi_{\text{norm}} \left(\frac{\sum_J \theta'_J \mathbf{n}'_J}{\sum_J \theta'_J} \right), \quad (10)$$

where ξ represents the thickness of the membrane measured from surface to the mid-surface; J is the index of all neighboring triangles; θ_J is inner angle of triangle J with apex of vertex i ; \mathbf{n}_J is the triangle normal, and $\text{norm}(\cdot)$ is used to normalize a vector. The superscript ' for θ and \mathbf{n} is used to denote quantities in the deformed configuration. As θ_J remains to be unchanged through affine transform, the only condition we have to verify is that $\forall J, \mathbf{n}'_J = \mathbf{R}\mathbf{n}_J$. Then we have: $\mathbf{n}'_J = \text{norm}((\mathbf{y}_J^2 - \mathbf{y}_J^0) \times (\mathbf{y}_J^1 - \mathbf{y}_J^0)) = \text{norm}((\mathbf{R}(\mathbf{x}_J^2 - \mathbf{x}_J^0)) \times (\mathbf{R}(\mathbf{x}_J^1 - \mathbf{x}_J^0))) = \text{norm}(\mathbf{R}((\mathbf{x}_J^2 - \mathbf{x}_J^0) \times (\mathbf{x}_J^1 - \mathbf{x}_J^0))) = \mathbf{R} \text{norm}((\mathbf{x}_J^2 - \mathbf{x}_J^0) \times (\mathbf{x}_J^1 - \mathbf{x}_J^0))$ from basic vector analysis rule, where the superscript $\cdot^{\{1,2,3\}}$ denotes three related vertex in triangle.

Together we have

$$\mathbf{D}_i = \left(\sum_j \omega_{ij} (\mathbf{y}_j - \mathbf{y}_i) (\mathbf{x}_j - \mathbf{x}_i)^T + \mathbf{y}_i^\perp \otimes \mathbf{x}_i^\perp \right) = \left(\sum_j \omega_{ij} \mathbf{R}(\mathbf{x}_j - \mathbf{x}_i) (\mathbf{x}_j - \mathbf{x}_i)^T + \mathbf{R}\mathbf{x}_i^\perp \otimes \mathbf{x}_i^\perp \right) = \mathbf{R}\mathbf{K}_i.$$

Then we have $\mathbf{F}_i = \mathbf{D}_i \mathbf{K}_i^{-1} = \mathbf{R}$.

2 Handle Contact Through Local Projection

2.1 Step Length of Gradient Method

In this section we will analysis how to choose step length λ to guarantee descending of the objective function in local step, i.e.,

$$\mathcal{B}(\mathbf{z}^m + \lambda \mathbf{g}) < \mathcal{B}(\mathbf{z}^m) \quad (11)$$

where

$$\mathcal{B}(\mathbf{z}) := \left(\sum_i \frac{1}{2} \|\mathbf{z}_i - \mathbf{y}_i^k\|_2^2 + \mu \sum_c B_c(\mathbf{z}) \right), \quad B(d(\mathbf{z})) = \begin{cases} -(d - \hat{d})^2 \log\left(\frac{d}{\hat{d}}\right), & 0 < d < \hat{d} \\ 0, & d \geq \hat{d} \end{cases}. \quad (12)$$

For such barrier energy function modeling the contact, since it is at least second-order smooth when actived (when $0 < d < \hat{d}$), the first-order partial derivative and second-order partial derivative are calculated and can be refered from [1](Supplement A) as:

$$\frac{\partial B(d)}{\partial d} := B'(d) = - \left(2(d - \hat{d}) \log\left(\frac{d}{\hat{d}}\right) + \frac{(d - \hat{d})^2}{d} \right) < 0, \quad (13)$$

and

$$\frac{\partial^2 B(d)}{\partial d^2} := B''(d) = - \left(2 \log\left(\frac{d}{\hat{d}}\right) + (d - \hat{d}) \frac{\hat{d} + 3d}{d^2} \right) > 0, \quad (14)$$

for $0 < d < \hat{d}$. After the mapping of $\hat{d}_{IJ} = d_{IJ} - \xi$, show on Fig. (9) and Eq. (51), we have both d , $-B' = |B'|$ and B'' get bounded, as $\lfloor d \rfloor = \varepsilon \xi$, $\lceil |B'(d)| \rceil = 2(\hat{d} - \varepsilon \xi) \log\left(\frac{\hat{d}}{\varepsilon \xi}\right) + \frac{(\varepsilon \xi - \hat{d})^2}{\varepsilon \xi}$ and $\lceil B''(d) \rceil = 2 \log\left(\frac{\hat{d}}{\varepsilon \xi}\right) + (\hat{d} - \varepsilon \xi) \frac{\hat{d} + 3\varepsilon \xi}{(\varepsilon \xi)^2}$, where $\lfloor \cdot \rfloor$ denotes the lower bound and $\lceil \cdot \rceil$ denotes the upper bound.

We start to show that, the above condition makes the objective function \mathcal{B} has a *Lipschitz continuous gradient* with constant L as the Lipschitz parameter, so that a fixed step with constant step length λ can be chosen as if $0 < \lambda \leq 1/L$ to guarantee convergency. We would like to provide generally definition and theorem and then look inside to the previous contact problem.

Definition 1 *L-smooth Function.*

A function f has a Lipschitz continuous gradient with constant $L > 0$, when

$$\|\nabla f(\mathbf{x}) - \nabla f(\mathbf{y})\|_2 \leq L \|\mathbf{x} - \mathbf{y}\|_2, \quad \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n.$$

We denote this kind function as $\mathcal{C}_L^1(\mathbb{R}^n)$ and L -smooth function.

Theorem 2 For any $f \in \mathcal{C}_L^1(\mathbb{R}^n)$, we have:

$$f(\mathbf{y}) \leq f(\mathbf{x}) + \nabla f(\mathbf{x})^T (\mathbf{y} - \mathbf{x}) + \frac{L}{2} \|\mathbf{y} - \mathbf{x}\|_2^2, \quad \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n.$$

Since the proof of Theo. (2) can be found in any Mathematical Analysis textbook, we would like to show it directly.

Theorem 3 *Fixed step convergency for gradient method.*

For gradient method and a L -smooth objective function $f \in \mathcal{C}_L^1(\mathbb{R}^n)$, GM guaranteed to converge for any $0 < \lambda \leq 1/L$, i.e.,

$$\begin{aligned} \mathbf{x}^+ &= \mathbf{x} - \lambda \nabla f(\mathbf{x}), \\ f(\mathbf{x}^+) &\leq f(\mathbf{x}), \quad \forall \lambda, 0 < \lambda \leq 1/L. \end{aligned}$$

Since the proof of Theo. (3) can also be found in any optimization textbook, here we would like to provide a yet simple proof.

Proof. Consider a single iteration

$$\mathbf{x}^+ = \mathbf{x} - \lambda \nabla f(\mathbf{x}).$$

Then for any $\mathbf{x} \in \mathbb{R}^n$ and $\lambda \in (0, 1/L]$, we have:

$$\begin{aligned} f(\mathbf{x}^+) &\leq f(\mathbf{x}) + \nabla f(\mathbf{x})^T(\mathbf{x}^+ - \mathbf{x}) + \frac{L}{2} \|\mathbf{x}^+ - \mathbf{x}\|_2^2 \quad (\text{Theo. (2)}) \\ &= f(\mathbf{x}) + \nabla f(\mathbf{x})^T(-\lambda \nabla f(\mathbf{x})) + \frac{L\lambda^2}{2} \|\nabla f(\mathbf{x})\|_2^2 \\ &\leq f(\mathbf{x}) - \lambda \|\nabla f(\mathbf{x})\|_2^2 + \frac{\lambda}{2} \|\nabla f(\mathbf{x})\|_2^2 \\ &= f(\mathbf{x}) - \frac{\lambda}{2} \|\nabla f(\mathbf{x})\|_2^2 \\ &\leq f(\mathbf{x}). \end{aligned}$$

□

Following the above theorem, the question remains to be, how to find L constant for our objective function \mathcal{B} . As \mathcal{B} is second-order differentiable, then we have $L = \max(|\lambda_i|) = \|\nabla^2 \mathcal{B}\|_2$, $\forall \lambda_i$ are the eigenvalue of $\nabla^2 \mathcal{B}$, deduced from the inequation $\|\nabla \mathcal{B}(\mathbf{y}) - \nabla \mathcal{B}(\mathbf{x})\|_2 \leq \|\nabla^2 \mathcal{B}\|_2 \cdot \|\mathbf{y} - \mathbf{x}\|_2$.

Look inside to the Hessian of the objective function \mathcal{B} , since the gradient method is done for each vertex \mathbf{z}_i , we only need to estimate a maximal absolute eigenvalue of Hessian derivative for each \mathbf{z}_i , then we can get an upper bound of the step length.

As

$$\begin{aligned} \nabla_{\mathbf{z}_i} \mathcal{B} &= -\mathbf{g}_i \\ &= -\mu \sum_I \alpha_I^i \left[\frac{(d_{IJ} - \hat{d})^2}{d_{IJ}} + 2(d_{IJ} - \hat{d}) \log \left(\frac{d_{IJ}}{\hat{d}} \right) \right] \text{norm}(\mathbf{d}_{IJ}) - \mathbf{y}_i^k + \mathbf{z}_i \\ &= \mu \sum_I \alpha_I^i B'(d_{IJ}) \text{norm}(\mathbf{d}_{IJ}) - \mathbf{y}_i^k + \mathbf{z}_i, \end{aligned} \quad (15)$$

where α_I^i represents the barycentric coordinate for vertex i in triangle I , and $\mathbf{d}_{IJ} = \alpha_I^i \mathbf{z}_i + \alpha_I^j \mathbf{z}_j + \alpha_I^k \mathbf{z}_k - \alpha_a^a \mathbf{z}_a - \alpha_b^b \mathbf{z}_b - \alpha_c^c \mathbf{z}_c$, where (i, j, k) and (a, b, c) are the corresponding vertex indexes of triangle I and J . Here we explicitly describe the proximal directional distance \mathbf{d}_{IJ} because we will then make derivative of \mathbf{z}_i on it. The proximal distance length $d_{IJ} = \|\mathbf{d}_{IJ}\|_2$, and $\text{norm}(\mathbf{d}_{IJ}) = \mathbf{d}_{IJ}/d_{IJ}$. Then we have

$$\nabla_{\mathbf{z}_i}^2 \mathcal{B} = \mu \sum_I \alpha_I^i \left(B''(d_{IJ}) \frac{\partial d_{IJ}}{\partial \mathbf{z}_i} \otimes \text{norm}(\mathbf{d}_{IJ}) + B'(d_{IJ}) \frac{\partial \text{norm}(\mathbf{d}_{IJ})}{\partial \mathbf{z}_i} \right) + \mathbf{I}, \quad (16)$$

where \mathbf{I} is an identity matrix of dimension $\mathbf{z}_i \times \text{dimension } \mathbf{z}_i$, i.e., 3×3 , and \otimes denotes the Kronecker product.

As we have (which can be found in [2] or simply make deduction)

$$\begin{aligned} \frac{\partial d_{IJ}}{\partial \mathbf{z}_i} &= \frac{\partial d_{IJ}}{\partial \mathbf{d}_{IJ}} \cdot \frac{\partial \mathbf{d}_{IJ}}{\partial \mathbf{z}_i} = \alpha_I^i \text{norm}(\mathbf{d}_{IJ}), \\ \frac{\partial \text{norm}(\mathbf{d}_{IJ})}{\partial \mathbf{z}_i} &= \frac{\partial \text{norm}(\mathbf{d}_{IJ})}{\partial \mathbf{d}_{IJ}} \cdot \frac{\partial \mathbf{d}_{IJ}}{\partial \mathbf{z}_i} = \frac{\alpha_I^i}{d_{IJ}} \left(\mathbf{I} - \text{norm}(\mathbf{d}_{IJ}) \otimes \text{norm}(\mathbf{d}_{IJ}) \right), \end{aligned}$$

which gives:

$$\begin{aligned} \nabla_{\mathbf{z}_i}^2 \mathcal{B} &= \mu \sum_I \alpha_I^i \left(B''(d_{IJ}) \frac{\partial d_{IJ}}{\partial \mathbf{z}_i} \otimes \text{norm}(\mathbf{d}_{IJ}) + B'(d_{IJ}) \frac{\partial \text{norm}(\mathbf{d}_{IJ})}{\partial \mathbf{z}_i} \right) + \mathbf{I} \\ &= \left(\mu \sum_I (\alpha_I^i)^2 \frac{B'(d_{IJ})}{d_{IJ}} + 1 \right) \mathbf{I} + \mu \sum_I \left[(\alpha_I^i)^2 \left(B''(d_{IJ}) + \frac{B'(d_{IJ})}{d_{IJ}} \right) \text{norm}(\mathbf{d}_{IJ}) \otimes \text{norm}(\mathbf{d}_{IJ}) \right] \\ &= FT + ST, \end{aligned} \quad (17)$$

where FT denotes the 'first term' and ST the 'second term'.

Since the d_{IJ} to active barrier energy is actually transformed to $\hat{d}_{IJ} = d_{IJ} - \xi$ when considering thickness modeling, such linear transform does change derivative and hessian, only changing the B', B'' terms. Then this d will go through our mapping function and get $\lfloor d \rfloor = \varepsilon \xi$ as minimal separation. In such mapping, we literally consider d as a calculated coefficient free of \mathbf{z}_i (so that there will be no $\frac{\partial f(d)}{\partial d}$ exist in the Chain Rule).

Although this analytical representation of Eq. (17) can be calculated to get $L = \|\nabla_{\mathbf{z}_i}^2 \mathcal{B}\|_2$, such process will be unnecessarily time costing. We choose to estimate an upper bound of $\|\nabla_{\mathbf{z}_i}^2 \mathcal{B}\|_2$. As $\|\nabla_{\mathbf{z}_i}^2 \mathcal{B}\|_2 \leq \|FT\|_2 + \|ST\|_2$, we will look inside to these terms each by each.

From FT , we have:

$$\begin{aligned}
\|FT\|_2 &\leq \left| \mu \sum_I (\alpha_I^i)^2 \frac{B'}{d} + 1 \right| \cdot \|\mathbf{I}\|_2 \\
&\leq \left| \mu \sum_I (\alpha_I^i)^2 \frac{B'}{d} + 1 \right| \\
&< \left| \mu \sum_I (\alpha_I^i)^2 \frac{B'}{d} \right| + |1| \\
&\leq \left| \mu \sum_I \alpha_I^i \frac{B'}{d} \right| + \left| \frac{\hat{d}}{\lfloor d \rfloor} \right| \quad (\text{since } 0 \leq \alpha_I^i \leq 1 \text{ and } 0 < d < \hat{d}) \\
&\leq \mu \alpha_0 \left| \frac{B'}{\lfloor d \rfloor} \right| + \frac{\hat{d}}{\lfloor d \rfloor} \quad (\text{here denotes } \alpha_0 := \left\lceil \sum_I \alpha_I^i \right\rceil) \\
&\leq \frac{1}{\lfloor d \rfloor} \left(\mu \alpha_0 \lceil |B'(d)| \rceil + \hat{d} \right) =: \lceil \|FT\|_2 \rceil.
\end{aligned} \tag{18}$$

Although the $\|ST\|_2$ can also get an upper bound $\mu \alpha_0 \left(\frac{\lceil |B'| \rceil}{\lfloor d \rfloor} + \lceil B'' \rceil \right)$ through similar deduction, however, this upper bound possess too large expansion, since '=' can only be reached when all $\alpha_I^i = 1$ and \mathbf{d}_{IJ} are paralleled, because of the existence of $\|\sum_I \text{norm}(\mathbf{d}_{IJ}) \otimes \text{norm}(\mathbf{d}_{IJ})\|_2$. A too large expansion will make the estimation of Lipschitz constant L expanded and yet make the step length unnecessarily get smaller, so that the convergence process will be unnecessarily slow.

We will simply abandon the $\|ST\|_2$ term, since $\|FT\|_2$ has already provided enough expansion in estimation of L and thus will give a enough step length upper bound. So that the estimation of L become $L \leq \|\nabla_{\mathbf{z}_i}^2 \mathcal{B}\|_2 \leq \lceil \|FT\|_2 \rceil$. Then we have $1/L \geq 1/\lceil \|FT\|_2 \rceil$.

From Theo. (3) and Eq. (18), then we have:

$$\begin{aligned}
0 &< \lambda \leq 1/L \\
\text{which means } \lambda &\leq \lfloor 1/L \rfloor = \frac{\varepsilon \xi}{\mu \alpha_0 \lceil |B'(d)| \rceil + \hat{d}}.
\end{aligned} \tag{19}$$

This condition has a more intuitive illustration, i.e., requiring the maximum position change at each iteration of each vertex no greater than minimal separation $\varepsilon \xi$, i.e., $\max_i \|\lambda \mathbf{g}_i\|_2 \leq \varepsilon \xi$.

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