

# Seamless Cloth Simulation with High Geometric Fidelity

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**Abstract.** Despite a large volume of studies, a current particle dynamic system results plausible deformation of draping, folding, wrinkling, stretching, etc., but realistic fold patterns generated from body-pose based variations and motion ill pose the problem. In this paper, a new approach based on continuum mechanic is proposed, which

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## 1 Introduction

## 2

### 2.1 Deformation Gradient

Let  $\kappa$  be a reference configuration and  $\Gamma$  be an arbitrary configuration of body  $\mathcal{B}$ . Then the mapping

$$\mathbf{x} = \Gamma_\kappa(\mathbf{X}, t) \quad (1)$$

is called the *deformation* of body  $\mathcal{B}$  from  $\kappa$  to  $\Gamma$ .

The *deformation gradient*  $\mathbf{F}$  of  $\Gamma$  relative to  $\kappa$ , is defined by

$$\mathbf{F} = \nabla_{\mathbf{X}} \Gamma_\kappa = \frac{\partial \mathbf{x}}{\partial \mathbf{X}} = \begin{bmatrix} \frac{\partial x}{\partial X} & \frac{\partial x}{\partial Y} & \frac{\partial x}{\partial Z} \\ \frac{\partial y}{\partial X} & \frac{\partial y}{\partial Y} & \frac{\partial y}{\partial Z} \\ \frac{\partial z}{\partial X} & \frac{\partial z}{\partial Y} & \frac{\partial z}{\partial Z} \end{bmatrix}, \quad (2)$$

where the coordinates of  $\mathbf{X}$  are called the *template coordinates* and the coordinates of  $\mathbf{x}$  are called the *reference coordinates*. Moreover, the Eq. (2) is a measure of local deformation of the body. Hence the mapping  $\Gamma_\kappa$  is one-to-one and onto,  $\mathbf{F}$  is non-singular.

Applied by *decomposition theorem* on the deformation gradient [?], there exist two positive definite symmetric tensor  $\mathbf{U}$  and  $\mathbf{V}$ , and an orthogonal tensor  $\mathbf{R}$ , uniquely determined by  $\mathbf{F}$ , such that

$$\mathbf{F} = \mathbf{R}\mathbf{U} = \mathbf{V}\mathbf{R}, \quad (3)$$

where the  $\mathbf{U}$  and  $\mathbf{V}$  denote the *right* and *left stretch tensor*, and  $\mathbf{R}$  denotes the local *rotation tensor*. As a consequence of Eq. (3), we could obtain the pure stretches by applying

$$\mathbf{U}^2 = \mathbf{F}^T \mathbf{F} = \mathbf{C} \quad \text{and} \quad \mathbf{V}^2 = \mathbf{F} \mathbf{F}^T = \mathbf{B}, \quad (4)$$

where the  $\mathbf{C}$  and  $\mathbf{B}$  are called *right* and *left Cauchy-Green strain tensor*.

Since  $\mathbf{C}$  can be diagonalized as

$$\mathbf{C} = \lambda_1^2 \mathbf{e}_1 \mathbf{e}_1^T + \lambda_2^2 \mathbf{e}_2 \mathbf{e}_2^T + \lambda_3^2 \mathbf{e}_3 \mathbf{e}_3^T, \quad (5)$$

we could calculate the  $\mathbf{U}$  by applying

$$\mathbf{U} = \lambda_1 \mathbf{e}_1 \mathbf{e}_1^T + \lambda_2 \mathbf{e}_2 \mathbf{e}_2^T + \lambda_3 \mathbf{e}_3 \mathbf{e}_3^T. \quad (6)$$

The  $\lambda_1 < \lambda_2 < \lambda_3$  are real, positive eigenvalues and the corresponding  $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$  are unit eigenvectors, defining main stretch or shrinkage directions [?].

Following by [?], the *wrinkle vector field*  $\mathbf{v}$  describes the rate of shrinkage per unit of length wherever shrinkage occurs and its direction is orthogonal to the main shrinkage direction, which is given by

$$\mathbf{v} = \max(1 - \lambda_1, 0) \mathbf{e}_2. \quad (7)$$

## 2.2 Velocity Gradient

Whereas the deformation gradient measures the local deformation, the *spacial velocity gradient*  $\mathbf{L}$  describes the rate of deformation, given by

$$\mathbf{L} = \text{grad} \mathbf{v} = \dot{\mathbf{F}} \mathbf{F}^{-1}, \quad (8)$$

where  $\mathbf{v}$  denotes the *velocity* of the material point  $\mathbf{X}$ , and  $\dot{\mathbf{F}}$  denotes the material time derivative of deformation gradient  $\mathbf{F}$ .

Analogically, we apply the polar decomposition on the spacial velocity gradient  $\mathbf{L}$  [?], and obtain

$$\mathbf{L} = \mathbf{D} + \mathbf{W}. \quad (9)$$

Thus, the tensor  $\mathbf{L}$  could be decomposed into its symmetric part  $\mathbf{D}$  and skew-symmetric part  $\mathbf{W}$

$$\mathbf{D} = \frac{1}{2}(\mathbf{L} + \mathbf{L}^T), \quad (10)$$

$$\mathbf{W} = \frac{1}{2}(\mathbf{L} - \mathbf{L}^T). \quad (11)$$

where  $\mathbf{D}$  is called the *rate of strain tensor*, and  $\mathbf{W}$  is called the *rate of rotation tensor*.

Assuming  $d\mathbf{x}$  is a material line element in the current configuration, the rate of change of its length  $\dot{\epsilon}_{ii}$  and angle  $\dot{\gamma}_{ij}$  is measured by means of  $\mathbf{D}$  [?].

$$\dot{\epsilon}_{ii} = \mathbf{e}_i \mathbf{D} \mathbf{e}_i = D_{ii} \quad (12)$$

$$\dot{\gamma}_{ij} = 2\mathbf{e}_i \mathbf{D} \mathbf{e}_j = D_{ij} \quad (13)$$

where  $D_{ij}$  are the  $ij$ -entries of  $\mathbf{D}$ . Additionally, if  $\mathbf{D}$  has the same base  $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$  as in the eq.(6), then

$$D_{11} = \frac{\dot{\lambda}_1}{\lambda_1} \quad (14)$$

$$D_{22} = \frac{\dot{\lambda}_2}{\lambda_3} \quad (15)$$

$$D_{33} = \frac{\dot{\lambda}_2}{\lambda_3} \quad (16)$$

Given the deformation gradients at the beginning  $F_t$  and end  $F_{t+\Delta t}$  of a time step, we use *forward difference* [?] to calculate the derivation of deformation gradient and take the inverse of the later one.

### 2.3 Cross-covariance Matrix

Let  $\Phi$  denote the points set in the template frame, and  $\Psi$  denote the points set in the reference frame. Therefore, each set of points can be described as

$$\Phi = \begin{bmatrix} X_1 & Y_1 & Z_1 \\ X_2 & Y_2 & Z_2 \\ \vdots & \vdots & \vdots \\ X_N & Y_N & Z_N \end{bmatrix} \quad \text{and} \quad \Psi = \begin{bmatrix} x_1 & y_1 & z_1 \\ x_2 & y_2 & z_2 \\ \vdots & \vdots & \vdots \\ x_N & y_N & z_N \end{bmatrix}, \quad (17)$$

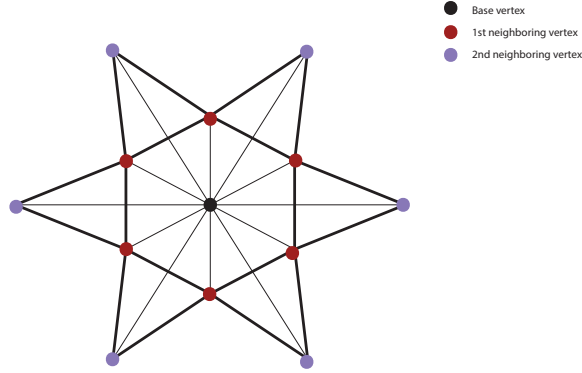
where  $N$  is the number of the points in the points set.

A point  $\bar{\mathbf{j}}$  is called the *neighboring point* of point  $\mathbf{j}$ , when  $\bar{\mathbf{j}}$  and  $\mathbf{j}$  are connected by a same edge. If only the data set without any further information is given, we shall use the *kd-tree* [?] as a data structure to search for the neighboring points in no structural points set.

Let  $\bar{\Phi} \subset \Phi$  define the points set of the neighboring points of  $\mathbf{X}_i$  in the template frame, and  $\bar{\Psi} \subset \Psi$  define the points set of the neighboring points of  $\mathbf{x}_i$  in the reference frame. Therefore, we could construct a  $M \times 3$  matrix  $\mathbf{P}_i$  for each point  $\mathbf{X}_i \in \Phi$ , as well as a matrix  $\mathbf{Q}_i$  for each point  $\mathbf{x}_i \in \Psi$ ,

$$\mathbf{P}_i = \begin{bmatrix} X_1 - X_i & Y_1 - Y_i & Z_1 - Z_i \\ X_2 - X_i & Y_2 - Y_i & Z_2 - Z_i \\ \vdots & \vdots & \vdots \\ X_M - X_i & Y_M - Y_i & Z_M - Z_i \end{bmatrix} \quad \text{and} \quad \mathbf{Q}_i = \begin{bmatrix} x_1 - x_i & y_1 - y_i & z_1 - z_i \\ x_2 - x_i & y_2 - y_i & z_2 - z_i \\ \vdots & \vdots & \vdots \\ x_M - x_i & y_M - y_i & z_M - z_i \end{bmatrix}, \quad (18)$$

where  $\mathbf{X}_1, \mathbf{X}_2, \dots, \mathbf{X}_M \in \bar{\Phi}$  and  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_M \in \bar{\Psi}$ , and  $M$  is the number of the points in the neighboring points set. The matrices  $\mathbf{P}_i$  and  $\mathbf{Q}_i$  represent the distribution of local shapes near the point  $\mathbf{X}_i$  and  $\mathbf{x}_i$ , see Fig. 1.



**Fig. 1.** The neighboring vertices of one base vertex in a structural points set.

Proceeding to define

$$\mathbf{H} = \mathbf{P}^T \mathbf{Q}. \quad (19)$$

$\mathbf{H}$  is called the *cross-covariance matrix*, which is a measure of similarity of  $\mathbf{P}$  and  $\mathbf{Q}$  [?].

The matrix  $\mathbf{H}$  is derived from the *orthogonal Procrustes problem*[?]. It is defined as the least-squares problem of transforming a given matrix  $\mathbf{P}$  into a given matrix  $\mathbf{Q}$  by an orthogonal transformation matrix  $\mathbf{R}$ , such that the sums of squares of the residual matrix  $\mathbf{E} = \mathbf{\Omega P} - \mathbf{Q}$  is a minimum

$$\mathbf{R} = \arg \min_{\mathbf{\Omega}} \|\mathbf{\Omega P} - \mathbf{Q}\|_F \quad \text{subject to} \quad \mathbf{\Omega}^T \mathbf{\Omega} = \mathbf{I}, \quad (20)$$

where  $\|\cdot\|_F$  is the Frobenius norm. Moreover, it can be shown[?] that this problem is equivalent to find the nearest orthogonal matrix  $\mathbf{R}$  to a given matrix  $\mathbf{H} = \mathbf{P}^T \mathbf{Q}$ , which most closely maps  $\mathbf{P}$  to  $\mathbf{Q}$ . Thus, we could use  $\mathbf{H}$  to approximate the deformation gradient  $\mathbf{F}$  in Eq. (2)

$$\tilde{\mathbf{F}} = \mathbf{H}. \quad (21)$$

Applied by Eq. (3) (4) (6), we could obtain the approximated stretch tensor  $\tilde{\mathbf{U}}$  and its eigenvalues  $\tilde{\lambda}_1, \tilde{\lambda}_2, \tilde{\lambda}_3$ .

### 3 Animation of Wrinkle Deformation Field

#### 3.1 Structural Points Sets Using 1st Neighboring Points

#### 3.2 Structural Points Sets Using 2nd Neighboring Points

#### 3.3 Non-structural Points Sets Using Binary Search Tree

### 4 Conclusion