Macrophage trajectories smoothing by evolving curves

PhD candidate : Giulia Lupi Supervisor : Prof. Karol Mikula Team members: Seol Ah Park

Slovak University of Technology

Discrete Duality Finite Volume Method and Applications 17-21 October 2022

- E - F

Motivation



PhD candidate : Giulia Lupi Supervisor : Prc Sn

Smoothing by evolving curves

1 Mathematical Model

< ∃ →

Presentation Overview

1 Mathematical Model

2 Numerical Discretization

Space Discretization Inflow-implicit/Outflow-explicit scheme Overall time discretization Attracting term

< ∃ >

-

Presentation Overview

Mathematical Model

2 Numerical Discretization

Space Discretization Inflow-implicit/Outflow-explicit scheme Overall time discretization Attracting term

Numerical Experiments

 Influence of the attracting term
 Macrophages trajectories
 Smoothed Velocity Estimation

∃ >



PhD candidate : Giulia Lupi Supervisor : Prc Smoothing by evolving curves

- < ≣ >

-



• $\delta > 0$ constant, *k* curvature, **N** unit normal vector



- $\delta > 0$ constant, *k* curvature, **N** unit normal vector
- $\lambda > 0$ constant,**x**₀ initial condition,(**x**₀ **x**) smooth function, \cdot scalar product



- $\delta > 0$ constant, *k* curvature, **N** unit normal vector
- $\lambda > 0$ constant,**x**₀ initial condition,(**x**₀ **x**) smooth function, \cdot scalar product
- α tangential velocity, **T** unit tangent vector

PhD candidate : Giulia Lupi Supervisor : Prc Smoothing by evolving curves 4/25

Attracting term



where

$$\chi_{u} = \{ \mathbf{x} \in \mathbb{R}^{2} : \mathbf{v} = \mathbf{x}_{0}(q) - \mathbf{x}(u), q \in [0, 1] \}$$
(4)

< ∃ >

포 사 문

Intrinsic form of the PDE

Consider $\mathbf{x}_t = -\delta k \mathbf{N} + \lambda w \mathbf{N} + \alpha \mathbf{T}$, where $w = [(\mathbf{x}_0 - \mathbf{x}) \cdot \mathbf{N}]$.

▲御 ▶ ▲ 唐 ▶ ▲ 唐 ▶

э

Consider $\mathbf{x}_t = -\delta k \mathbf{N} + \lambda w \mathbf{N} + \alpha \mathbf{T}$, where $w = [(\mathbf{x}_0 - \mathbf{x}) \cdot \mathbf{N}]$. The curve is discretized to a set of points: $\mathbf{x}_0, \mathbf{x}_1, ..., \mathbf{x}_{n+1}$.

▲御 ▶ ▲ 憲 ▶

Consider $\mathbf{x}_t = -\delta k \mathbf{N} + \lambda w \mathbf{N} + \alpha \mathbf{T}$, where $w = [(\mathbf{x}_0 - \mathbf{x}) \cdot \mathbf{N}]$. The curve is discretized to a set of points: $\mathbf{x}_0, \mathbf{x}_1, ..., \mathbf{x}_{n+1}$.



• Let *s* be the unit arc-length parametrization.

Consider $\mathbf{x}_t = -\delta k \mathbf{N} + \lambda w \mathbf{N} + \alpha \mathbf{T}$, where $w = [(\mathbf{x}_0 - \mathbf{x}) \cdot \mathbf{N}]$. The curve is discretized to a set of points: $\mathbf{x}_0, \mathbf{x}_1, ..., \mathbf{x}_{n+1}$.



- Let *s* be the unit arc-length parametrization.
- Define $\mathbf{T} = \mathbf{x}_s$ and $\mathbf{N} = \mathbf{x}_s^{\perp}$ such that $\mathbf{T} \wedge \mathbf{N} = -1$.

Consider $\mathbf{x}_t = -\delta k \mathbf{N} + \lambda w \mathbf{N} + \alpha \mathbf{T}$, where $w = [(\mathbf{x}_0 - \mathbf{x}) \cdot \mathbf{N}]$. The curve is discretized to a set of points: $\mathbf{x}_0, \mathbf{x}_1, ..., \mathbf{x}_{n+1}$.



- Let *s* be the unit arc-length parametrization.
- Define $\mathbf{T} = \mathbf{x}_s$ and $\mathbf{N} = \mathbf{x}_s^{\perp}$ such that $\mathbf{T} \wedge \mathbf{N} = -1$.
- From the Frenet-Serret formulas we get $\mathbf{T}_s = -k\mathbf{N}$. Then

$$-k\mathbf{N}=\mathbf{T}_{s}=(\mathbf{x}_{s})_{s}=\mathbf{x}_{ss}.$$

Consider $\mathbf{x}_t = -\delta k \mathbf{N} + \lambda w \mathbf{N} + \alpha \mathbf{T}$, where $w = [(\mathbf{x}_0 - \mathbf{x}) \cdot \mathbf{N}]$. The curve is discretized to a set of points: $\mathbf{x}_0, \mathbf{x}_1, ..., \mathbf{x}_{n+1}$.



- Let *s* be the unit arc-length parametrization.
- Define $\mathbf{T} = \mathbf{x}_s$ and $\mathbf{N} = \mathbf{x}_s^{\perp}$ such that $\mathbf{T} \wedge \mathbf{N} = -1$.
- From the Frenet-Serret formulas we get $\mathbf{T}_s = -k\mathbf{N}$. Then

 $-k\mathbf{N} = \mathbf{T}_s = (\mathbf{x}_s)_s = \mathbf{x}_{ss}.$ We obtain the form of the so-called intrinsic partial differential equation: $\mathbf{x}_t = \delta \mathbf{x}_{ss} + \alpha \mathbf{x}_s + \lambda w \mathbf{x}_s^{\perp}.$

Choice of the tangential velocity

Consider the local and the global length of the evolving curve

$$g = |\mathbf{x}_u|, \quad L = \int_0^1 g du, \qquad \frac{g}{L} \approx \frac{\frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{h}}{L} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{Lh} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{\frac{L}{h+1}}$$

御 と く き と く き と

where $h = \frac{1}{n+1}$ and n + 1 is the number of segments.

Choice of the tangential velocity

Consider the local and the global length of the evolving curve

$$g = |\mathbf{x}_u|, \quad L = \int_0^1 g du, \qquad \frac{g}{L} \approx \frac{\frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{h}}{L} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{Lh} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{\frac{L}{n+1}}$$

where $h = \frac{1}{n+1}$ and n+1 is the number of segments. We want uniformly distributed points, then $\frac{g}{T} \rightarrow 1$

▲御▶ ▲漫▶ ▲漫▶

Choice of the tangential velocity

Consider the local and the global length of the evolving curve

$$g = |\mathbf{x}_u|, \quad L = \int_0^1 g du, \qquad \frac{g}{L} \approx \frac{\frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{h}}{L} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{Lh} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{\frac{L}{n+1}}$$

where $h = \frac{1}{n+1}$ and n+1 is the number of segments. We want uniformly distributed points, then $\frac{g}{L} \rightarrow 1$ Using the Frenet-Serret formulas we obtain

$$(\frac{g}{L})_t = \frac{g}{L}(k\beta + \alpha_s - \langle k\beta \rangle_{\Gamma})$$

where $\langle k\beta \rangle_{\Gamma} = \frac{1}{L} \int_{\Gamma} k\beta$.

伺 ト イ ヨ ト イ ヨ ト

Choice of the tangential velocity

Consider the local and the global length of the evolving curve

$$g = |\mathbf{x}_u|, \quad L = \int_0^1 g du, \qquad \frac{g}{L} \approx \frac{\frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{h}}{L} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{Lh} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{\frac{L}{n+1}}$$

where $h = \frac{1}{n+1}$ and n+1 is the number of segments. We want uniformly distributed points, then $\frac{g}{L} \rightarrow 1$ Using the Frenet-Serret formulas we obtain

$$(\frac{g}{L})_t = \frac{g}{L}(k\beta + \alpha_s - \langle k\beta \rangle_{\Gamma})$$

where $\langle k\beta \rangle_{\Gamma} = \frac{1}{L} \int_{\Gamma} k\beta$. Define

$$(rac{g}{L})_t = \omega(1-rac{g}{L}) \qquad \omega > 0$$

・ 戸 ト ・ ヨ ト ・ ヨ ト

Choice of the tangential velocity

Consider the local and the global length of the evolving curve

$$g = |\mathbf{x}_u|, \quad L = \int_0^1 g du, \qquad \frac{g}{L} \approx \frac{\frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{h}}{L} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{Lh} = \frac{|\mathbf{x}_i - \mathbf{x}_{i-1}|}{\frac{L}{n+1}}$$

where $h = \frac{1}{n+1}$ and n+1 is the number of segments. We want uniformly distributed points, then $\frac{g}{L} \rightarrow 1$ Using the Frenet-Serret formulas we obtain

$$(\frac{g}{L})_t = \frac{g}{L}(k\beta + \alpha_s - \langle k\beta \rangle_{\Gamma})$$

where $\langle k\beta \rangle_{\Gamma} = \frac{1}{L} \int_{\Gamma} k\beta$. Define

$$(rac{g}{L})_t = \omega(1-rac{g}{L}) \qquad \omega > 0$$

We obtain

$$\alpha_{s} = \langle \boldsymbol{k}\beta \rangle_{\Gamma} - \boldsymbol{k}\beta + \omega(\frac{L}{g} - 1)$$

Flowing finite volume method

Let us consider the intrinsic form of the PDE

$$\mathbf{x}_t - \alpha \mathbf{x}_s = \delta \mathbf{x}_{ss} + \lambda w \mathbf{x}_s^{\perp}$$



PhD candidate : Giulia Lupi Supervisor : Prc

Integrating over the finite volume \mathbf{p}_i we obtain

$$\int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} \mathbf{x}_t ds - \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} \alpha \mathbf{x}_s ds = \delta \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} \mathbf{x}_{ss} ds + \lambda \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} w \mathbf{x}_s^{\perp} ds \quad (5)$$

< ∃ >

The values α , *w* are considered to be constant over the finite volume and will be indicated as α_i , *w*_{*i*}.

Integrating over the finite volume \mathbf{p}_i we obtain

$$\int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} \mathbf{x}_t ds - \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} \alpha \mathbf{x}_s ds = \delta \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} \mathbf{x}_{ss} ds + \lambda \int_{x_{i-\frac{1}{2}}}^{x_{i+\frac{1}{2}}} w \mathbf{x}_s^{\perp} ds \quad (5)$$

The values α , *w* are considered to be constant over the finite volume and will be indicated as α_i , *w*_{*i*}.

Using the Newton-Leibniz formula and approximating \mathbf{x}_s by a finite difference we obtain

$$\frac{h_{i}+h_{i+1}}{2}(\mathbf{x}_{i})_{t} + \frac{\alpha_{i}}{2}(\mathbf{x}_{i}-\mathbf{x}_{i+1}) - \frac{\alpha_{i}}{2}(\mathbf{x}_{i}-\mathbf{x}_{i-1}) = \\ = \delta(\frac{\mathbf{x}_{i+1}-\mathbf{x}_{i}}{h_{i+1}} - \frac{\mathbf{x}_{i}-\mathbf{x}_{i-1}}{h_{i}}) + \lambda w_{i}(\frac{\mathbf{x}_{i+1}-\mathbf{x}_{i-1}}{2})^{\perp}$$
(6)

ヘロア 人間 アメヨア 小田 アー

Inflow-implicit/Outflow-explicit scheme

$$\mathbf{x}_t + \alpha \mathbf{x}_s = \mathbf{0}$$

PhD candidate : Giulia Lupi Supervisor : Prc Smoothing by evolving curves 10

ヘロト 人間 ト 人 思 ト 人 思 ト

э

Inflow-implicit/Outflow-explicit scheme

$$\mathbf{x}_t + \alpha \mathbf{x}_s = \mathbf{0}$$



ヘロト 人間 ト 人 思 ト 人 思 ト

æ

Inflow-implicit/Outflow-explicit scheme

$$\mathbf{x}_t + \alpha \mathbf{x}_s = \mathbf{0}$$



æ

Inflow-implicit/Outflow-explicit scheme

$$\frac{h_{i}+h_{i+1}}{2}(\mathbf{x}_{i})_{t} + \frac{\alpha_{i}}{2}(\mathbf{x}_{i}-\mathbf{x}_{i+1}) - \frac{\alpha_{i}}{2}(\mathbf{x}_{i}-\mathbf{x}_{i-1}) = \\ = \delta_{i}(\frac{\mathbf{x}_{i+1}-\mathbf{x}_{i}}{h_{i+1}} - \frac{\mathbf{x}_{i}-\mathbf{x}_{i-1}}{h_{i}}) + \lambda w_{i}(\frac{\mathbf{x}_{i+1}-\mathbf{x}_{i-1}}{2})^{\perp}$$
(7)

ヘロアス 留マス ほどん ほどう

3

Inflow-implicit/Outflow-explicit scheme

$$\frac{h_{i}+h_{i+1}}{2}(\mathbf{x}_{i})_{t} + \frac{\alpha_{i}}{2}(\mathbf{x}_{i}-\mathbf{x}_{i+1}) - \frac{\alpha_{i}}{2}(\mathbf{x}_{i}-\mathbf{x}_{i-1}) =$$

$$= \delta_{i}(\frac{\mathbf{x}_{i+1}-\mathbf{x}_{i}}{h_{i+1}} - \frac{\mathbf{x}_{i}-\mathbf{x}_{i-1}}{h_{i}}) + \lambda w_{i}(\frac{\mathbf{x}_{i+1}-\mathbf{x}_{i-1}}{2})^{\perp}$$

$$\mathbf{x}_{t} - \alpha \mathbf{x}_{s} = \delta \mathbf{x}_{ss} + \lambda w \mathbf{x}_{s}^{\perp}$$
(7)

ヘロト 人間 とくほ とくほとう

э

1

Inflow-implicit/Outflow-explicit scheme

$$\frac{h_{i} + h_{i+1}}{2}(\mathbf{x}_{i})_{t} + \frac{\alpha_{i}}{2}(\mathbf{x}_{i} - \mathbf{x}_{i+1}) - \frac{\alpha_{i}}{2}(\mathbf{x}_{i} - \mathbf{x}_{i-1}) =$$

$$= \delta_{i}(\frac{\mathbf{x}_{i+1} - \mathbf{x}_{i}}{h_{i+1}} - \frac{\mathbf{x}_{i} - \mathbf{x}_{i-1}}{h_{i}}) + \lambda w_{i}(\frac{\mathbf{x}_{i+1} - \mathbf{x}_{i-1}}{2})^{\perp}$$

$$\mathbf{x}_{t} - \alpha \mathbf{x}_{s} = \delta \mathbf{x}_{ss} + \lambda w \mathbf{x}_{s}^{\perp}$$

$$b_{i-\frac{1}{2}}^{in} = max(-\alpha_{i}, 0), b_{i-\frac{1}{2}}^{out} = min(-\alpha_{i}, 0)$$

$$b_{i+\frac{1}{2}}^{in} = max(\alpha_{i}, 0), b_{i+\frac{1}{2}}^{out} = min(\alpha_{i}, 0)$$
(8)

PhD candidate : Giulia Lupi Supervisor : Prc Smoothing by evolving curves

ヘロト 人間 ト 人 思 ト 人 思 ト

э

Inflow-implicit/Outflow-explicit scheme

$$\frac{h_{i} + h_{i+1}}{2} (\mathbf{x}_{i})_{t} + \frac{\alpha_{i}}{2} (\mathbf{x}_{i} - \mathbf{x}_{i+1}) - \frac{\alpha_{i}}{2} (\mathbf{x}_{i} - \mathbf{x}_{i-1}) = \\ = \delta_{i} (\frac{\mathbf{x}_{i+1} - \mathbf{x}_{i}}{h_{i+1}} - \frac{\mathbf{x}_{i} - \mathbf{x}_{i-1}}{h_{i}}) + \lambda w_{i} (\frac{\mathbf{x}_{i+1} - \mathbf{x}_{i-1}}{2})^{\perp}$$

$$\mathbf{x}_{t} - \alpha \mathbf{x}_{s} = \delta \mathbf{x}_{ss} + \lambda w \mathbf{x}_{s}^{\perp}$$

$$b_{i-\frac{1}{2}}^{in} = max(-\alpha_{i}, 0), b_{i-\frac{1}{2}}^{out} = min(-\alpha_{i}, 0) \\ b_{i+\frac{1}{2}}^{in} = max(\alpha_{i}, 0), b_{i+\frac{1}{2}}^{out} = min(\alpha_{i}, 0)$$

$$\frac{h_{i} + h_{i+1}}{2} (\mathbf{x}_{i})_{t} + \frac{1}{2} (b_{i+\frac{1}{2}}^{in} + b_{i+\frac{1}{2}}^{out}) (\mathbf{x}_{i} - \mathbf{x}_{i+1}) + \frac{1}{2} (b_{i-\frac{1}{2}}^{in} + b_{i-\frac{1}{2}}^{out}) (\mathbf{x}_{i} - \mathbf{x}_{i-1}) \\ = \delta (\frac{\mathbf{x}_{i+1} - \mathbf{x}_{i}}{h_{i+1}} - \frac{\mathbf{x}_{i} - \mathbf{x}_{i-1}}{h_{i}}) + \lambda w_{i} (\frac{\mathbf{x}_{i+1} - \mathbf{x}_{i-1}}{2})^{\perp}$$
(7)

PhD candidate : Giulia Lupi Supervisor : Prc

・ロト ・ 聞 ト ・ 思 ト ・ 思 ト

ъ

Let *m* be the time step index and τ the length of the discrete time step.

• time derivative: finite difference $\mathbf{x}_t = \frac{\mathbf{x}_i^{m+1} - \mathbf{x}_i^m}{\tau}$

Let *m* be the time step index and τ the length of the discrete time step.

- time derivative: finite difference $\mathbf{x}_t = \frac{\mathbf{x}_i^{m+1} \mathbf{x}_i^m}{\tau}$
- unknowns in the inflow part of the advection term implicitly

Let *m* be the time step index and τ the length of the discrete time step.

- time derivative: finite difference $\mathbf{x}_t = \frac{\mathbf{x}_i^{m+1} \mathbf{x}_i^m}{\tau}$
- unknowns in the inflow part of the advection term implicitly
- unknowns in the outflow part of the advection term explicitly

< ⊒ >

Let *m* be the time step index and τ the length of the discrete time step.

- time derivative: finite difference $\mathbf{x}_t = \frac{\mathbf{x}_i^{m+1} \mathbf{x}_i^m}{\tau}$
- unknowns in the inflow part of the advection term implicitly
- unknowns in the outflow part of the advection term explicitly

< ∃ >

• diffusion term implicitly

Let *m* be the time step index and τ the length of the discrete time step.

- time derivative: finite difference $\mathbf{x}_t = \frac{\mathbf{x}_i^{m+1} \mathbf{x}_i^m}{\tau}$
- unknowns in the inflow part of the advection term implicitly
- unknowns in the outflow part of the advection term explicitly

I ≥ ▶

- diffusion term implicitly
- attracting term explicitly

Let *m* be the time step index and τ the length of the discrete time step.

- time derivative: finite difference $\mathbf{x}_t = \frac{\mathbf{x}_i^{m+1} \mathbf{x}_i^m}{\tau}$
- unknowns in the inflow part of the advection term implicitly
- unknowns in the outflow part of the advection term explicitly
- diffusion term implicitly
- attracting term explicitly

We obtain



Attracting term

Fix \mathbf{x}_i^m and for j = 1, ..., n + 1

• Consider the line *r* passing through the points $\mathbf{x}_{i-1}^0, \mathbf{x}_i^0$,



-∢ ⊒ ▶

Attracting term

Fix \mathbf{x}_i^m and for j = 1, ..., n + 1

- Consider the line *r* passing through the points $\mathbf{x}_{i-1}^0, \mathbf{x}_i^0$,
- Find the line *s* such that: $s \perp r$ and $\mathbf{x}_i^m \in s$. Find $\mathbf{q} \in r \cap s$.



Attracting term

Fix \mathbf{x}_i^m and for j = 1, ..., n + 1

- Consider the line *r* passing through the points $\mathbf{x}_{i-1}^0, \mathbf{x}_i^0$,
- Find the line *s* such that: $s \perp r$ and $\mathbf{x}_i^m \in s$. Find $\mathbf{q} \in r \cap s$.
- Consider $d = d(\mathbf{q}, \mathbf{x}_i^m)$. If $\mathbf{q} \in [\mathbf{x}_{j-1}^0, \mathbf{x}_j^0]$ and d < minD then: $\mathbf{x}^0 - \mathbf{x}_i^m = \mathbf{q} - \mathbf{x}_i^m$



Influence of the attracting term

Comparison



Figure: Comparison of the evolution of the initial curve (red) for δ fixed and different values of λ after 400 time steps.Results are shown for $\lambda = 0$ (blue), $\lambda = 1$ (light blue), $\lambda = 5$ (pink) and $\lambda = 10$ (green)

Macrophages trajectories

PhD candidate : Giulia Lupi Supervisor : Prc Smoothing by evolving curves 157.

э

Macrophages trajectories

PhD candidate : Giulia Lupi Supervisor : Prc Smoothing by evolving curves 16/2

э

Macrophages trajectories



ヘロト 人間 ト 人 思 ト 人 思 ト

æ

We considered the mean Hausdorff distance between two discrete curves \mathcal{A}, \mathcal{B} defined as

$$\bar{\delta}_{H}(\mathcal{A},\mathcal{B}) = \frac{1}{n} \sum_{i=1}^{n} \min_{\mathbf{b} \in B} d(\mathbf{a}_{i},\mathbf{b}),$$

$$\bar{\delta}_{H}(\mathcal{B},\mathcal{A}) = \frac{1}{n} \sum_{i=1}^{n} \min_{\mathbf{a} \in \mathcal{A}} d(\mathbf{b}_{i},\mathbf{a}),$$

$$\bar{d}_{H}(\mathcal{A},\mathcal{B}) = \frac{\bar{\delta}_{H}(\mathcal{A},\mathcal{B}) + \bar{\delta}_{H}(\mathcal{B},\mathcal{A})}{2},$$

(11)

where $A = \{a_0, ..., a_{n+1}\}$, $B = \{b_0, ..., b_{n+1}\}$ are discrete sets and $A = \{a_1, ..., a_{n+1}\}$, $B = \{b_1, ..., b_{n+1}\}$ are sets of segments

Macrophage trajectories



PhD candidate : Giulia Lupi Supervisor : Prc Smoothing by evolving curves 19/25

< □ ▶ < 团 ▶ < 厘 ▶ 19/25 э

Macrophage trajectories



ヘロマ ヘロマ 小田 ア

э

э

Smoothed Velocity Estimation

The formula for the new real-time is given by

$$T_i^{n+1} = T_i^n + \frac{\alpha_i^n \tau}{L^{n+1}} \tag{12}$$

・ 同 ト ・ ヨ ト ・ ヨ ト

$$T_{i}^{n+1} = T_{i}^{n} + \frac{\alpha_{i}^{n}\tau}{L^{n+1}}$$
(12)

• • = • • = •

• T_i^{n+1} and T_i^n are the new real-time and the real-time in the previous time step

$$T_{i}^{n+1} = T_{i}^{n} + \frac{\alpha_{i}^{n}\tau}{L^{n+1}}$$
(12)

A B M A B M

- T_i^{n+1} and T_i^n are the new real-time and the real-time in the previous time step
- α_i^n is the tangential velocity

$$T_{i}^{n+1} = T_{i}^{n} + \frac{\alpha_{i}^{n}\tau}{L^{n+1}}$$
(12)

A B F A B F

- T_i^{n+1} and T_i^n are the new real-time and the real-time in the previous time step
- α_i^n is the tangential velocity
- au the time step

$$T_{i}^{n+1} = T_{i}^{n} + \frac{\alpha_{i}^{n}\tau}{L^{n+1}}$$
(12)

- T_i^{n+1} and T_i^n are the new real-time and the real-time in the previous time step
- α_i^n is the tangential velocity
- *τ* the time step
- L^{n+1} is the total length of the curve in the current time step

• • = • • = •

Finally, we calculated the velocity using the central difference scheme

$$V(x,y) = \left(\frac{\mathbf{x}_{1}^{i+1} - \mathbf{x}_{1}^{i-1}}{T_{i+1} - T_{i-1}}, \frac{\mathbf{x}_{2}^{i+1} - \mathbf{x}_{2}^{i-1}}{T_{i+1} - T_{i-1}}\right)$$
(13)

- 4 同 ト 4 ヨ ト

Time and Velocity Estimation

Velocity Estimation



Э

Time and Velocity Estimation

Velocity Estimation



◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のへで

S T U

SLOVAK UNIVERSITY OF TECHNOLOGY IN BRATISLAVA

- Seol Ah Park
- Karol Mikula





- Mai Nguyen-Chi
- Georges Lutfalla
- Tamara Sipka

Thank You