

Cage-based deformations for implicit surfaces

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at University Paris-Est Marne-la-Vallée
in Internship supervised by Loïc Barthe[†] and Pascal Romon[‡]

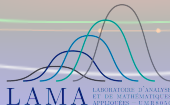
July 08, 2019



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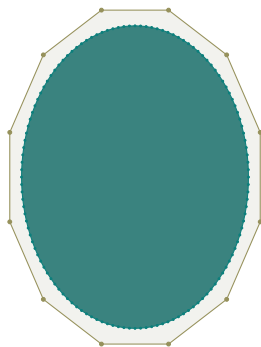


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Topics of the presentation :

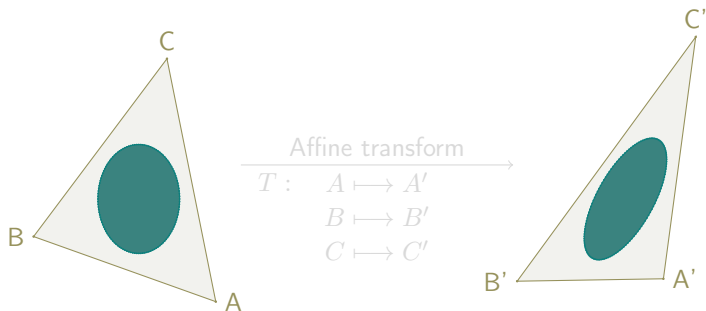
- Introduction to cage-based deformations.
 - Cages and interest as deformation.
 - Generalized barycentric coordinates in cages for deformation.
- How to deform an implicit surface with a cage.
 - Inverse position problem.
 - State of the art : Free-Form Deformations special type of cages and implicit surface deformation.
 - Our methods and flexible solving architecture.
 - Cage auto-intersection field-operator solving trials.
- Future work and interest for Implicit skinning.

Cage on surface



- Bounding simplification of a discrete **surface**.
- Define **control positions**.
- Allow **smooth deformations** of the surface.

Barycentric coordinates and affine transform



$$P = \alpha A + \beta B + \gamma C$$

$$\alpha + \beta + \gamma = 1$$

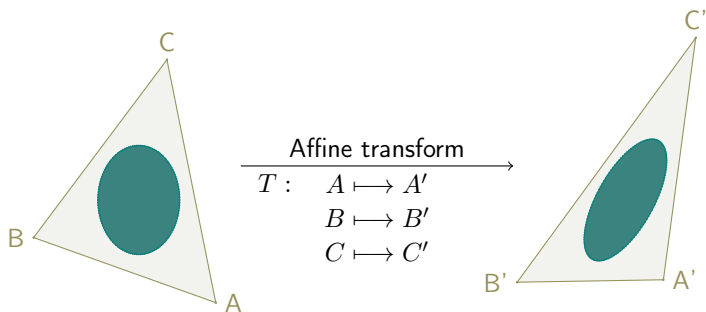
$$\Rightarrow \alpha \vec{PA} + \beta \vec{PB} + \gamma \vec{PC} = 0$$

$$Q = T(P) = T(\alpha A + \beta B + \gamma C)$$

$$Q = \alpha T(A) + \beta T(B) + \gamma T(C)$$

$$Q = \alpha A' + \beta B' + \gamma C'$$

Barycentric coordinates and affine transform



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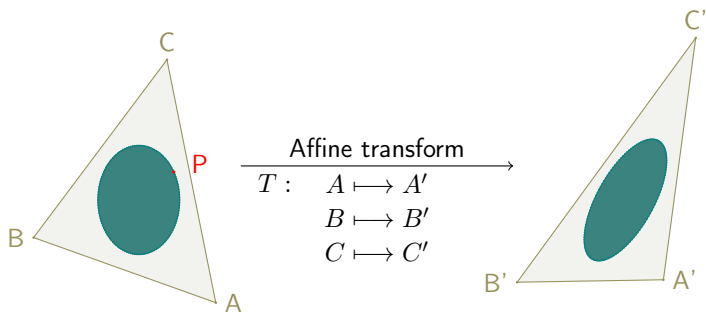
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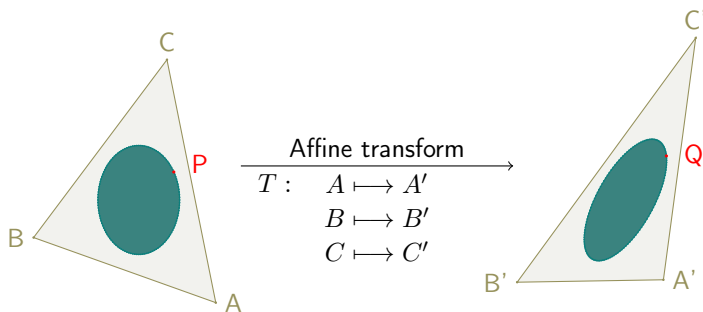
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Barycentric coordinates and affine transform



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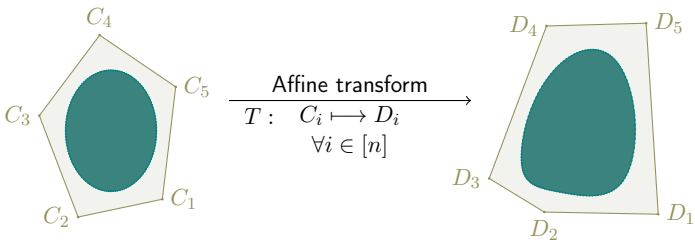
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Generalized barycentric coordinates to cages



$$P = \sum_{i \in [n]} \alpha_i(P) C_i$$

$$\sum_{i \in [n]} \alpha_i(P) = 1$$

Binding step :

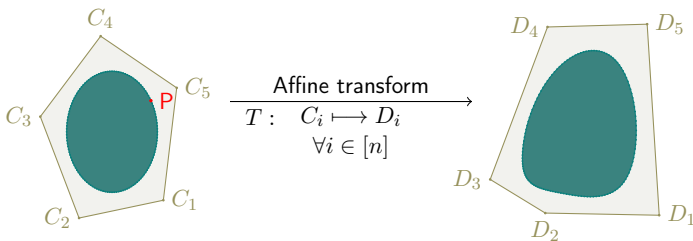
compute weights $(\alpha_i(P))_{i \in [n]}$

$$Q = T(P) = T(\sum_{i \in [n]} \alpha_i(P) C_i)$$

$$Q = \sum_{i \in [n]} \alpha_i(P) T(C_i)$$

$$Q = \sum_{i \in [n]} \alpha_i(P) D_i$$

Generalized barycentric coordinates to cages



$$P = \sum_{i \in [n]} \alpha_i(P) C_i$$

$$\sum_{i \in [n]} \alpha_i(P) = 1$$

Binding step :

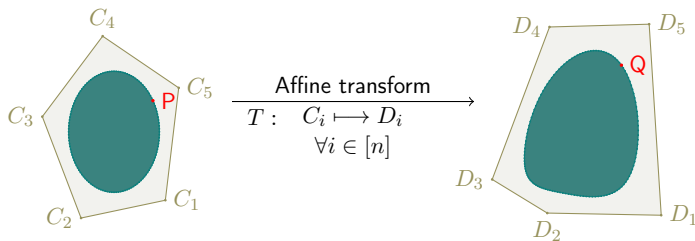
compute weights $(\alpha_i(P))_{i \in [n]}$

$$Q = T(P) = T\left(\sum_{i \in [n]} \alpha_i(P) C_i\right)$$

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Generalized barycentric coordinates to cages



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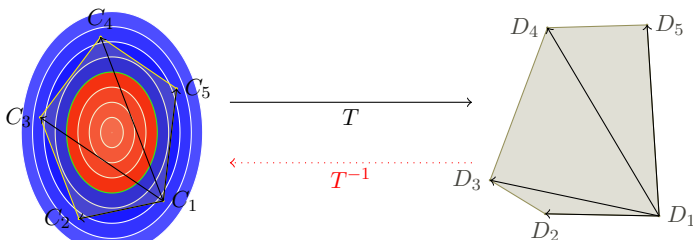
$$Q = \sum_{i \in [n]} \alpha_i(P) T(C_i)$$

$$Q = \sum_{i \in [n]} \alpha_i(P) D_i$$

$$\begin{aligned} Q &= \sum_{i \in [n]} \alpha_i(P)(C_i + \vec{u}_i) \\ Q &= P + \sum_{i \in [n]} \alpha_i(P)\vec{u}_i \\ P &= Q - \sum_{i \in [n]} \alpha_i(P)\vec{u}_i \end{aligned}$$

7 / 54

Inverse position problem



$$g(Q) = f(P)$$

$$g(Q) = f(T^{-1}(Q))$$

$$Q = \sum_{i \in [n]} \alpha_i(P) D_i$$

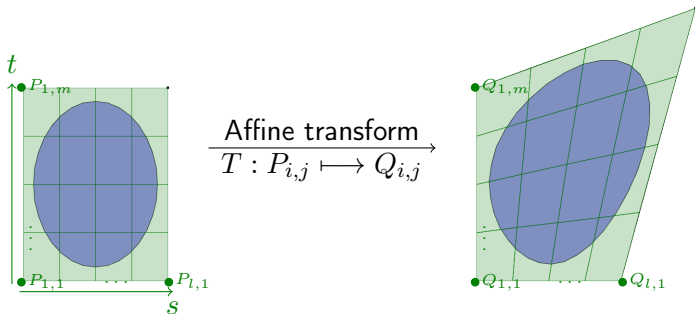
$$Q = \sum_{i \in [n]} \alpha_i(P) (C_i + \vec{u}_i)$$

$$Q = P + \sum_{i \in [n]} \alpha_i(P) \vec{u}_i$$

$$\mathbf{P} = Q - \sum_{i \in [n]} \alpha_i(\mathbf{P}) \vec{u}_i$$

$\{\overrightarrow{C_1 C_i}\}_{i \in [n] \setminus \{1\}}$ linearly dependent.

Free-Form-Deformations

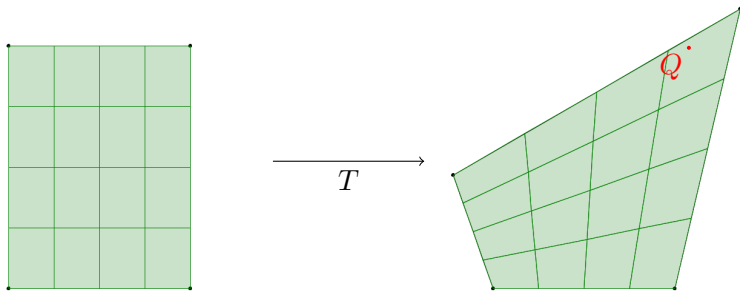


(s, t) coordinates in $(P_{1,1}, \overrightarrow{P_{1,1}P_{l,1}}, \overrightarrow{P_{1,1}P_{1,m}})$.

$P_{i,j}$ and $Q_{i,j}$ bilinear interpolation of the parallelepipeds.

$$T(s, t) = \sum_{(i, j) \in [l] \times [m]} \binom{l}{i} \binom{m}{j} s^i (1-s)^{l-i} t^j (1-t)^{m-j} T(P_{i, j})$$

Free-Form-Deformations and inverse problem solving



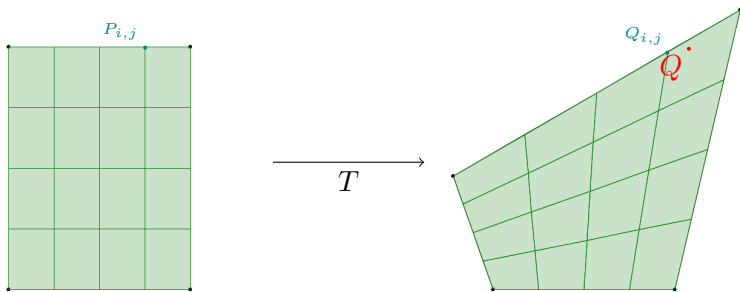
Get nearest $Q_{i,j}$,
hence $P_{i,j}$ equivalent.
Solve $T(s, t) - Q = 0$
using Newton method.

$$J_T(s, t) = \left(\frac{\partial}{\partial s} T(s, t), \frac{\partial}{\partial t} T(s, t) \right)$$

$$X_0 = \left(\frac{i}{l}, \frac{j}{m} \right)$$

$$X_{n+1} = X_n - J_T^{-1}(X_n) (T(X_n) - Q)$$

Free-Form-Deformations and inverse problem solving



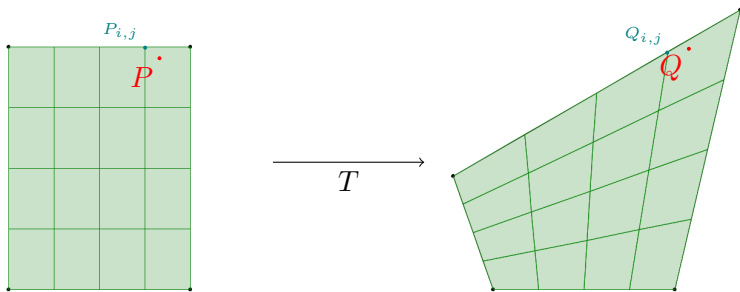
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Free-Form-Deformations and inverse problem solving



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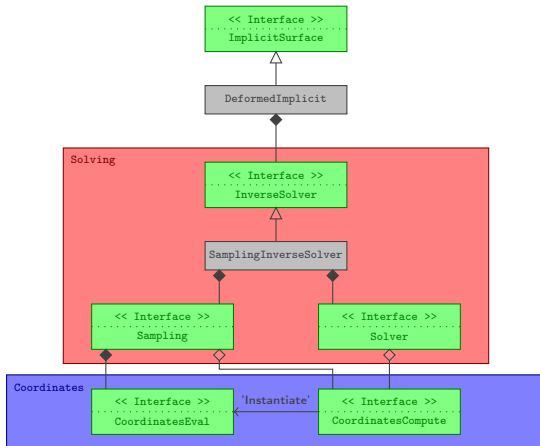
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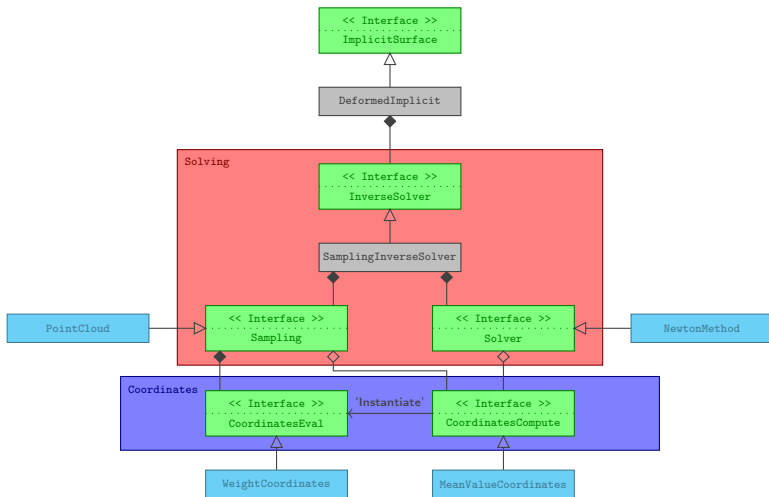
Free-Form-Deformations implicit surface example in 2D

Example in 2D (render done in java using Processing IDE).

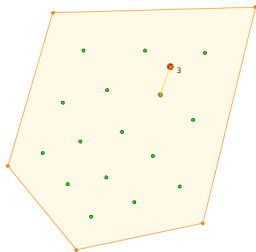
Our general architecture for inverse position solving



State of the art inspired idea : architecture



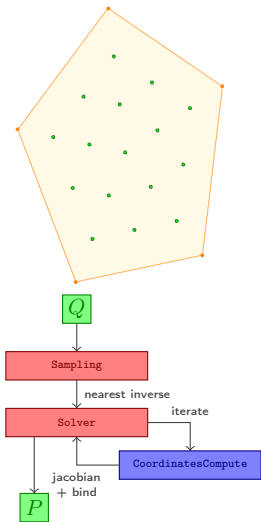
State of the art inspired idea : method



P_0 nearest sample.

$$P_{n+1} = P_n - J_T^{-1}(P_n) (T(P_n) - Q)$$

State of the art inspired idea : method



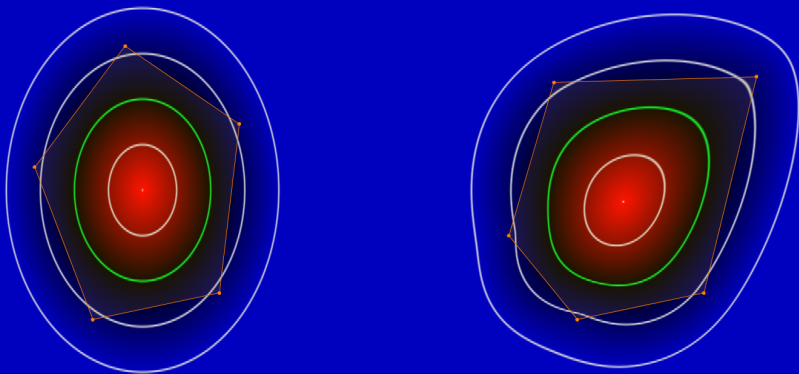
Solve $T(P) - Q = 0$:

$$J_T(x, y) = \left(\frac{\partial}{\partial x} T(x, y), \frac{\partial}{\partial y} T(x, y) \right)$$

P_0 nearest sample.

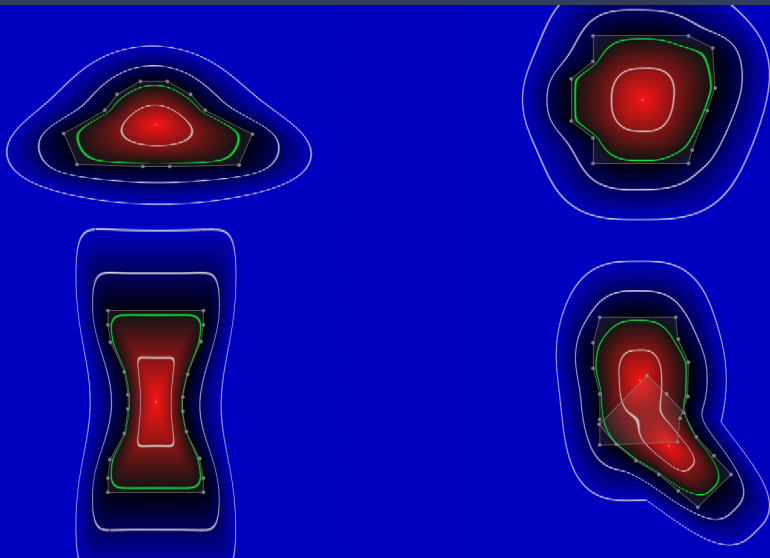
$$P_{n+1} = P_n - J_T^{-1}(P_n) (T(P_n) - Q)$$

State of the art inspired idea : result in 2D

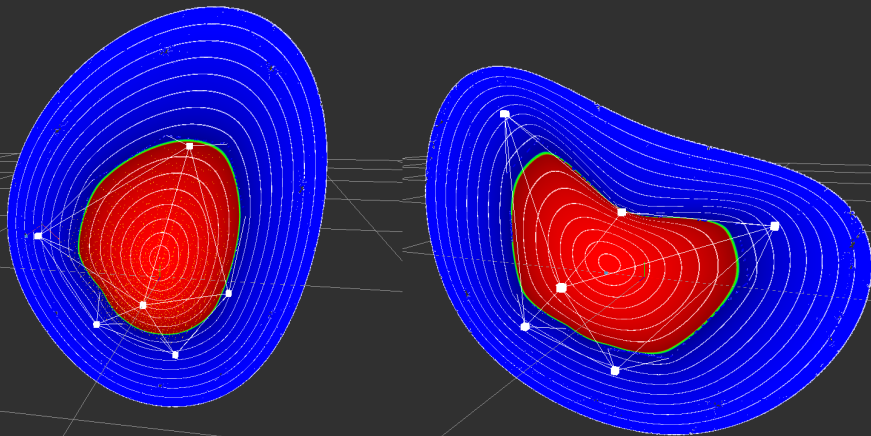


Example in 2D (render done in java using Processing IDE).

State of the art inspired idea : examples in 2D

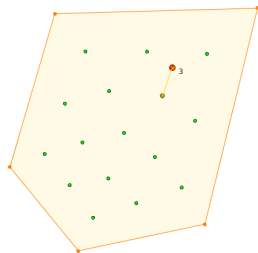
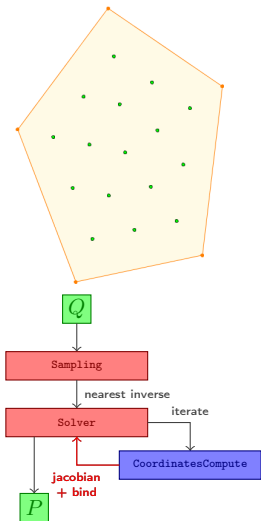


State of the art inspired idea : result in 3D



Example in 3D (render done in C++ using Radium Engine).

State of the art inspired idea : method



Solve $T(P) - Q = 0$:

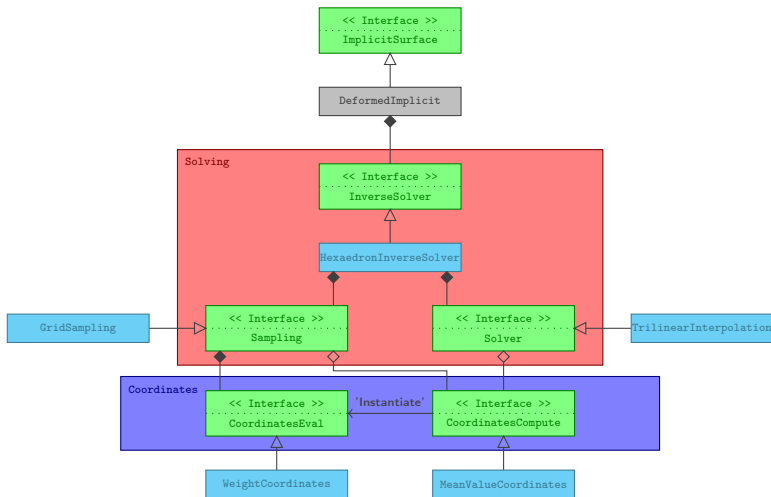
$$J_T(x, y) = \left(\frac{\partial}{\partial x} T(x, y), \frac{\partial}{\partial y} T(x, y) \right)$$

P_0 nearest sample.

$$P_{n+1} = P_n - J_T^{-1}(P_n) (T(P_n) - Q)$$

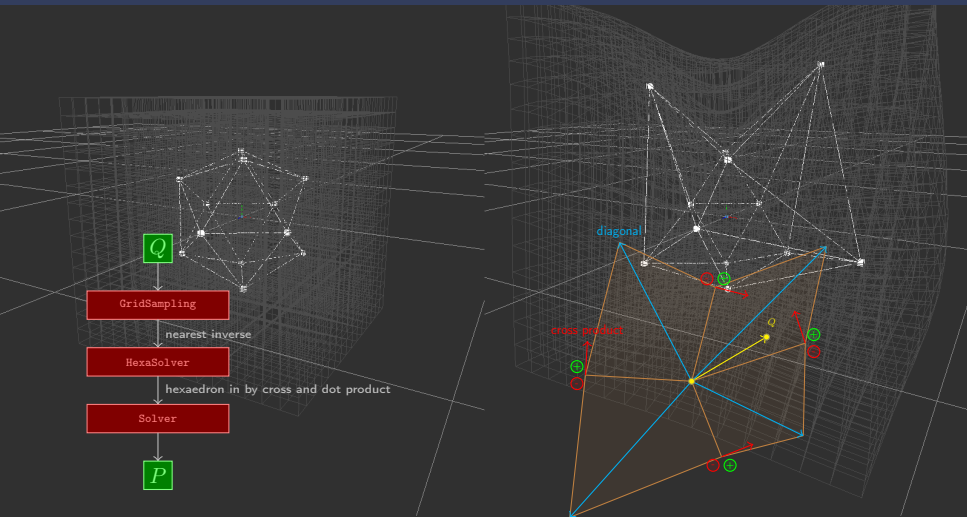
Bounding grid sampling and trilinear approximation by nearest neighbor detect cross and dot product method

Bounding grid and trilinear approximation : architecture



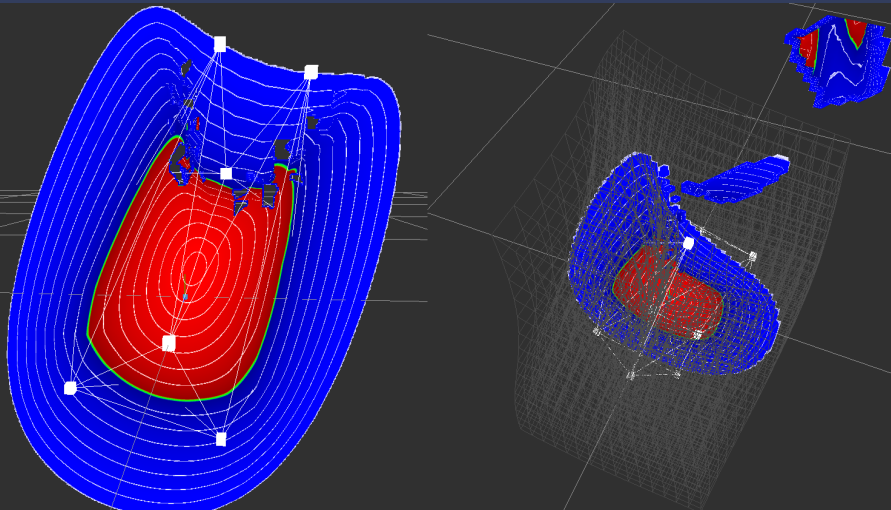
Bounding grid sampling and trilinear approximation by nearest neighbor detect cross and dot product method

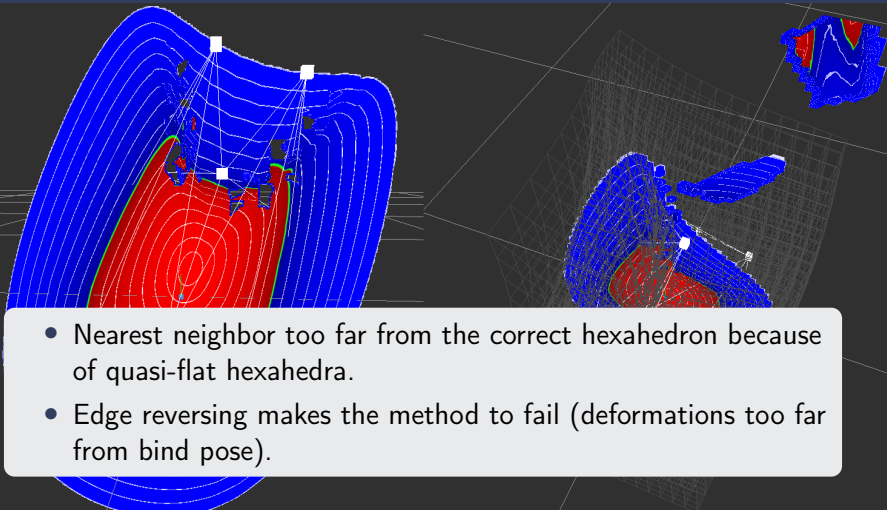
Bounding grid and trilinear approximation : method



Bounding grid sampling and trilinear approximation by nearest neighbor detect cross and dot product method

Bounding grid nearest neighbor : robustness to deformations problems

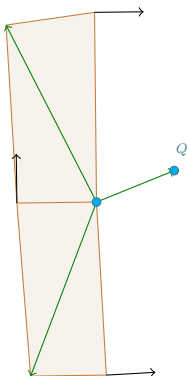




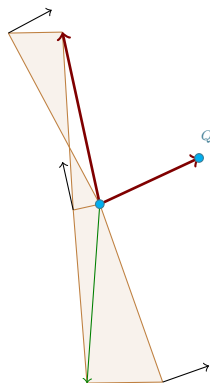
Edge reversed :

Grid reversal fail : example

Correctly oriented :

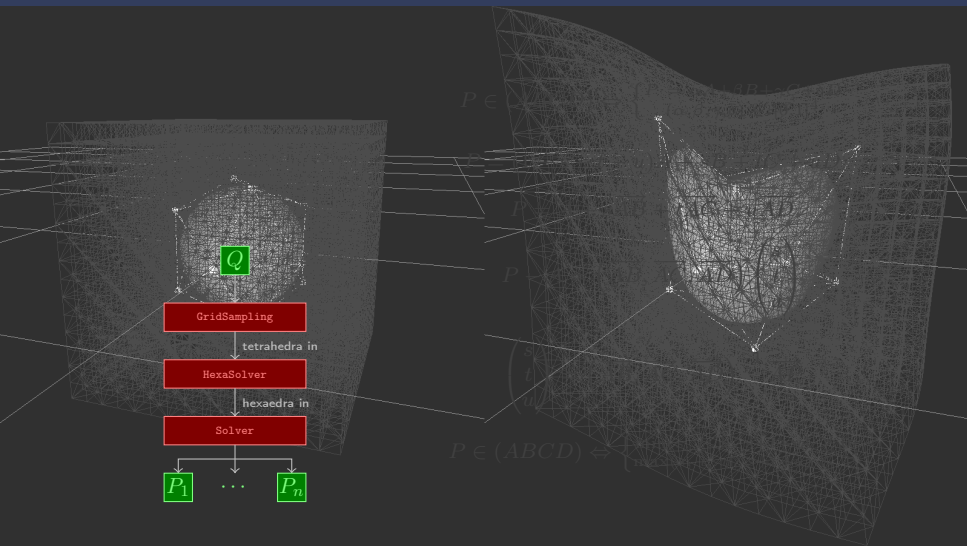


Edge reversed :

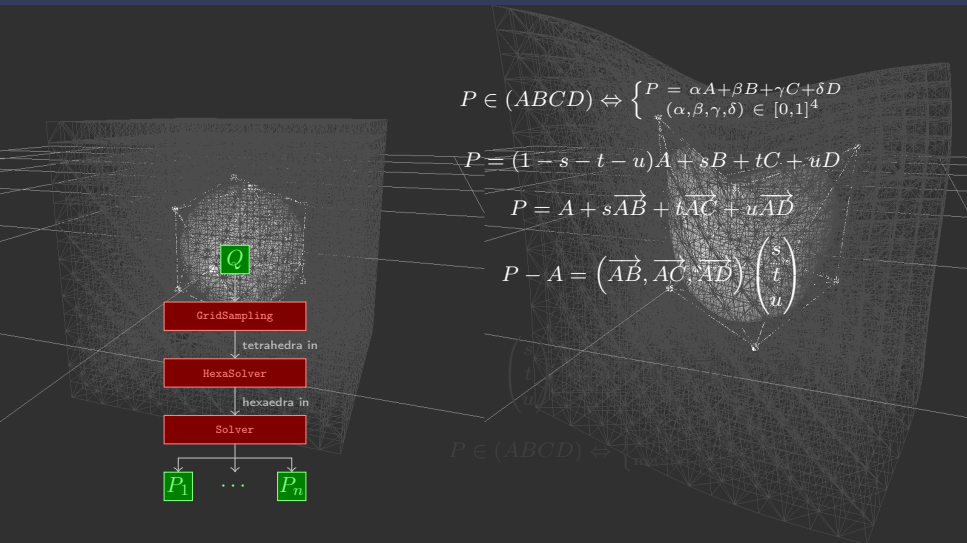


Bounding grid tetrahedral cut : method

Bounding grid tetrahedral cut : method



Bounding grid tetrahedral cut : method



Bounding grid tetrahedral cut : method



$$P \in (ABCD) \Leftrightarrow \begin{cases} P = \alpha A + \beta B + \gamma C + \delta D \\ (\alpha, \beta, \gamma, \delta) \in [0, 1]^4 \end{cases}$$

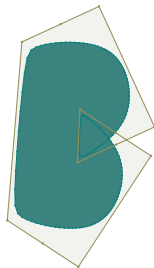
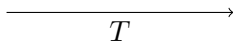
$$P = (1 - s - t - u)A + sB + tC + uD$$

$$P = A + s\overrightarrow{AB} + t\overrightarrow{AC} + u\overrightarrow{AD}$$

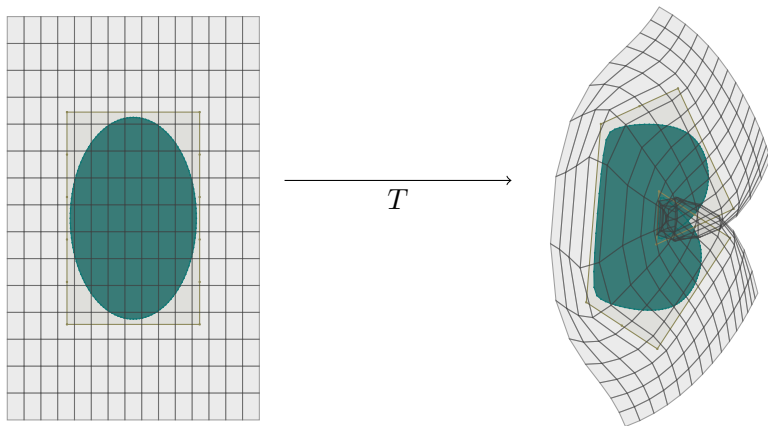
$$P - A = \begin{pmatrix} \overrightarrow{AB} & \overrightarrow{AC} & \overrightarrow{AD} \end{pmatrix} \begin{pmatrix} s \\ t \\ u \end{pmatrix}$$

$$\begin{pmatrix} s \\ t \\ u \end{pmatrix} = \begin{pmatrix} \overrightarrow{AB} & \overrightarrow{AC} & \overrightarrow{AD} \end{pmatrix}^{-1} \overrightarrow{AP}$$

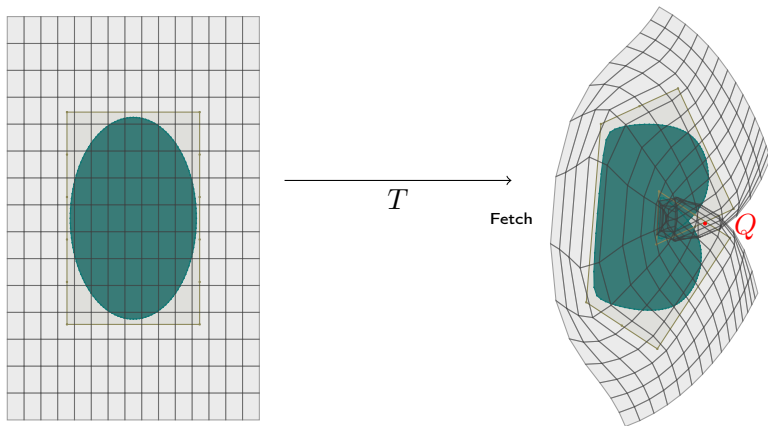
$$P \in (ABCD) \Leftrightarrow \begin{cases} \min(s, t, u, s+t+u) \geq 0 \\ \max(s, t, u, s+t+u) \leq 1 \end{cases}$$



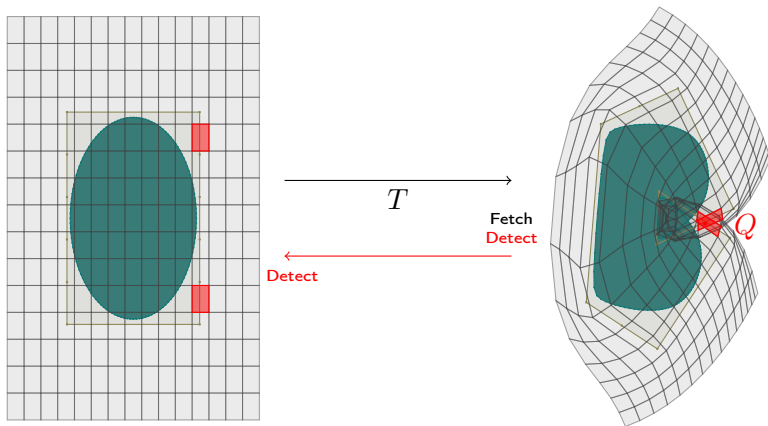
Multiple equivalents solving : example



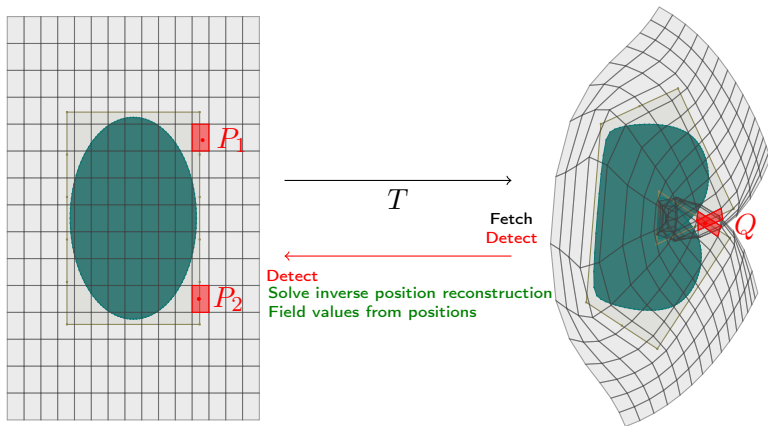
Multiple equivalents solving : example



Multiple equivalents solving : example



Multiple equivalents solving : example



Bounding grid tetrahedral cut : benchmark display planes

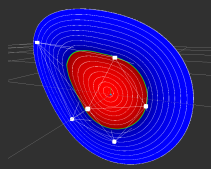
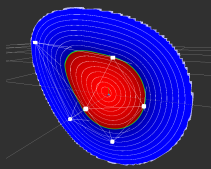
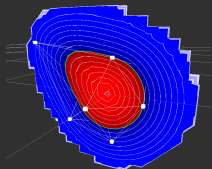
Implicit blob sphere deformed by cage (12 vertices, 20 faces) :

resolution
time
visualize

$32 \times 32 (2^{10})$
 $2.7 \cdot 10^7 ns (27ms)$

$128 \times 128 (2^{14})$
 $6 \cdot 10^7 ns (60ms)$

$512 \times 512 (2^{18})$
 $8.05 \cdot 10^8 ns (805ms)$



BIH and sampling update time : $1.7 \cdot 10^7 ns (17ms)$ in average.

Benchmark overview

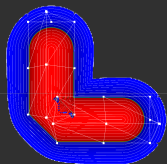
Implicit blob sphere deformed by cage (12 vertices, 20 faces) :

Method :	State of the art inspired	Bounding grid nearest neighbor	Bounding grid tetrahedral cut
<i>Update time</i>	$2.4 \cdot 10^6 ns(2.4ms)$	$2.1 \cdot 10^7 ns(21ms)$	$1.7 \cdot 10^7 ns(17ms)$
<i>Marching cubes :</i>			
$16 \times 16 \times 16 (2^{12})$	$9.24 \cdot 10^8 ns(0.924s)$	$9.9 \cdot 10^7 ns(99ms)$	$8.5 \cdot 10^7 ns(85ms)$
$32 \times 32 \times 32 (2^{15})$	$4.6 \cdot 10^9 ns(4.6s)$	$4.1 \cdot 10^8 ns(410ms)$	$3.31 \cdot 10^8 ns(331ms)$
$64 \times 64 \times 64 (2^{18})$	$2.65 \cdot 10^{10} ns(26.5s)$	$1.8 \cdot 10^9 ns(1.8s)$	$1.37 \cdot 10^9 ns(1.37s)$
<i>Display plane :</i>			
$32 \times 32 (2^{10})$	$1.62 \cdot 10^8 ns(162ms)$	$7.9 \cdot 10^6 ns(7.9ms)$	$2.7 \cdot 10^7 ns(27ms)$
$128 \times 128 (2^{14})$	$2.4 \cdot 10^9 ns(2.4s)$	$8 \cdot 10^7 ns(80ms)$	$6 \cdot 10^7 ns(60ms)$
$512 \times 512 (2^{18})$	$3.82 \cdot 10^{10} ns(38.2s)$	$1.29 \cdot 10^9 ns(1.29s)$	$8.05 \cdot 10^8 ns(805ms)$

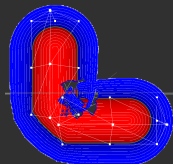
Multiple equivalent self-intersection results

Implicit blob capsule deformed by cage (26 vertices, 48 faces) :

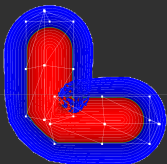
State of the art inspired
 $1 \cdot 10^{11} ns (100s)$



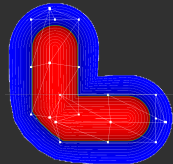
Bounding grid nearest neighbor
 $1.4 \cdot 10^9 ns (1.4s)$



Bounding grid tetrahedral detection
 $6.93 \cdot 10^8 ns (693ms)$



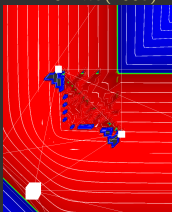
Bounding grid tetrahedral multiple detection
 $7.17 \cdot 10^8 ns (717ms)$



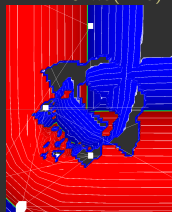
Multiple equivalent self-intersection results : zoom

Implicit blob capsule deformed by cage (26 vertices, 48 faces) :

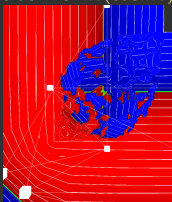
State of the art inspired
 $1 \cdot 10^{11} ns (100s)$



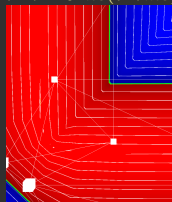
Bounding grid nearest neighbor
 $1.4 \cdot 10^9 ns (1.4s)$



Bounding grid tetrahedral detection
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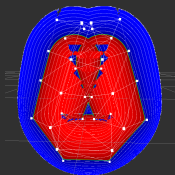
Bounding grid tetrahedral multiple detection
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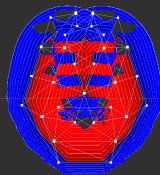
Multiple equivalent self-intersection and contact results

Implicit blob capsule deformed by cage (42 vertices, 80 faces) :

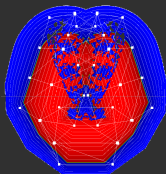
State of the art inspired
 $1.58 \cdot 10^{11} ns (158s)$



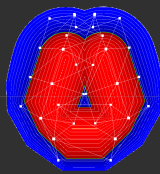
Bounding grid nearest neighbor
 $2.2 \cdot 10^9 ns (2.2s)$



Bounding grid tetrahedral detection
 $1.75 \cdot 10^9 ns (1.75s)$

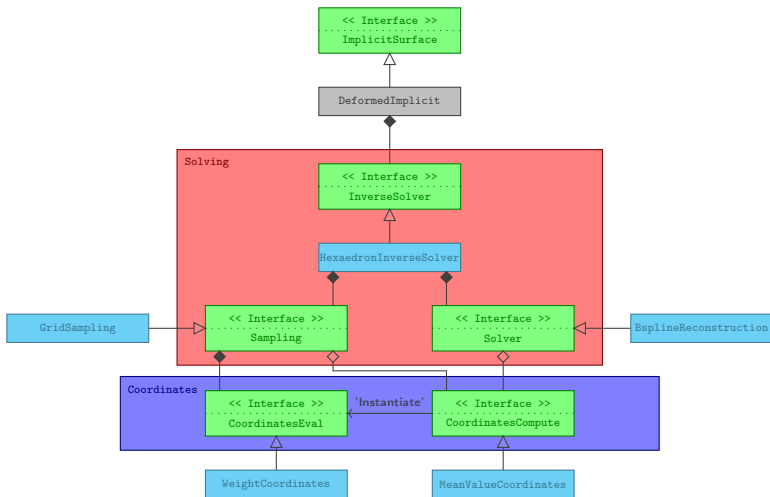


Bounding grid tetrahedral multiple detection
 $1.67 \cdot 10^9 ns (1.67s)$



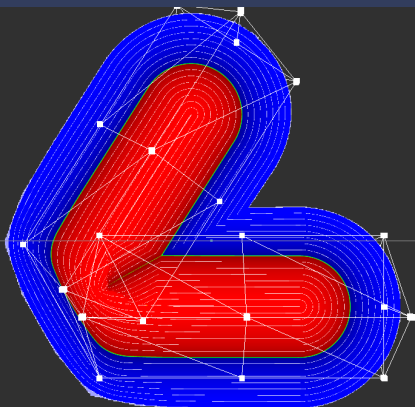
Field reconstruction from sampling : approximation from B-spline

Bounding grid B-spline reconstruction : architecture



Field reconstruction from sampling : approximation from B-spline

B-spline reconstruction



$$T(u, v, w) = \sum_{(i,j,k) \in [r]^3} \mathcal{N}_i^3(u) \mathcal{N}_j^3(v) \mathcal{N}_k^3(w)$$

$$J_T(u, v, w) = \left(\frac{\partial}{\partial u} T(u, v, w), \frac{\partial}{\partial v} T(u, v, w), \frac{\partial}{\partial w} T(u, v, w) \right)$$

$$\begin{pmatrix} u_{n+1} \\ v_{n+1} \\ w_{n+1} \end{pmatrix} = \begin{pmatrix} u_n \\ v_n \\ w_n \end{pmatrix} - J_T^{-1} \left(\begin{pmatrix} u_n \\ v_n \\ w_n \end{pmatrix} \right) \left(T \left(\begin{pmatrix} u_n \\ v_n \\ w_n \end{pmatrix} \right) - Q \right)$$

Hexahedral detection.

Get starting (u_0, v_0, w_0) .

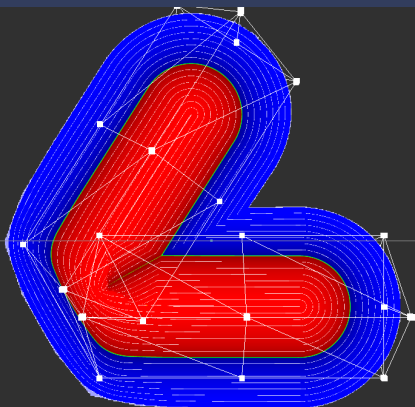
Solve $T(u, v, w) - Q = 0$
using Newton method.

Open nodal vector : $\left[0, 0, 0, \frac{1}{r-2}, \dots, \frac{r-3}{r-2}, 1, 1, 1 \right]$

Constraint : $(u_n, v_n, w_n) \in (0, 1)^3$

Field reconstruction from sampling : approximation from B-spline

B-spline reconstruction



$$T(u, v, w) = \sum_{(i,j,k) \in [r]^3} \mathcal{N}_i^3(u) \mathcal{N}_j^3(v) \mathcal{N}_k^3(w)$$

$$J_T(u, v, w) = \left(\frac{\partial}{\partial u} T(u, v, w), \frac{\partial}{\partial v} T(u, v, w), \frac{\partial}{\partial w} T(u, v, w) \right)$$

$$\begin{pmatrix} u_{n+1} \\ v_{n+1} \\ w_{n+1} \end{pmatrix} = \begin{pmatrix} u_n \\ v_n \\ w_n \end{pmatrix} - J_T^{-1} \left(\begin{pmatrix} u_n \\ v_n \\ w_n \end{pmatrix} \right) \left(T \left(\begin{pmatrix} u_n \\ v_n \\ w_n \end{pmatrix} \right) - Q \right)$$

Hexahedral detection.

Get starting (u_0, v_0, w_0) .

Solve $T(u, v, w) - Q = 0$
using Newton method.

Open nodal vector : $\left[0, 0, 0, \frac{1}{r-2}, \dots, \frac{r-3}{r-2}, 1, 1, 1 \right]$

Constraint : $(u_n, v_n, w_n) \in (0, 1)^3$

Field reconstruction from sampling : approximation from B-spline

Reconstruction compare

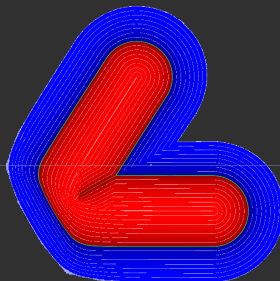
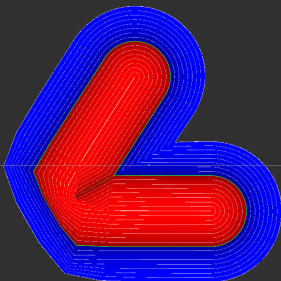
Implicit blob capsule deformed by cage (26 vertices, 48 faces) :

Trilinear reconstruction

$3.21 \cdot 10^9 ns (3.2s)$

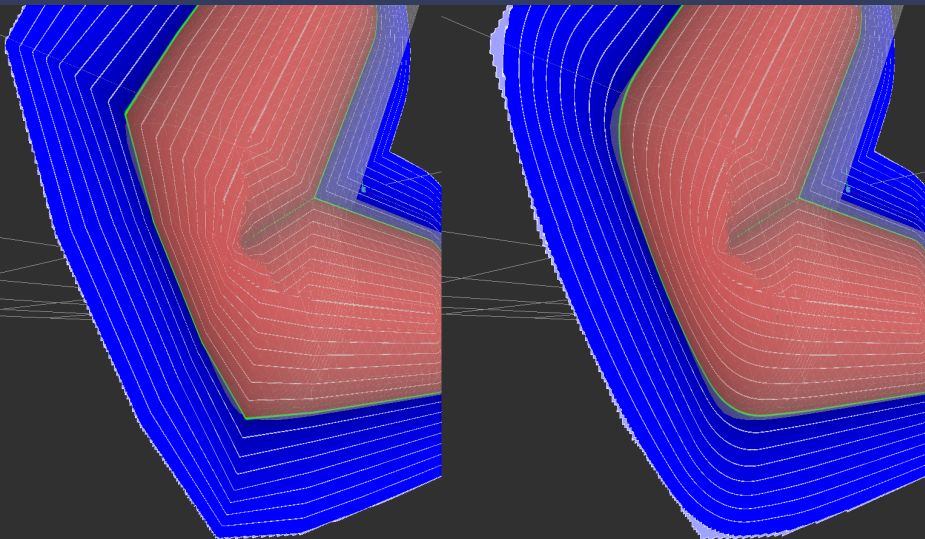
B-spline reconstruction

$1.19 \cdot 10^{11} ns (119s)$



Field reconstruction from sampling : approximation from B-spline

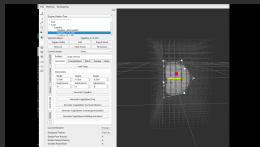
Reconstruction and coordinates system fitting



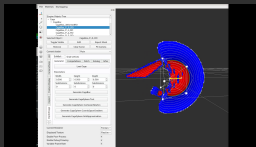
Space fold-over : movie time

Space fold-over solutions overview :

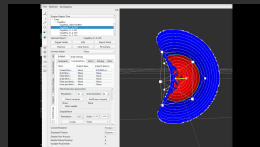
Bounding grid deform



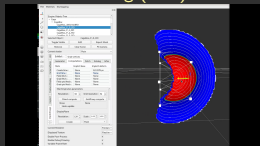
Bounding grid nearest neighbor



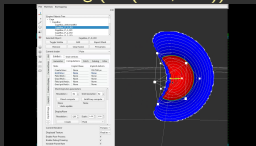
Bounding grid tetrahedral detection



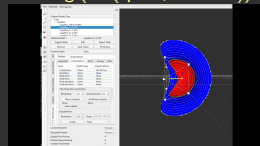
Bounding grid tetrahedral multiple detection solving (max)



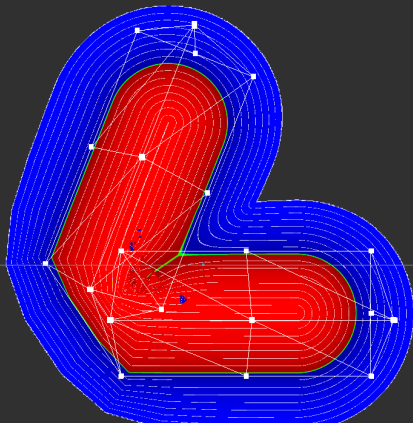
Bounding grid tetrahedral multiple detection solving (diff(max, min))



Bounding grid tetrahedral multiple and reversal detection solving (diff(space, reversed))

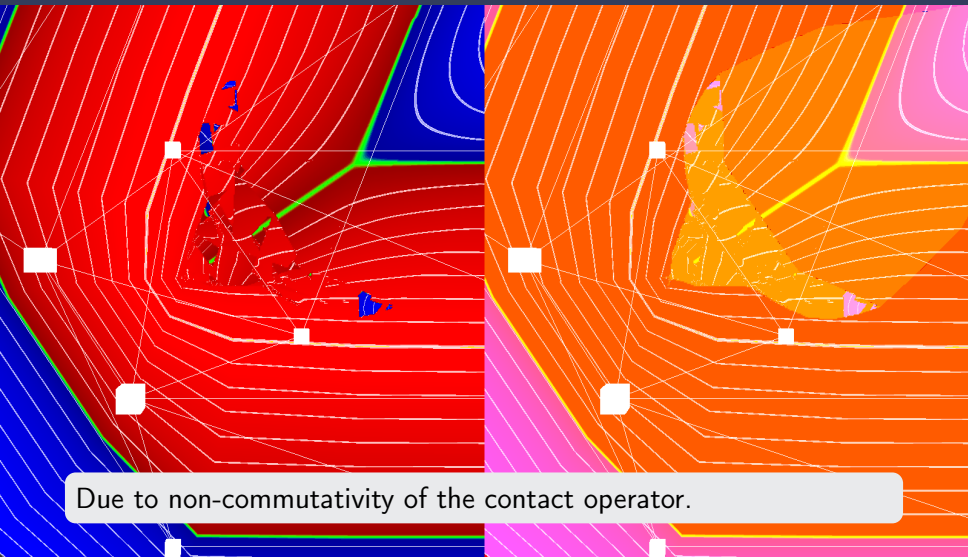


Multiple equivalents : order problem



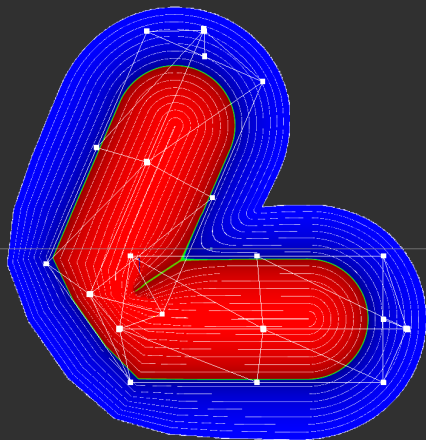
Discontinuity inside the 3-multiple equivalents' area.

Multiple equivalents : order problem (zoom)



Multiple field values : oriented-reversal classification and dual composition

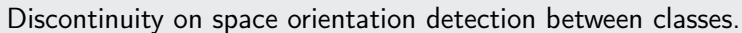
Classification and dual composition



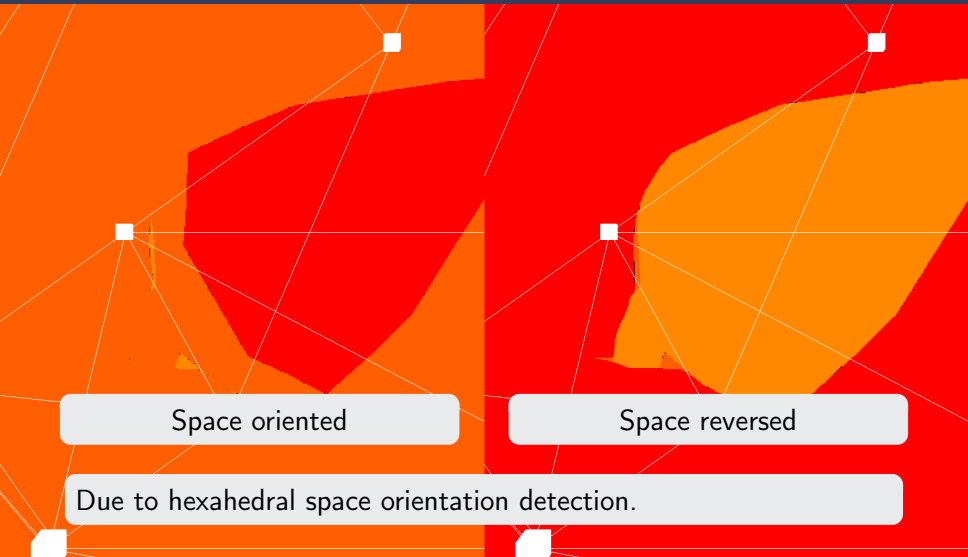


Compose other cases with max operator.

Classification and dual composition discontinuity

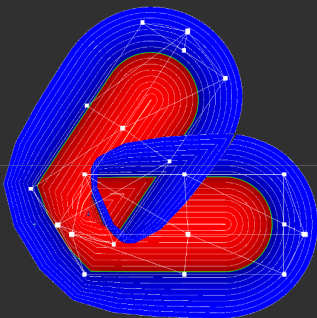


Classification and dual composition problem overview

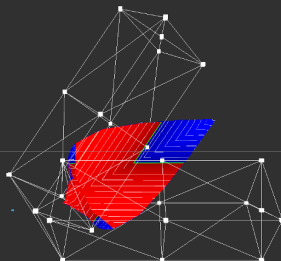


Multiple field values : implicit order solving

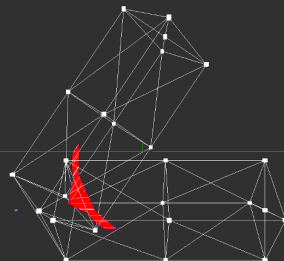
Implicit order : classes



Minimal value



Median value

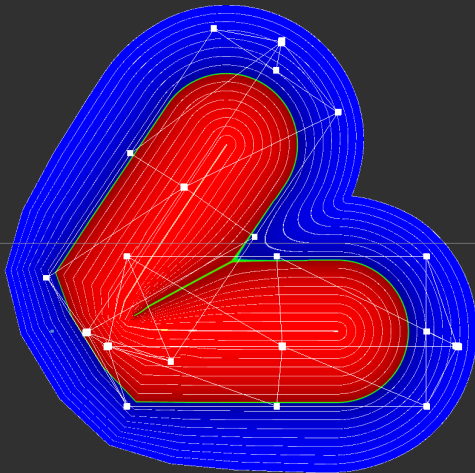


Maximal value

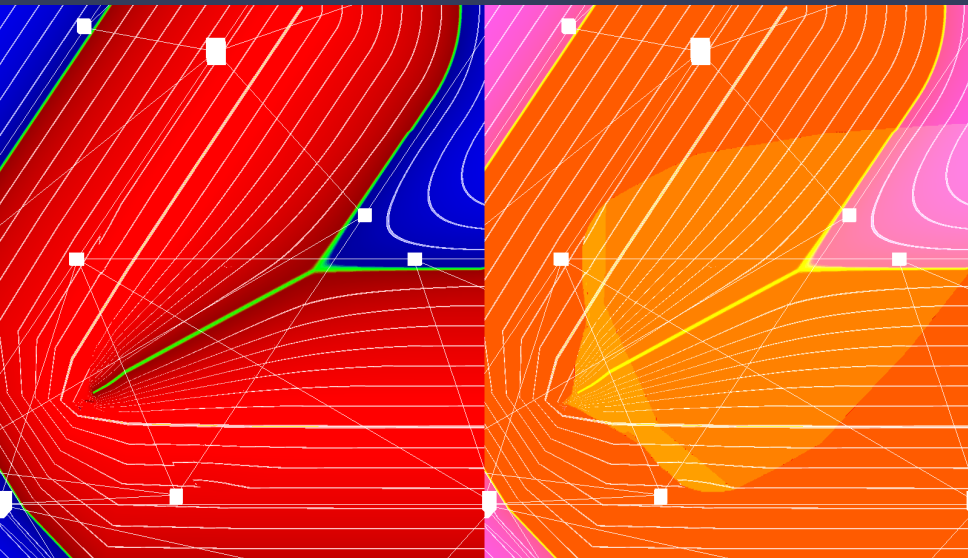
Sorting field value for each fetch.

Multiple field values : implicit order solving

Implicit order : composition result

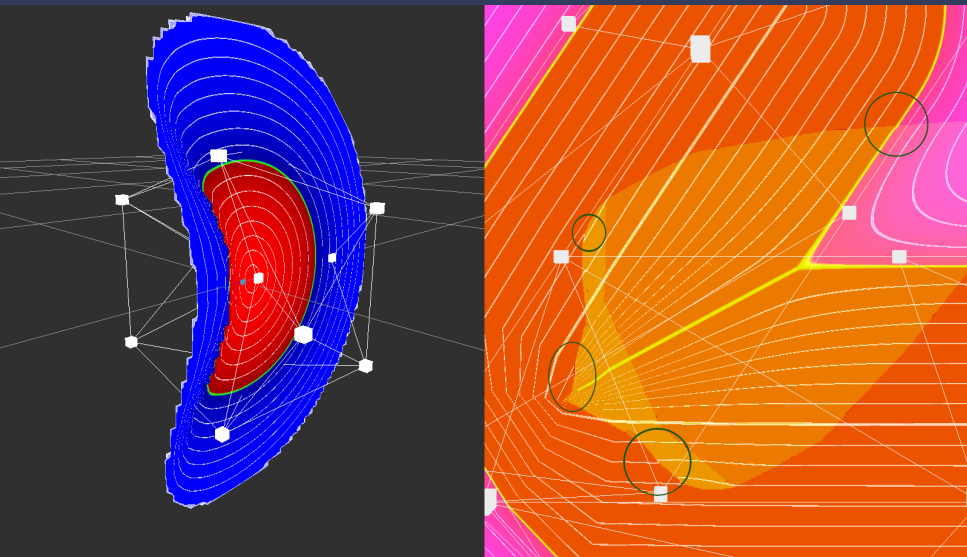


Implicit order : composition and discontinuities

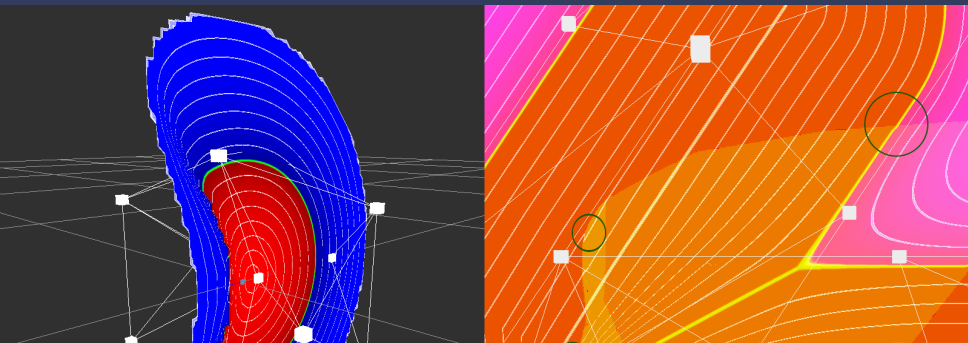


Discontinuity between the equivalents' classes.

Improve continuity of the method



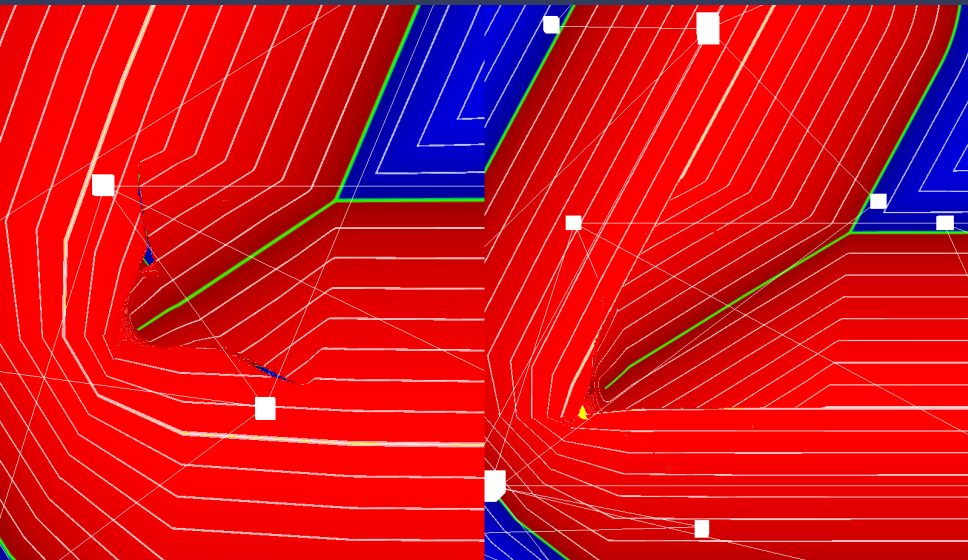
Improve continuity of the method



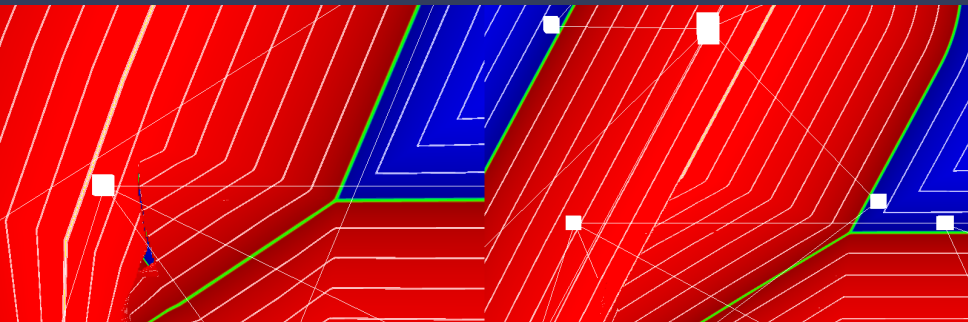
Continuity for implicit skinning :

- Define operator between cage and field to correct compressions of the field ?
- Find a better design for the sampling than an uniform grid ?

Limitation : Field reconstruction lack of relevancy



Limitation : Field reconstruction lack of relevancy

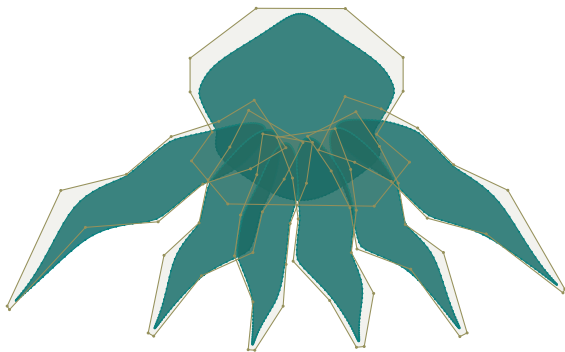


Field relevancy problems for implicit skinning :

- Compression of the field in too many equivalents areas.
- Design n-ary operator ?
- Find a better classification for binary composition ?
- Part of the multiple equivalents' classes missing ?

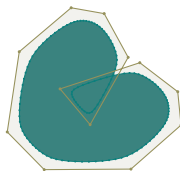
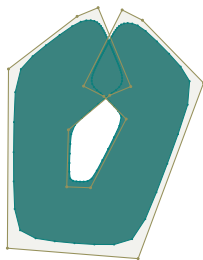
Integration to Implicit skinning

- Add smooth deformations.
- Allow free deformations for animators.

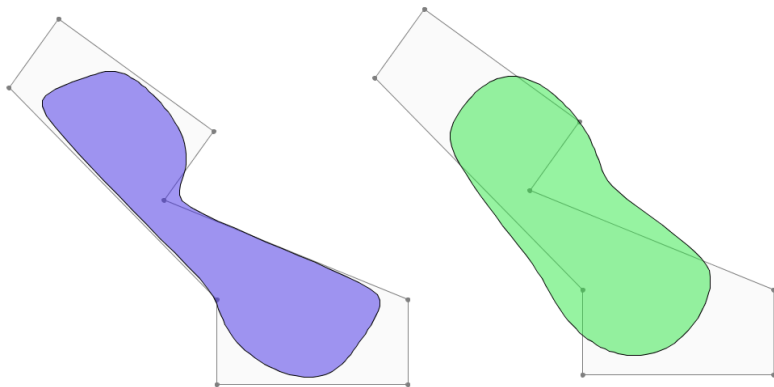


Correct mesh self-intersection

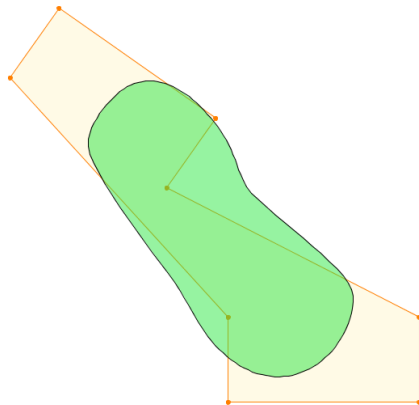
Case : Self-contact Self-intersection ...
 Solution : contact operator skinning operator
 Visualize :



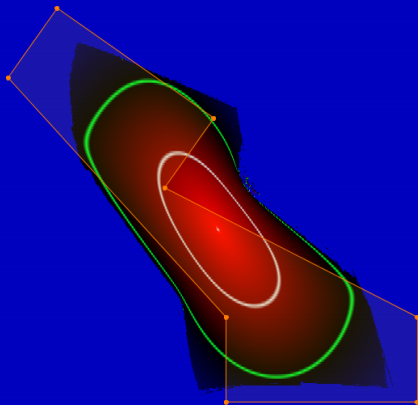
Green and MeanValue compare in 2d



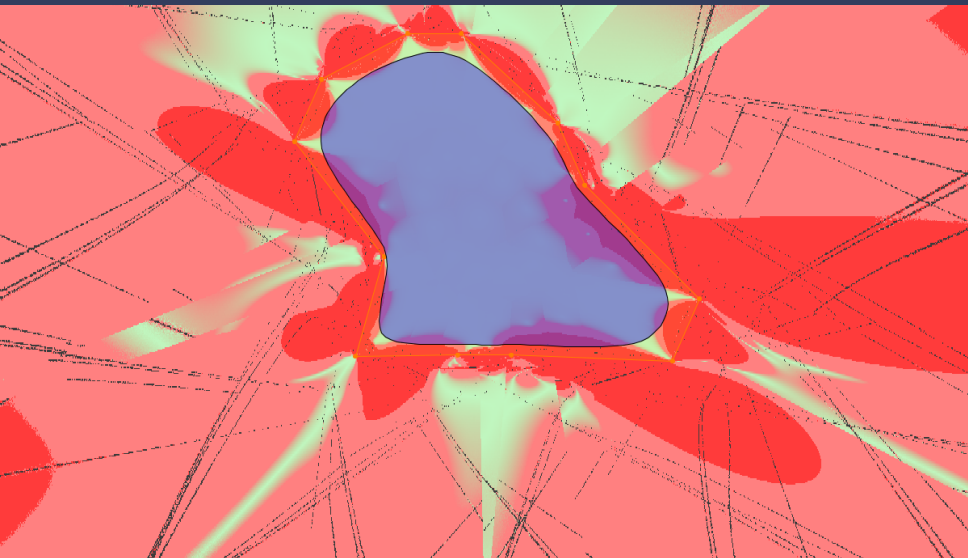
Green deform 2d



Green field 2d



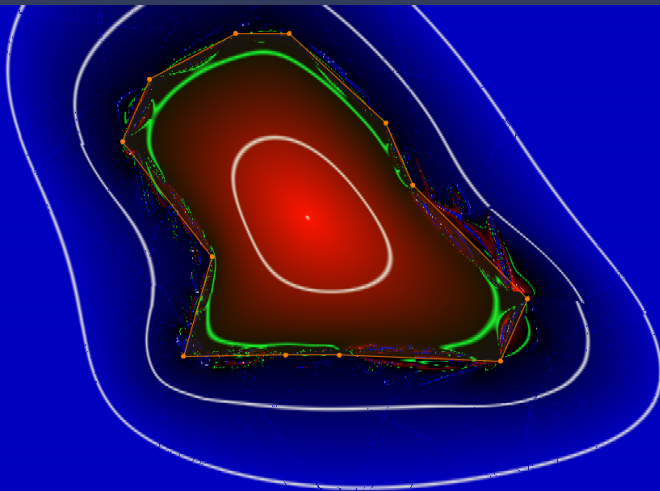
Using computed gradient : error



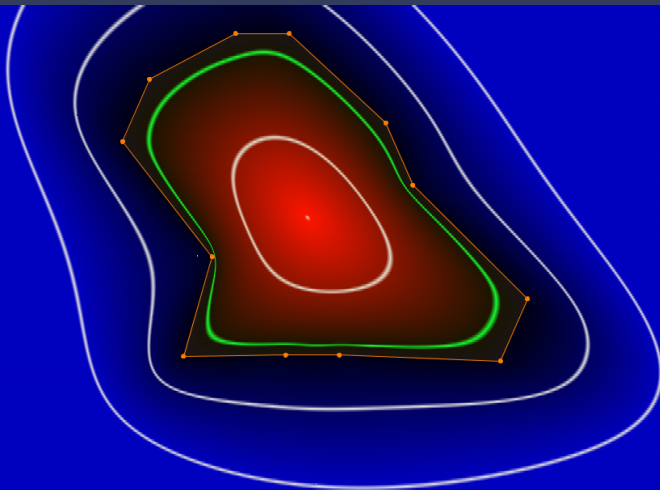
Using discrete gradient : error



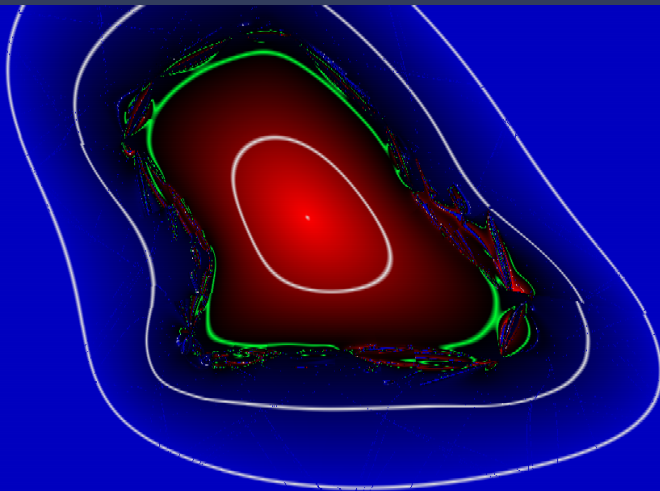
Using computed gradient : field



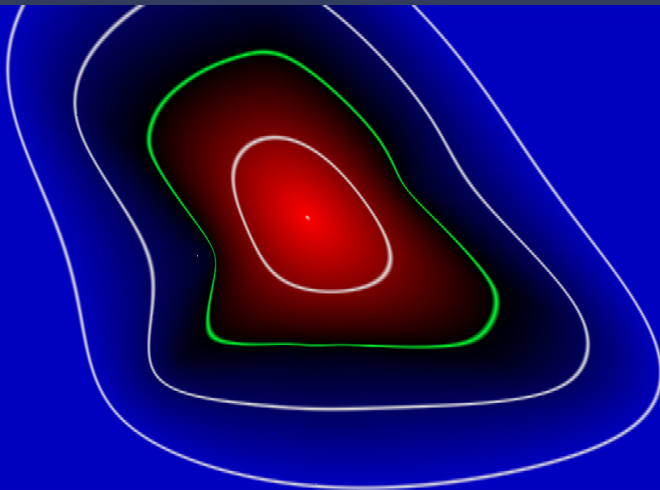
Using discrete gradient : field



Using computed gradient : field without cage



Using discrete gradient : field without cage



Gradient discontinuity on edges

$$a_i = C_i - P, \quad b_i = C_{i+1} - P, \quad \delta_i = \frac{\det(a_i, b_i)}{|\det(a_i, b_i)|} \frac{a_i \cdot b_i}{\|a_i\| \|b_i\|}$$

$$\omega_i = \frac{\tan\left(\frac{\delta_{i-1}}{2}\right) + \tan\left(\frac{\delta_i}{2}\right)}{\|a_i\|}, \quad P = \frac{\sum_{i \in [n]} \omega_i C_i}{\sum_{i \in [n]} \omega_i}$$

$$\frac{\partial}{\partial \gamma} \delta_i = \frac{\partial}{\partial \gamma} \frac{a_i \cdot b_i}{\|a_i\| \|b_i\|} = (a_i \cdot b_i) \frac{a_{i,\gamma} \|b_i\|^2 + b_{i,\gamma} \|a_i\|^2}{\|a_i\|^3 \|b_i\|^3} - \frac{a_{i,\gamma} + b_{i,\gamma}}{\|a_i\| \|b_i\|}$$

$$\frac{\partial}{\partial \gamma} \tan\left(\frac{\delta_i}{2}\right) = \frac{1 + \tan^2\left(\frac{\delta_i}{2}\right)}{2} \sqrt{\frac{\|a_i\| \|b_i\|}{\|a_i\| \|b_i\| - a_i \cdot b_i}} \frac{\partial}{\partial \gamma} \delta_i \frac{\det(a_i, b_i)}{|\det(a_i, b_i)|}$$

$$\frac{\partial}{\partial \gamma} \omega_i = \frac{\left(\frac{\partial}{\partial \gamma} \tan\left(\frac{\delta_{i-1}}{2}\right) + \frac{\partial}{\partial \gamma} \tan\left(\frac{\delta_i}{2}\right)\right) \|a_i\| + \frac{a_{i,\gamma}}{\|a_i\|} \left(\tan\left(\frac{\delta_{i-1}}{2}\right) + \tan\left(\frac{\delta_i}{2}\right)\right)}{\|a_i\|^2}$$

$$\frac{\partial}{\partial \gamma} T(P) = \frac{\left(\sum \frac{\partial}{\partial \gamma} \omega_i T(C_i)\right) (\sum \omega_i) - (\sum \omega_i T(C_i)) \left(\sum \frac{\partial}{\partial \gamma} \omega_i\right)}{(\sum \omega_i)^2}$$

Gradient discontinuity on edges

$$a_i = C_i - P, \quad b_i = C_{i+1} - P, \quad \delta_i = \frac{\det(a_i, b_i)}{|\det(a_i, b_i)|} \frac{a_i \cdot b_i}{||a_i||||b_i||}$$

$$\omega_i = \frac{\tan\left(\frac{\delta_{i-1}}{2}\right) + \tan\left(\frac{\delta_i}{2}\right)}{||a_i||}, \quad P = \frac{\sum_{i \in [n]} \omega_i C_i}{\sum_{i \in [n]} \omega_i}$$

$$\frac{\partial}{\partial \gamma} \delta_i = \frac{\partial}{\partial \gamma} \frac{a_i \cdot b_i}{||a_i||||b_i||} = (a_i \cdot b_i) \frac{a_{i,\gamma} ||b_i||^2 + b_{i,\gamma} ||a_i||^2}{||a_i||^3 ||b_i||^3} - \frac{a_{i,\gamma} + b_{i,\gamma}}{||a_i||||b_i||}$$

$$\frac{\partial}{\partial \gamma} \tan\left(\frac{\delta_i}{2}\right) = \frac{1 + \tan^2\left(\frac{\delta_i}{2}\right)}{2} \sqrt{\frac{||a_i||||b_i||}{||a_i||||b_i|| - a_i \cdot b_i}} \frac{\partial}{\partial \gamma} \delta_i \frac{\det(a_i, b_i)}{|\det(a_i, b_i)|}$$

$$\frac{\partial}{\partial \gamma} \omega_i = \frac{\left(\frac{\partial}{\partial \gamma} \tan\left(\frac{\delta_{i-1}}{2}\right) + \frac{\partial}{\partial \gamma} \tan\left(\frac{\delta_i}{2}\right)\right) ||a_i|| + \frac{a_{i,\gamma}}{||a_i||} \left(\tan\left(\frac{\delta_{i-1}}{2}\right) + \tan\left(\frac{\delta_i}{2}\right)\right)}{||a_i||^2}$$

$$\frac{\partial}{\partial \gamma} T(P) = \frac{\left(\sum \frac{\partial}{\partial \gamma} \omega_i T(C_i)\right) (\sum \omega_i) - (\sum \omega_i T(C_i)) \left(\sum \frac{\partial}{\partial \gamma} \omega_i\right)}{(\sum \omega_i)^2}$$

Gradient descent in Coordinates space

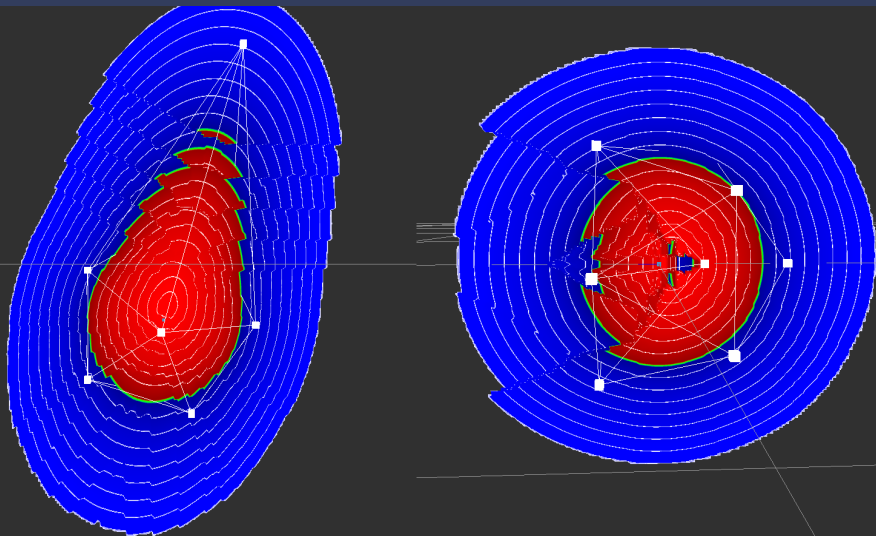
Minimize $\| \sum_{i \in [n]} \alpha_i T(C_i) - Q \|^2$.

$$\frac{\partial}{\partial \alpha_k} \| \sum_{i \in [n]} \alpha_i T(C_i) - Q \|^2 = 2T(C_k) \cdot \left(\sum_{i \in [n]} \alpha_i T(C_i) - Q \right)$$

$$\nabla \| \sum_{i \in [n]} \alpha_i T(C_i) - Q \|^2 = 2 \begin{pmatrix} T(C_1), T(C_2), \dots, T(C_n) \end{pmatrix} \cdot \left(\sum_{i \in [n]} \alpha_i T(C_i) - Q \right)$$

Do gradient descent.

Coordinates space gradient descent : result



Coordinates space gradient descent : benchmark display planes

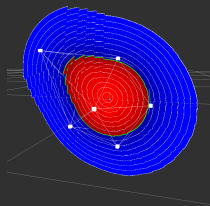
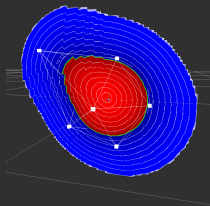
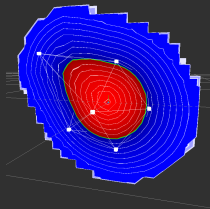
Implicit blob sphere deformed by cage (12 vertices, 20 faces) :

resolution
time
visualize

32×32 (2^{10})
 $1 \cdot 10^7 ns$ (10ms)

128×128 (2^{14})
 $1.58 \cdot 10^8 ns$ (158ms)

512×512 (2^{18})
 $2.4 \cdot 10^9 ns$ (2.4s)



Gradient descent in Coordinates space

Find $\{\mathfrak{C}_j\}_{j \in [l]}$ orthonormal basis of constraints from coordinates system.

Then do gradient descent on :

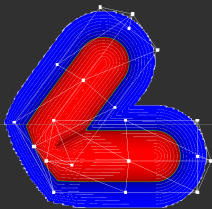
$$\nabla_{\mathfrak{C}} \left\| \sum_{i \in [n]} \alpha_i T(C_i) - Q \right\|^2 = \nabla \left\| \sum_{i \in [n]} \alpha_i T(C_i) - Q \right\|^2 - \sum_{j \in [l]} \frac{\nabla \left\| \sum_{i \in [n]} \alpha_i T(C_i) - Q \right\|^2 \cdot \mathfrak{C}_j}{\mathfrak{C}_j \cdot \mathfrak{C}_j} \mathfrak{C}_j$$

- Is constraints computation that relevant for time saving ?
- Nearest neighbor's like method : no multiple equivalent solving.
- Benchmark times near to Bounding grid methods.

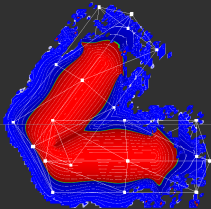
Reconstruction compare : fails

Implicit blob capsule deformed by cage (26 vertices, 48 faces) :

Trilinear reconstruction
 $2.16 \cdot 10^8 ns (216ms)$



Lagrange reconstruction
 $2.56 \cdot 10^{10} ns (25.6s)$



Bezier reconstruction
 $8.35 \cdot 10^{10} ns (83.5s)$

