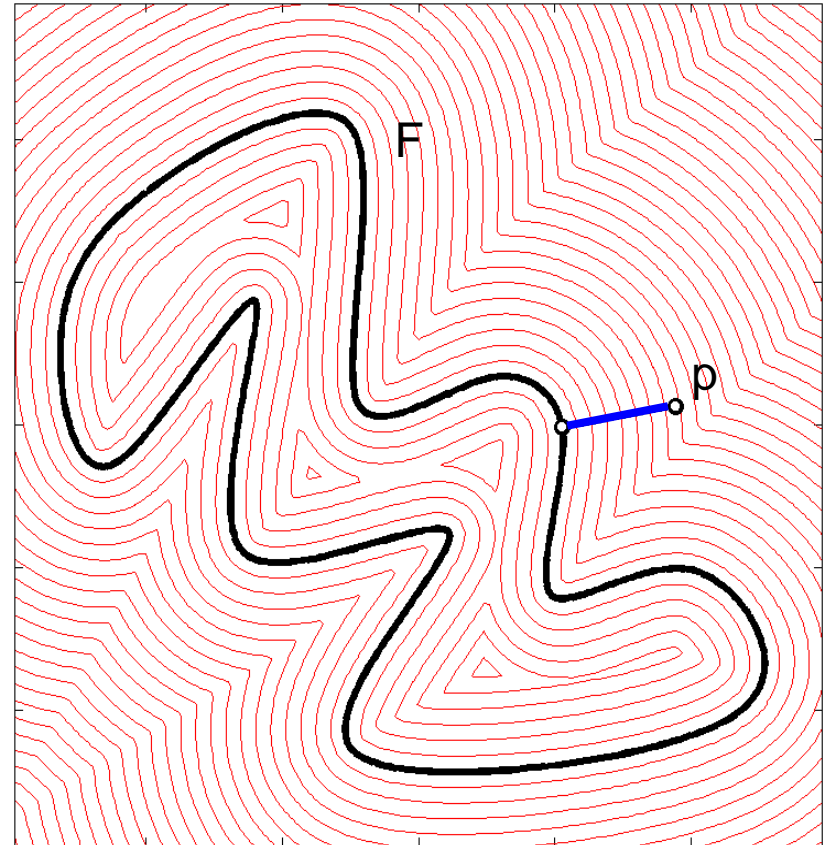


Distance Functions

Distance function

- Given: geometric object F (curve, surface, solid, ...)
- Assigns to each point the shortest distance from F
- Level sets of the distance function are **trimmed offsets**

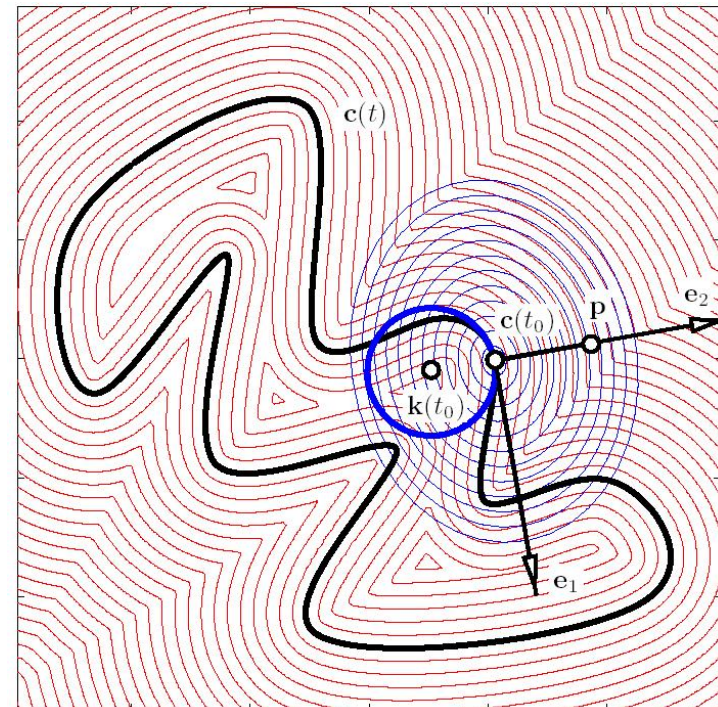
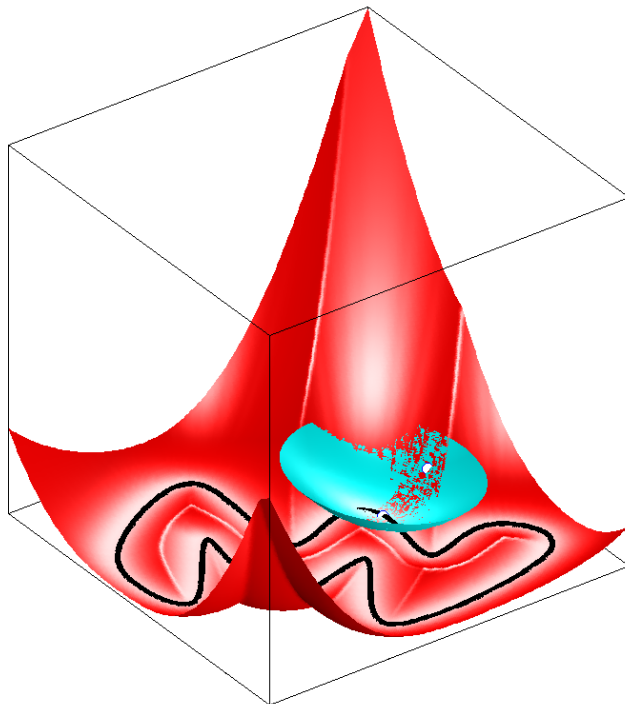


Graph surfaces of distance functions

- Graph surfaces of distance functions: (developable) surfaces of constant slope
- Each tangent plane has the inclination angle 45 degrees
- Not smooth at the base curve and at the cut locus



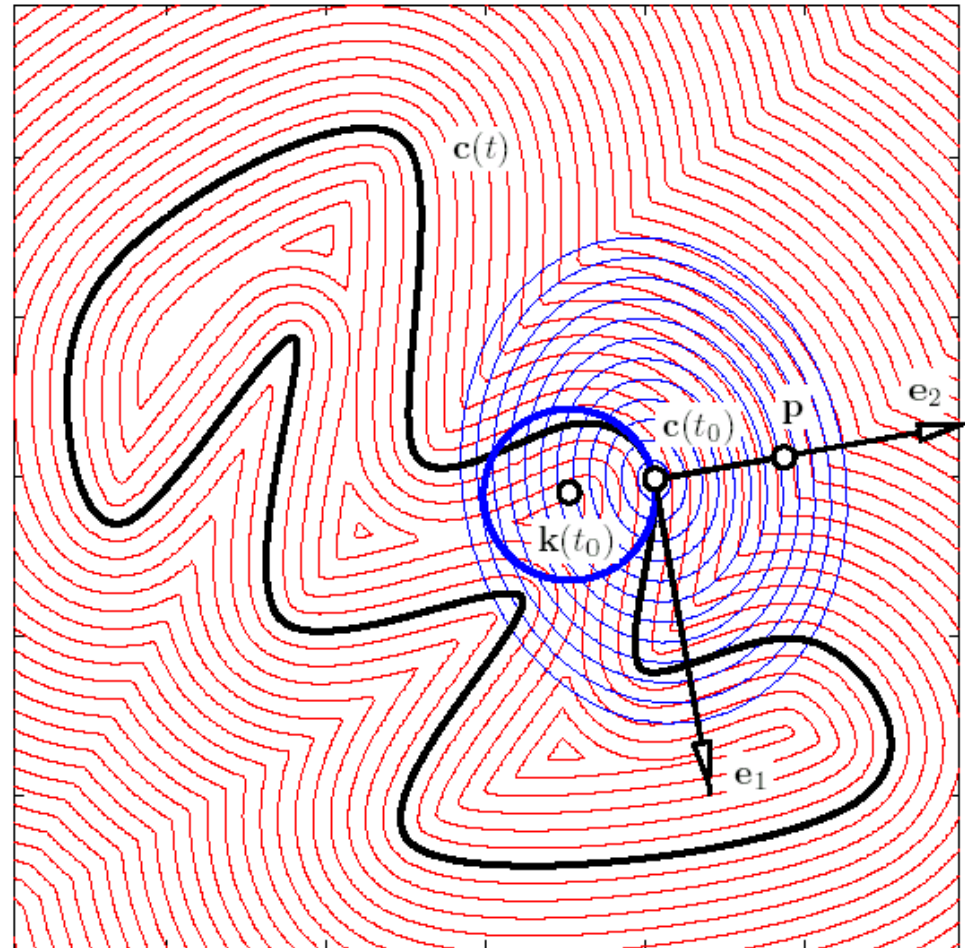
- For geometric optimization algorithms: *local quadratic approximation of the squared distance function*
- Outside cut locus



quadratic Taylor
approximant F_d of
the squared distance
function at p :

$$F_d(x_1, x_2) = \frac{d}{d - \rho} x_1^2 + x_2^2$$

x_1, x_2 are the
coordinates in the
Frenet frame at the
normal footpoint
 $c(t_0)$ of p .



Taylor approximant

$$F_d(x_1, x_2) = \frac{d}{d - \rho} x_1^2 + x_2^2$$

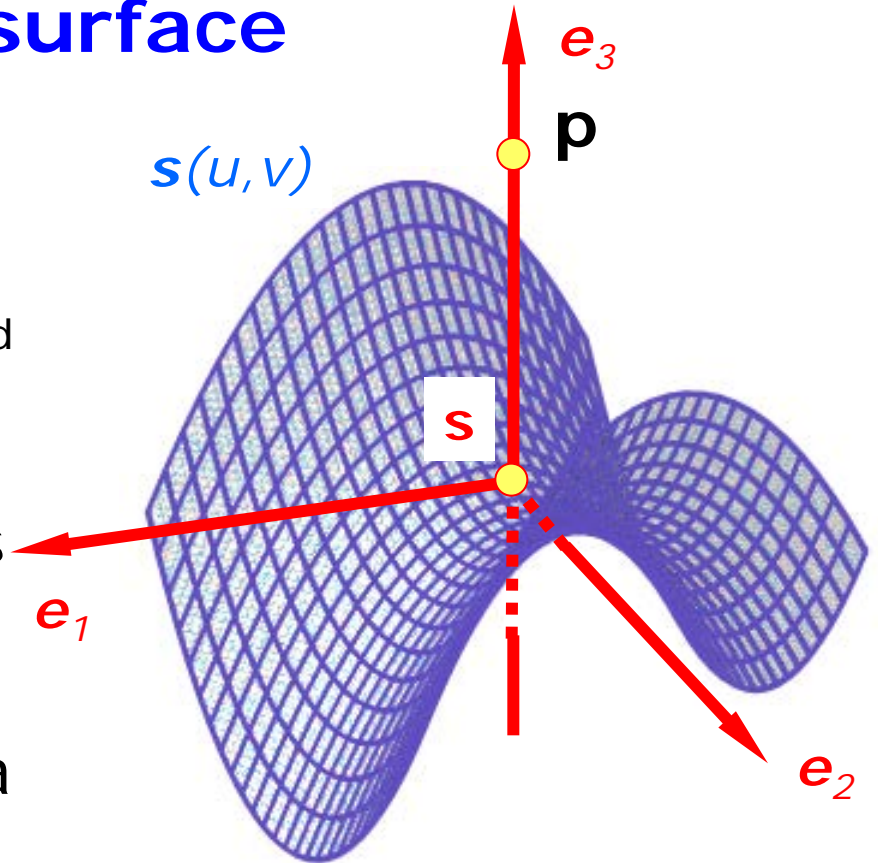
for $d=0$ (p on the curve):

$$F_d(x_1, x_2) = x_2^2$$

Squared distance to the tangent at $c(t_0) = p$.

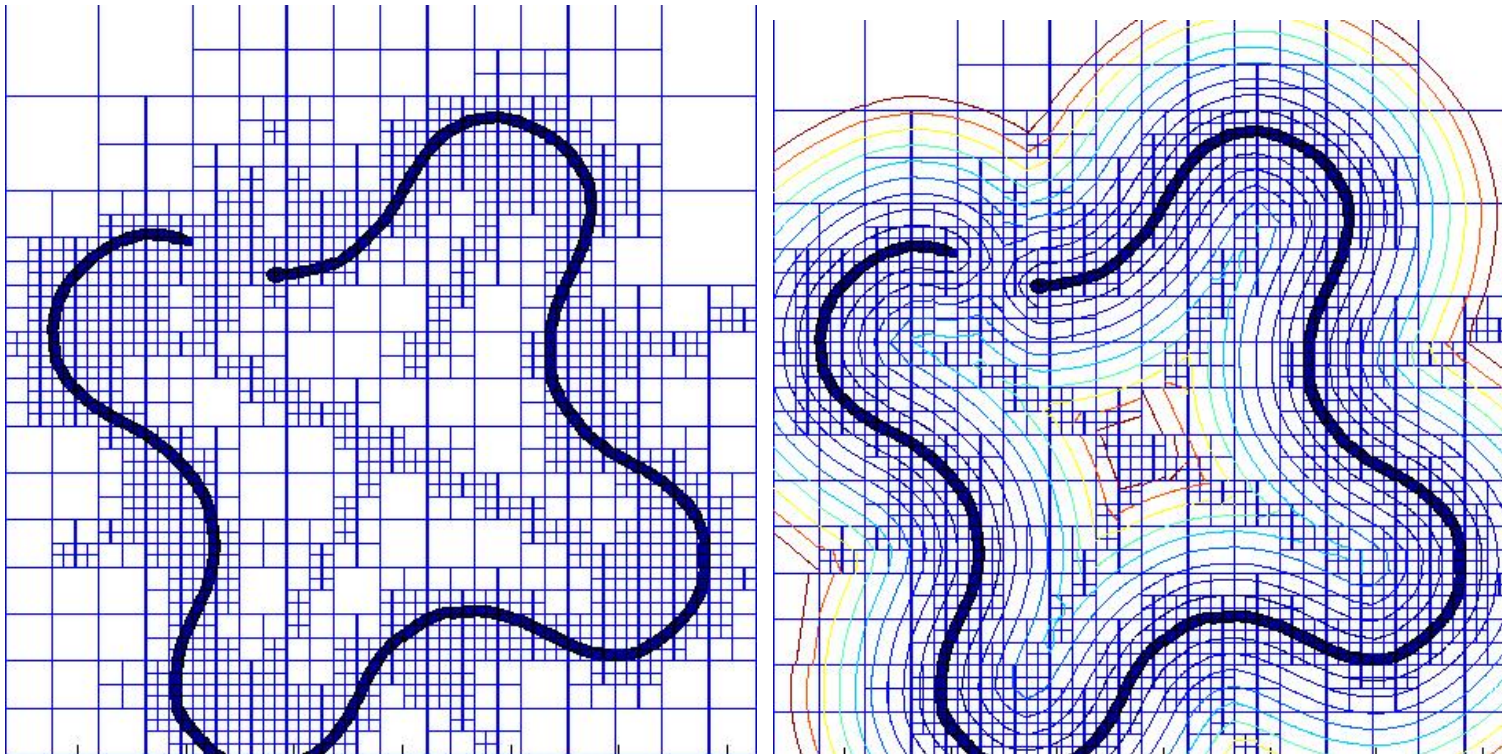
Second order approximant of the squared distance function to a **surface**

- The second order Taylor approximant F_d of the squared distance function to a surface at a point \mathbf{p} is expressed in the principal frame at the normal footpoint \mathbf{s} via

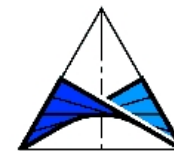


$$F_d(x_1, x_2, x_3) = \frac{d}{d - \rho_1} x_1^2 + \frac{d}{d - \rho_2} x_2^2 + x_3^2$$

- Adaptive data structure whose cells carry local quadratic approximants of the squared distance function

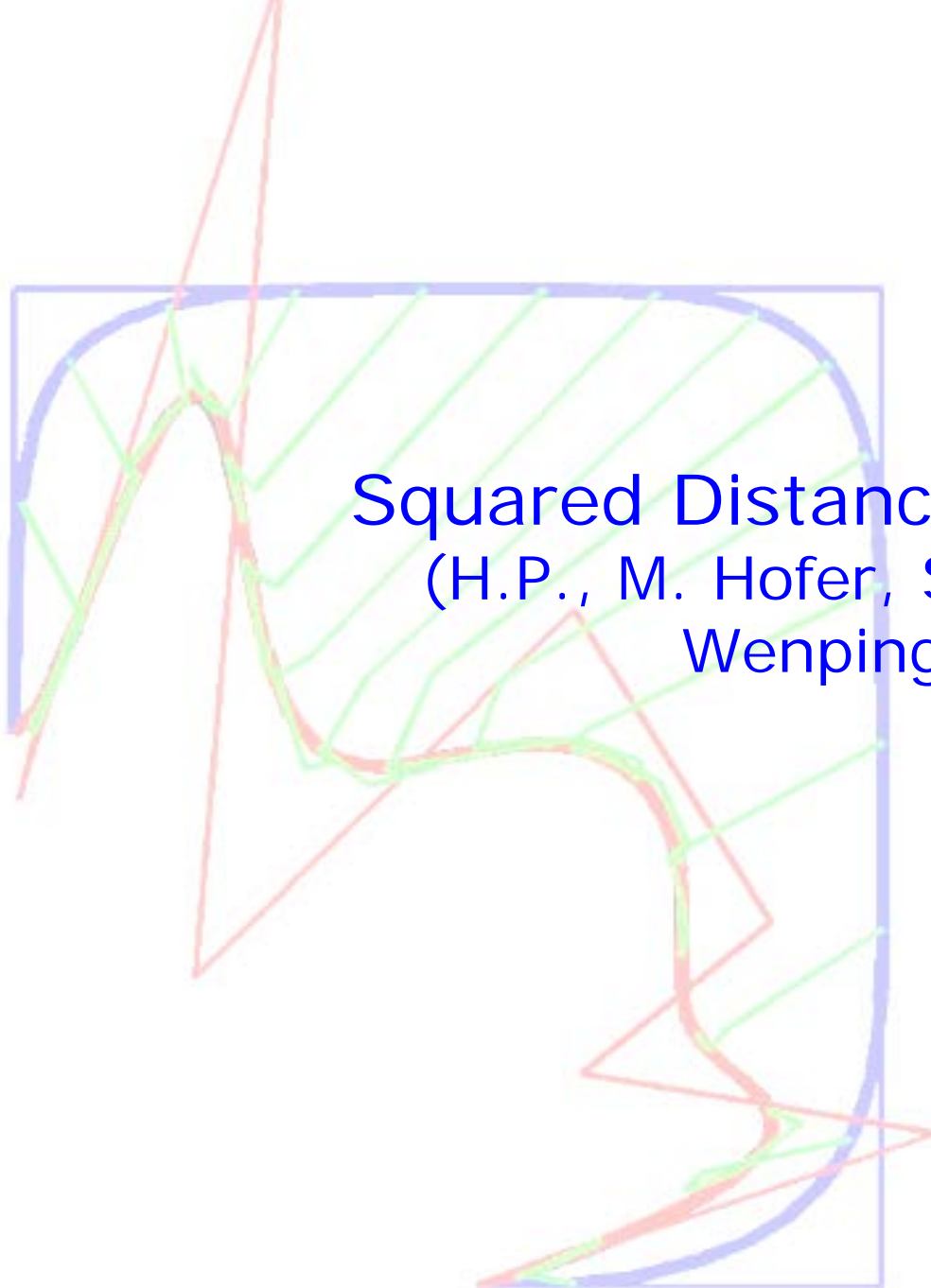


Curve and surface fitting

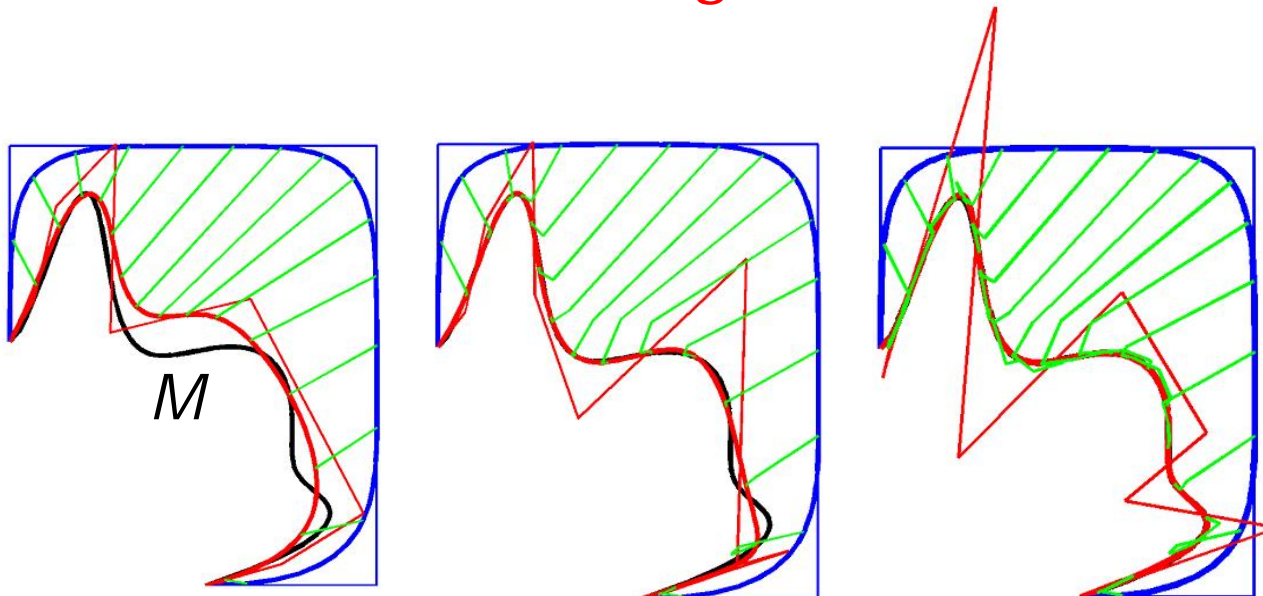


Squared Distance Minimization I

(H.P., M. Hofer, S. Leopoldseder,
Wenping Wang)

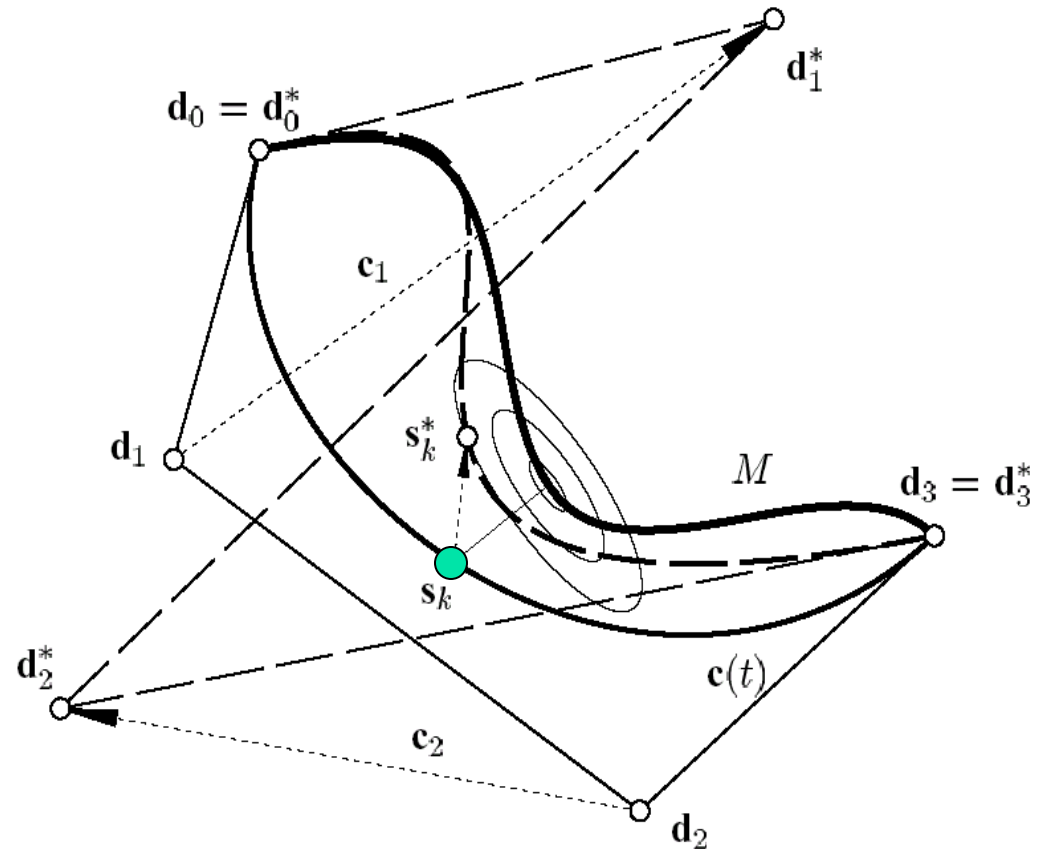


- B-spline curve or surface is iteratively deformed in the *squared distance field* of the model shape M (shape to be approximated: point cloud, mesh, implicit curve/surface,...)
- *Errors measured orthogonal to M*

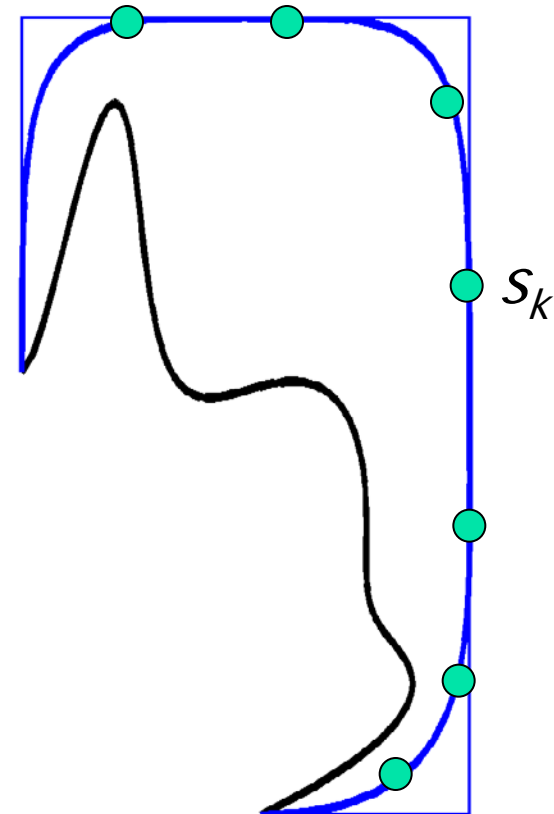


Iteration:

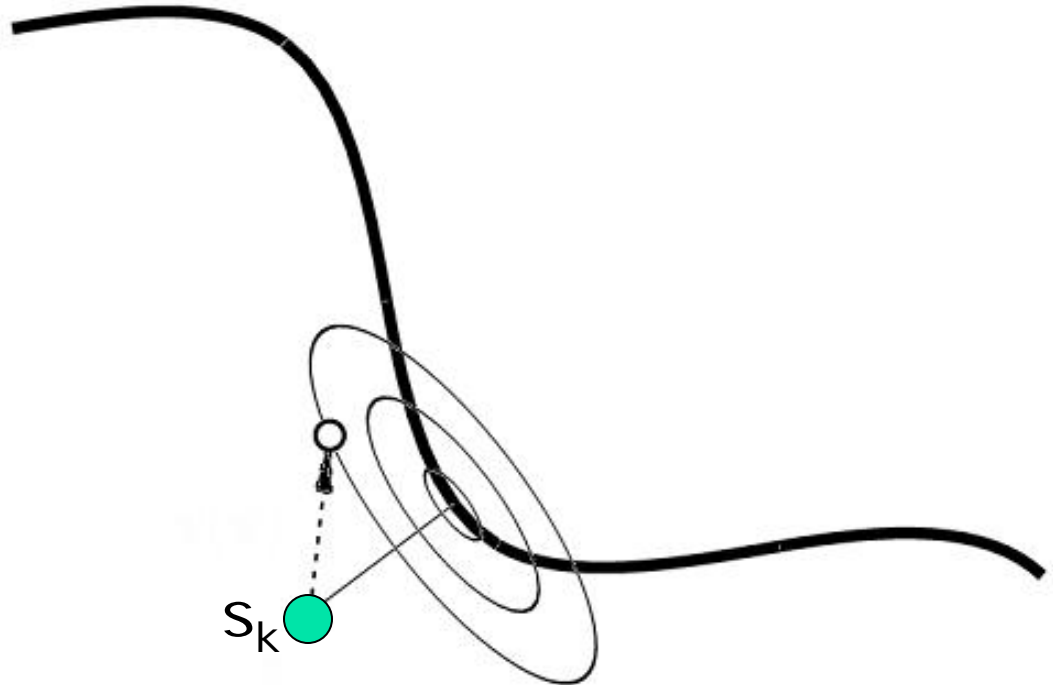
- evaluate current B-spline form at discrete number of 'sensor' points s_k
- compute local quadratic approximants of the squared distance field d^2 to M at s_k
- change control points such that sensors come closer to M (done with help of quadratic approximants to d^2)



- Choose an initial shape s (B-spline curve or surface)
- compute sufficient number of *sensor points* s_k on s (evaluation at chosen parameter values)
- Note: if control points are displaced with vectors c_i (knots and weights fixed), the new *sensor locations* s_k^* depend linearly on c_i



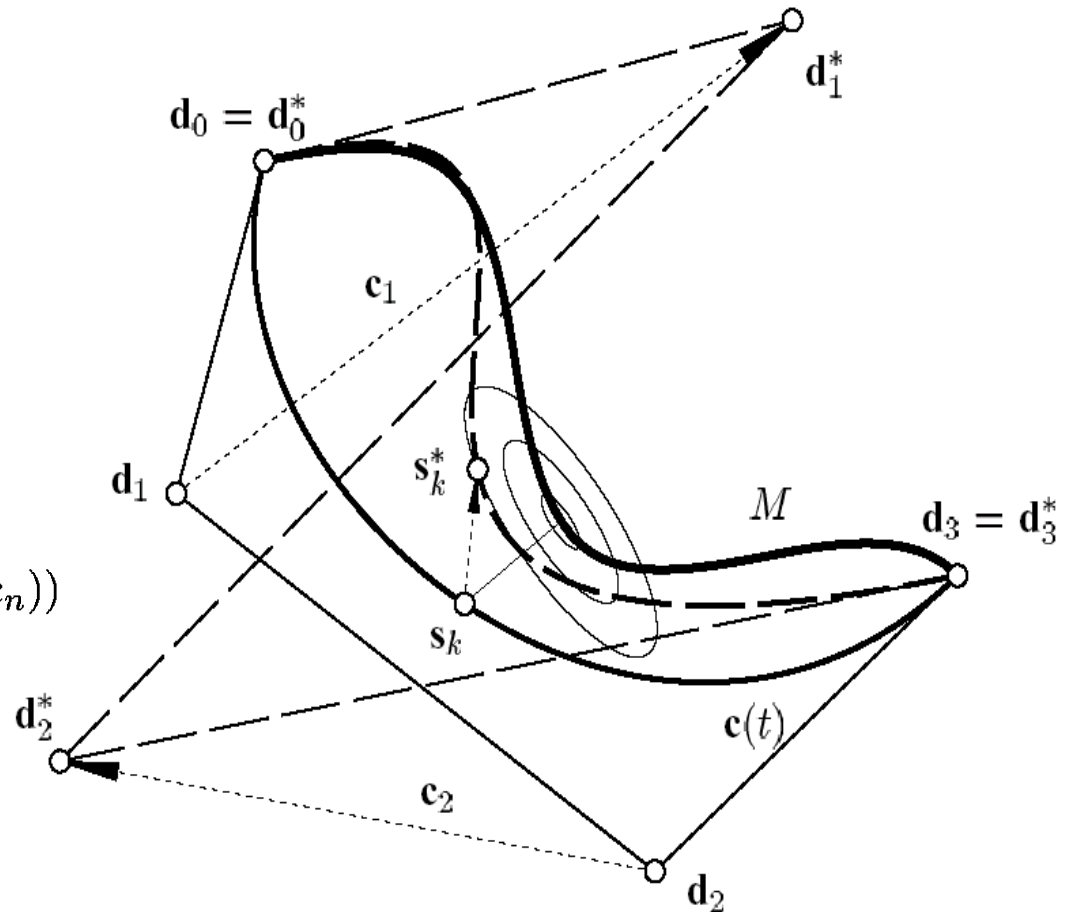
- At each sensor point s_k , compute a nonnegative local quadratic approximant to the squared distance field of model shape



- Compute displacement vectors \mathbf{c}_i of the control points such that sensors come closer to model shape by minimizing

$$F = \sum_{k=1}^N F_d^k(L_k(\mathbf{d}_1 + \mathbf{c}_1, \dots, \mathbf{d}_n + \mathbf{c}_n))$$

- This is a **quadratic function** in \mathbf{c}_i

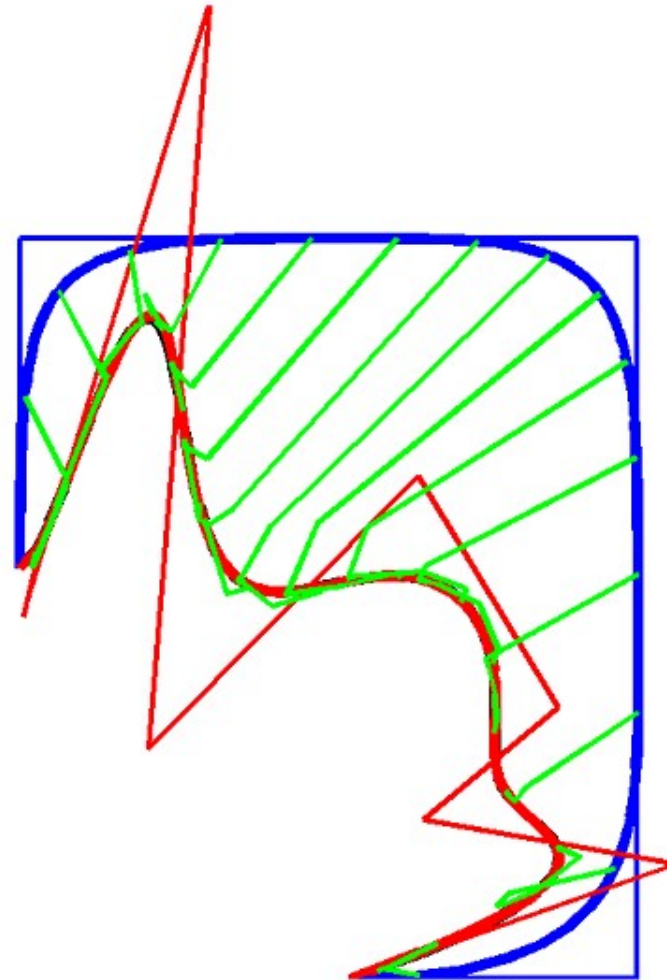


- For *regularization* and avoidance of overlappings/foldings of the final shape, an adequately weighted smoothing term F_s can be added to the functional, e.g. for surfaces:

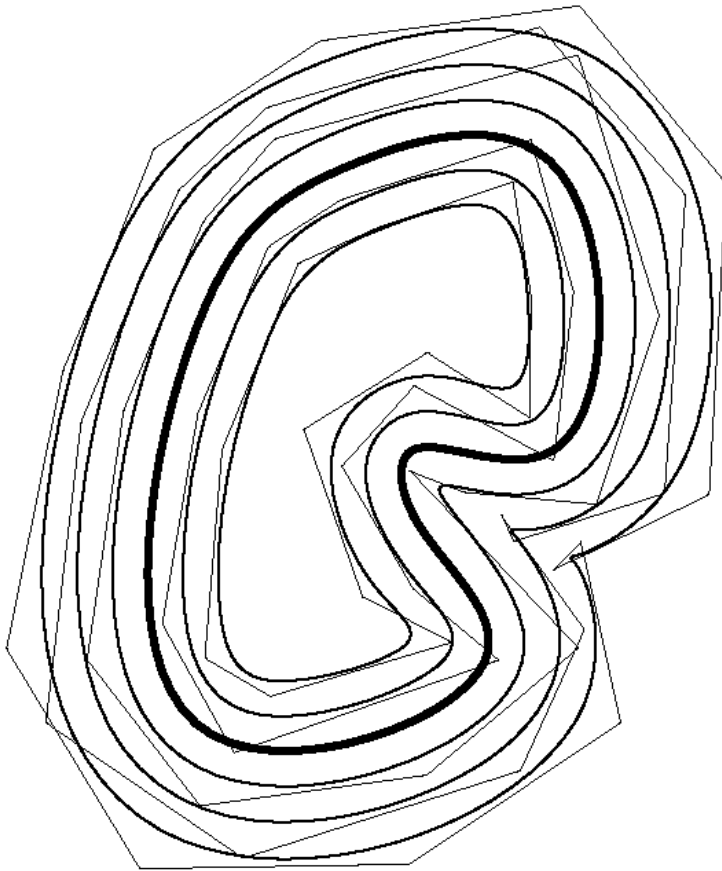
$$F_s = \int \int (s_{uu}^2 + 2s_{uv}^2 + s_{vv}^2) dudv$$

Amounts to minimization of quadratic function

- Displacement polygons of sensor points show that later moves are mainly in tangential direction
- Related to the computation of a good parameterization

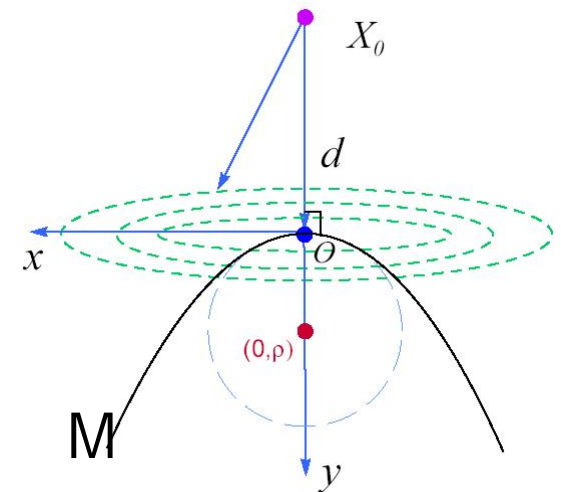


Example: Offset approximation



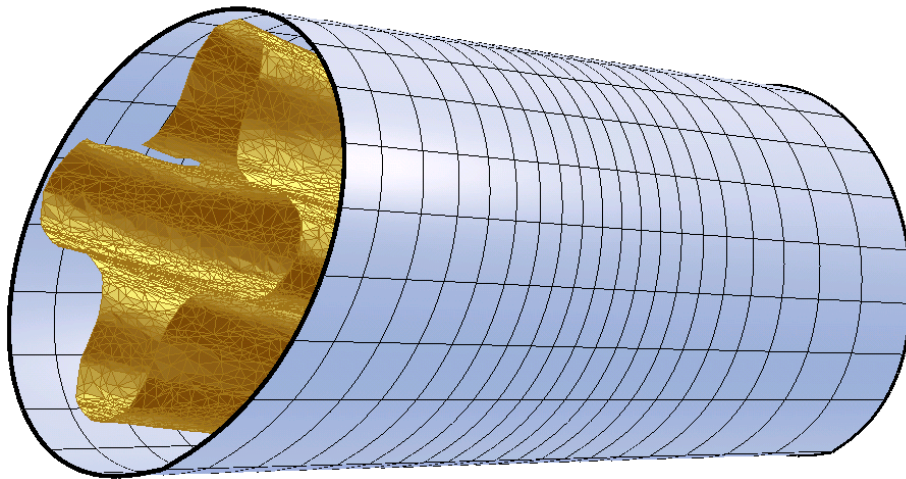
Sensor points do NOT move towards offset along normals of the progenitor curve

- **SDM** employs curvature at closest point to compute 2nd order approximation of squared distance function
- **TDM** uses only *squared distance to tangent at footpoint* (requires regularization and stepsize control)
- **PDM** (mostly used) employs squared distances to footpoint
- **TDM regularized with a PDM term** works very well

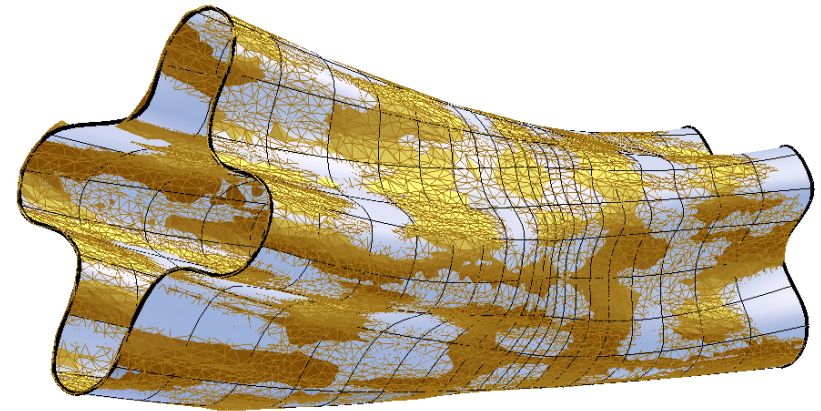


- **SDM**: Newton algorithm with *quadratic convergence*
- **TDM**: Gauss-Newton iteration for nonlinear least squares problem; *quadratic convergence for zero residual problem* and good initial position; requires **regularization** (add multiple of identity matrix to approximate Hessian or **add a PDM term with low weight**) and **step size control**
- **PDM**: only linearly convergent and prone to be trapped in local minimizer

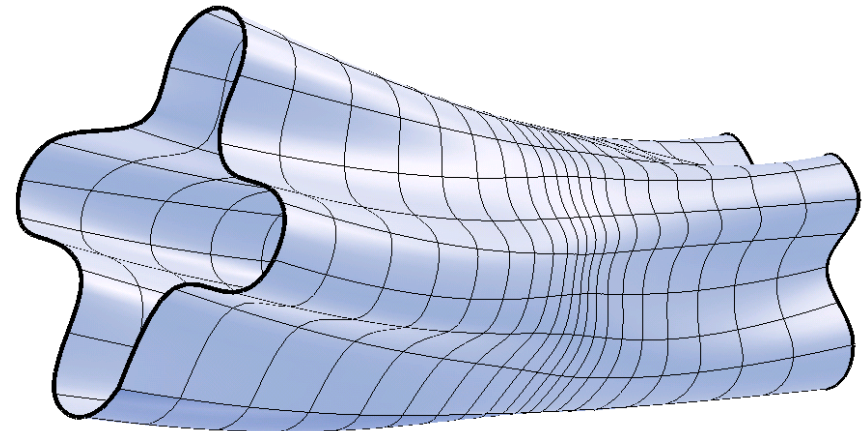
Example: NURBS surface approximation

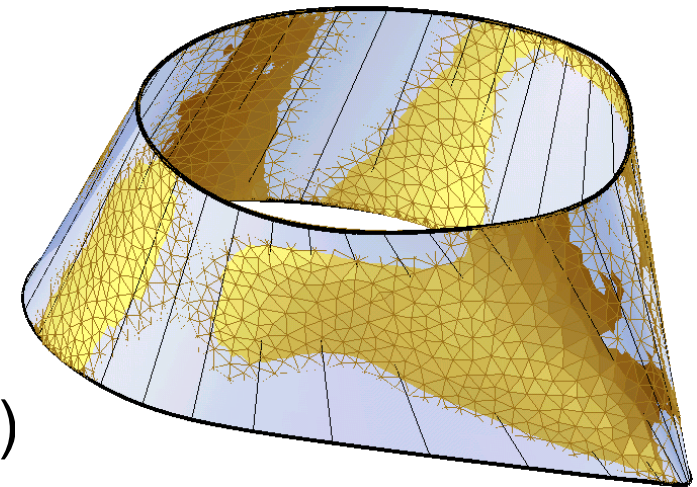
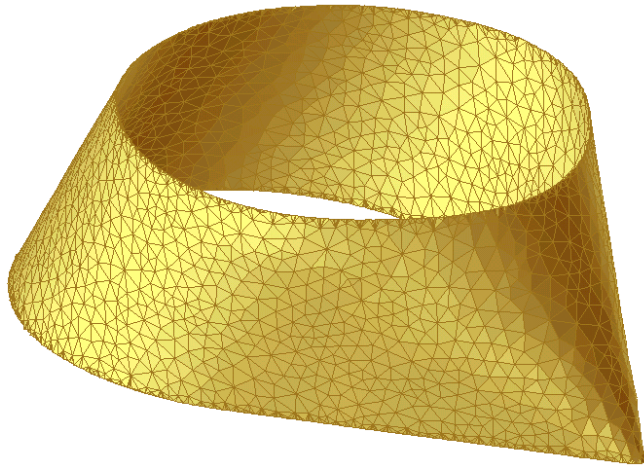
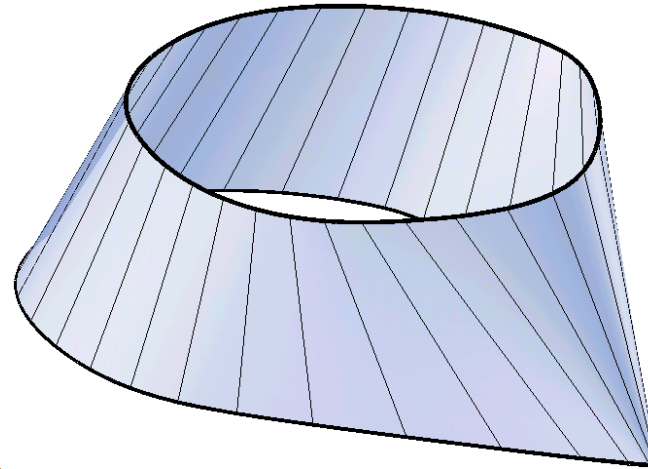


Model surface (gold) and
initial position of active
NURBS surface (blue)



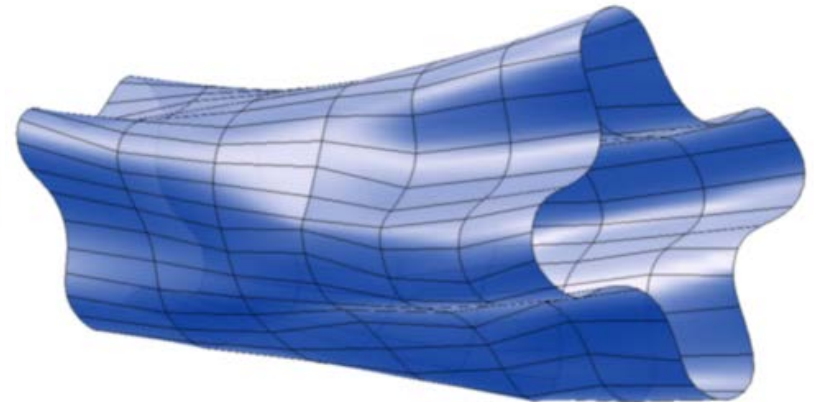
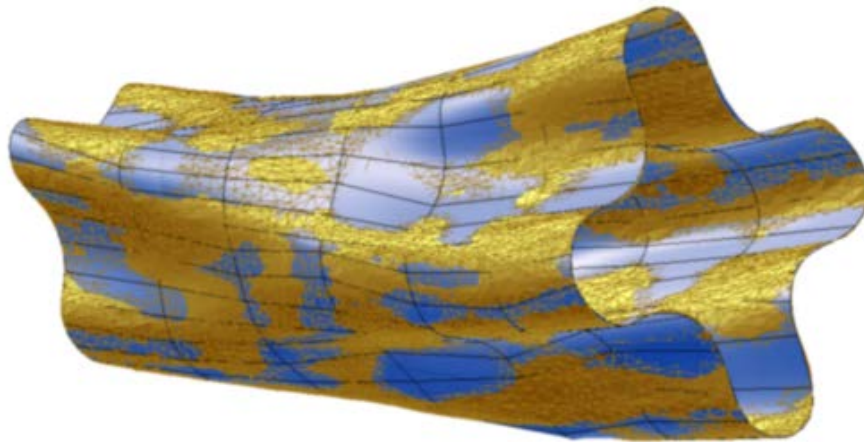
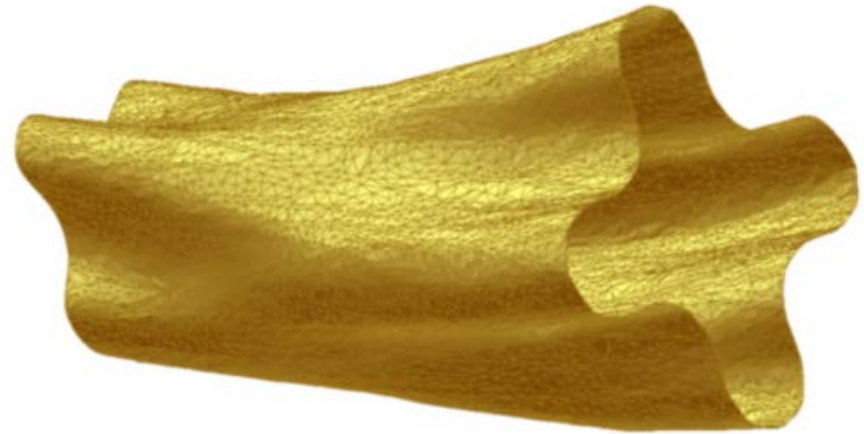
Final position of active
NURBS surface





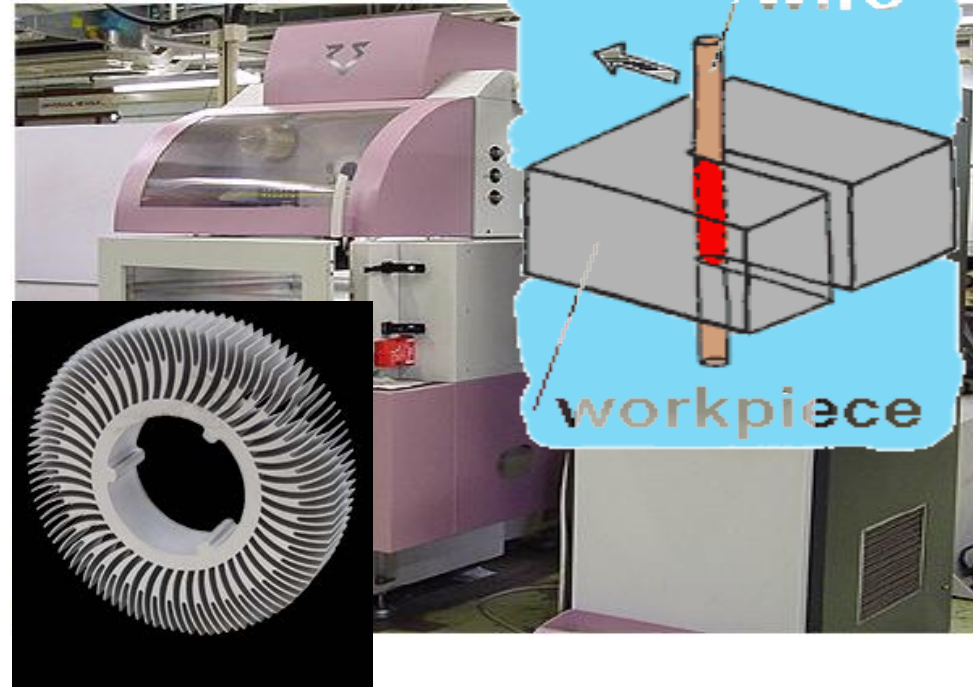
- active B-spline surface of degree $(1, n)$

Approximation by a piecewise ruled surface





hot wire cutting
(styrofoam)

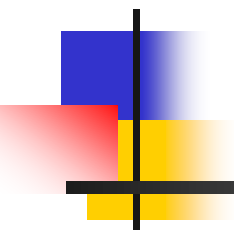
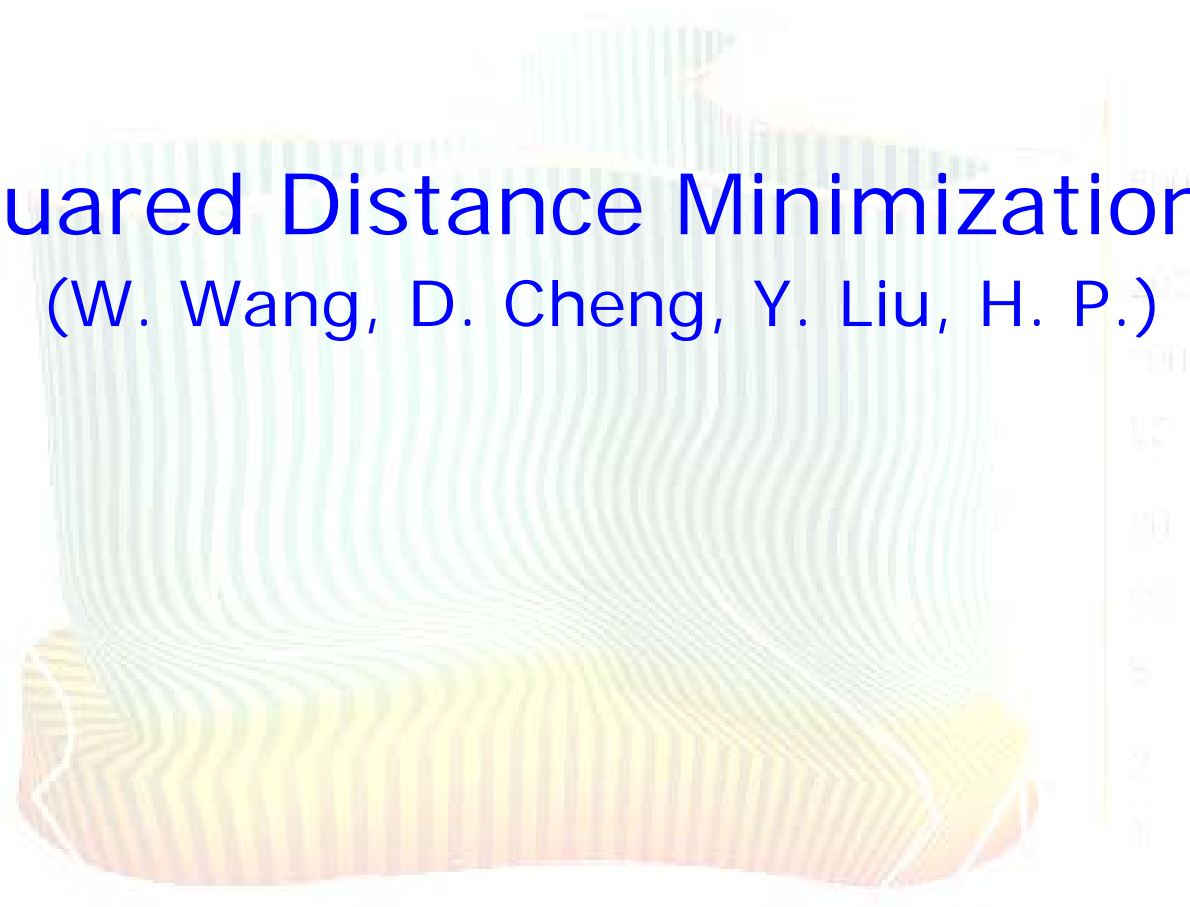


wire EDM
(metal)

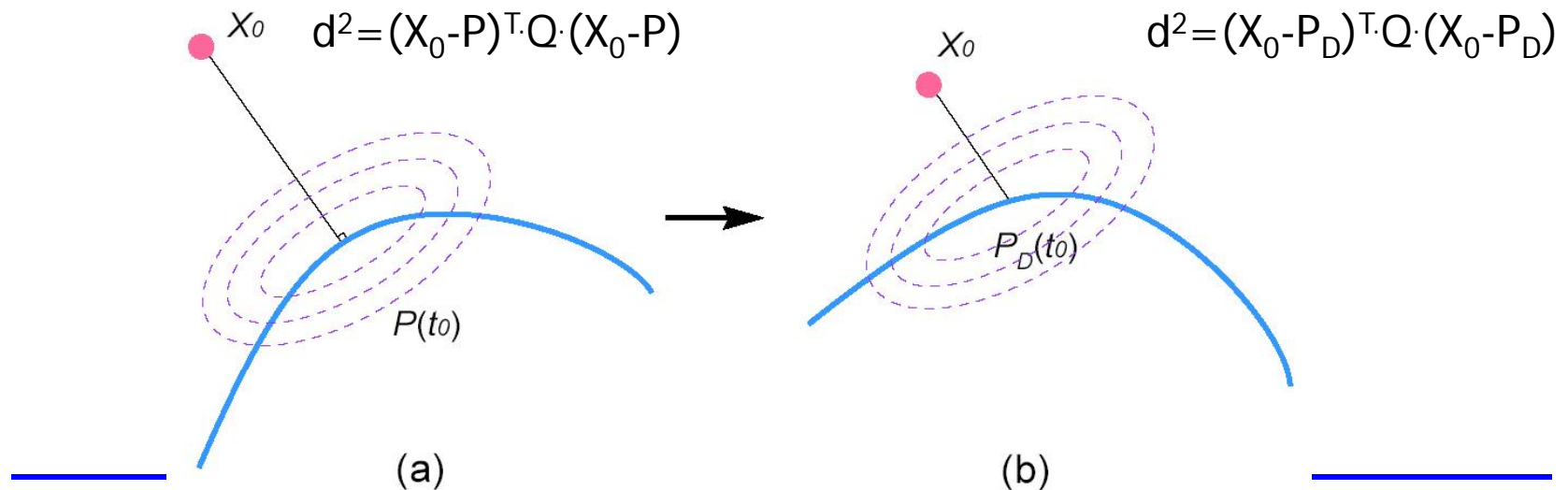


Squared Distance Minimization II

(W. Wang, D. Cheng, Y. Liu, H. P.)

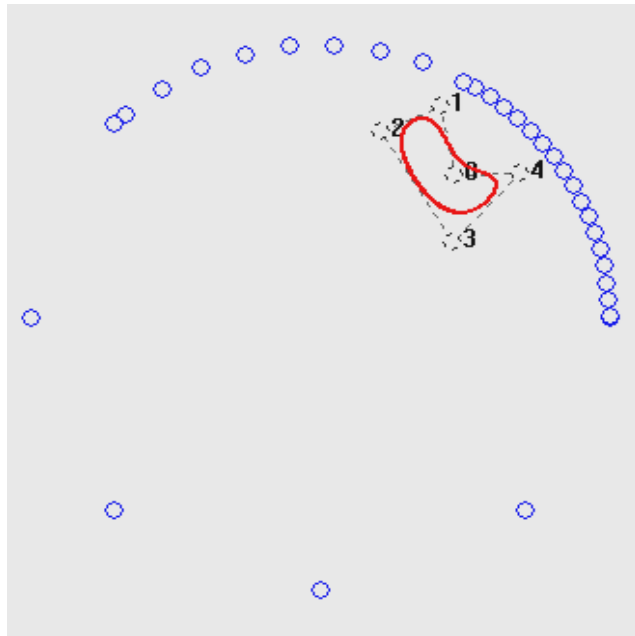


- *Error measurement orthogonal to the active curve/surface*
- Squared distance field (quadratic approximants) attached to the moving curve/surface
- Quasi-Newton method (neglects rotation of tangent and change of curvature)

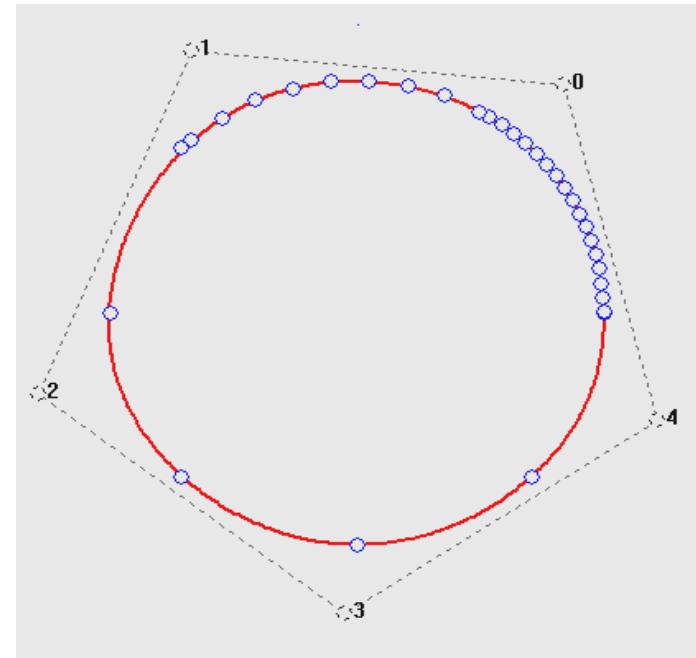


- TDM: measures the error via squared distance to tangent at closest point; requires regularization and step size control
- PDM (standard approach): measures error via squared distance to closest point
- **TDM regularized with a PDM term** (small weight) works very well

Example 1: Non-uniform data

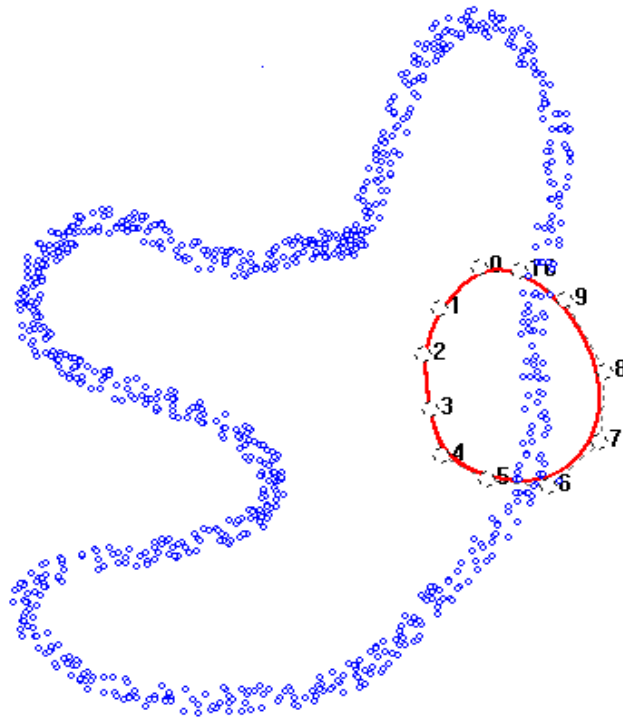


Initial position

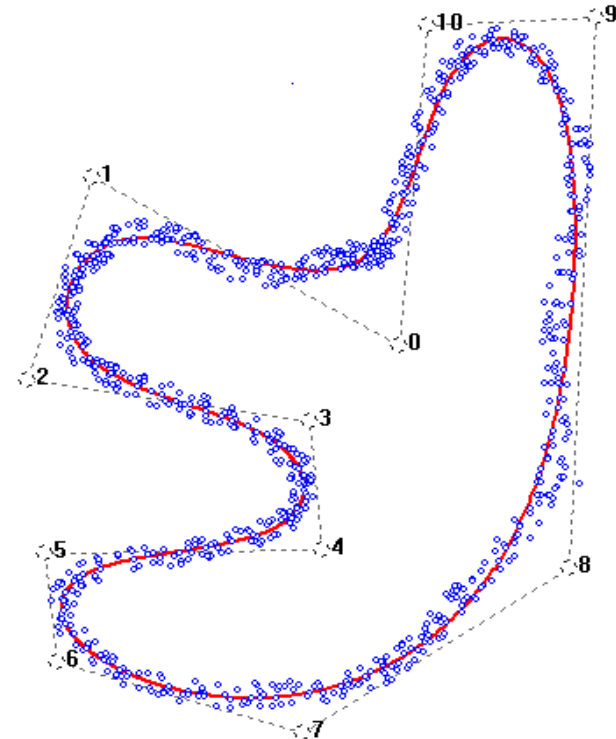


SDM after 8 iterations

Example 2: Thick point cloud

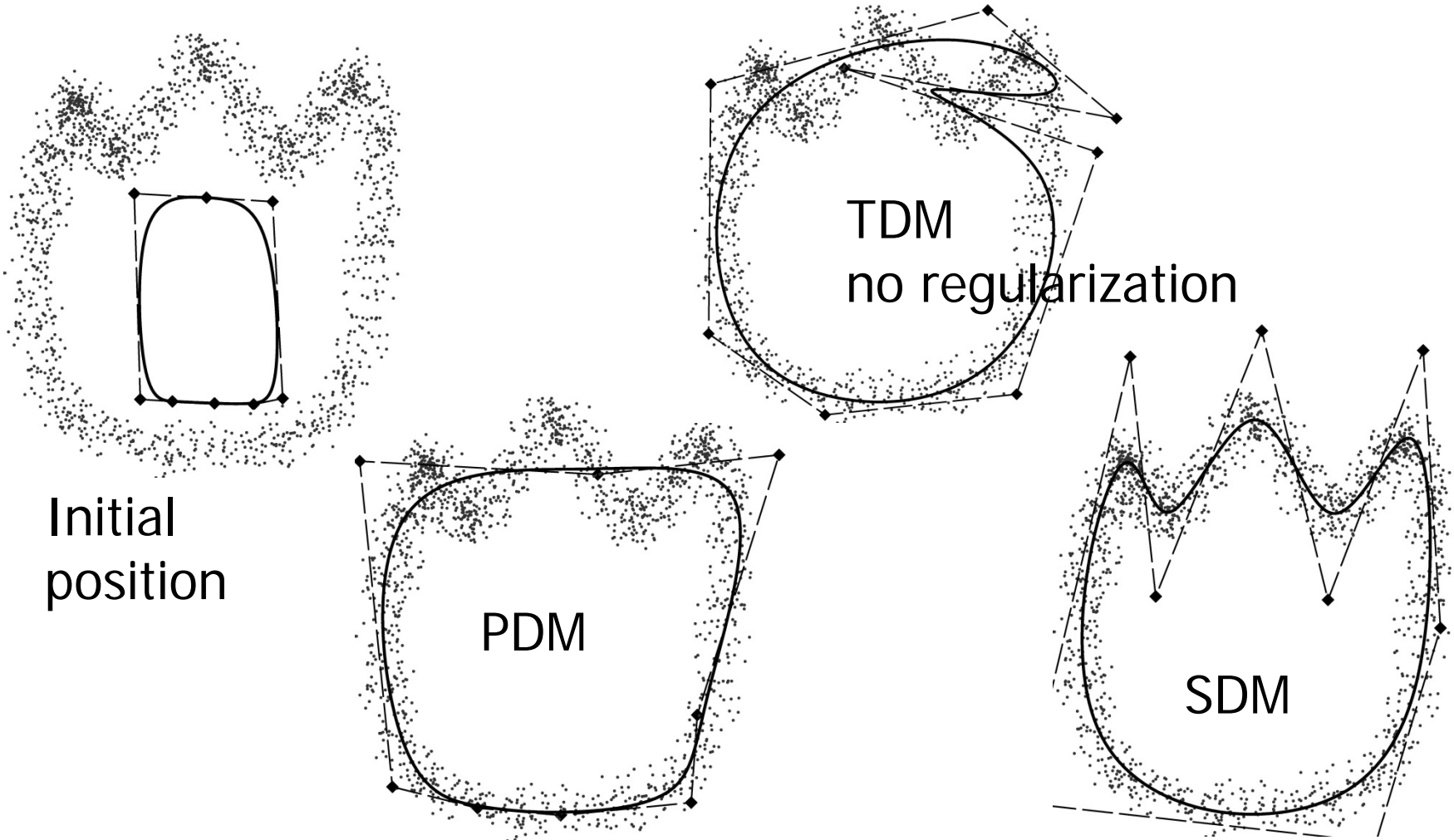


Initial position

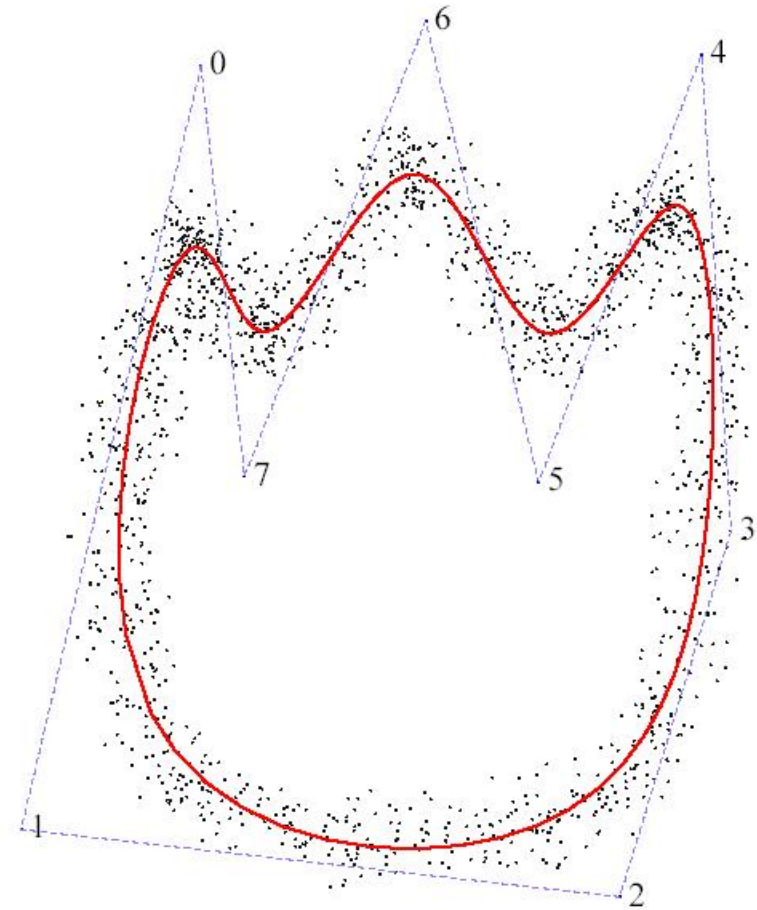
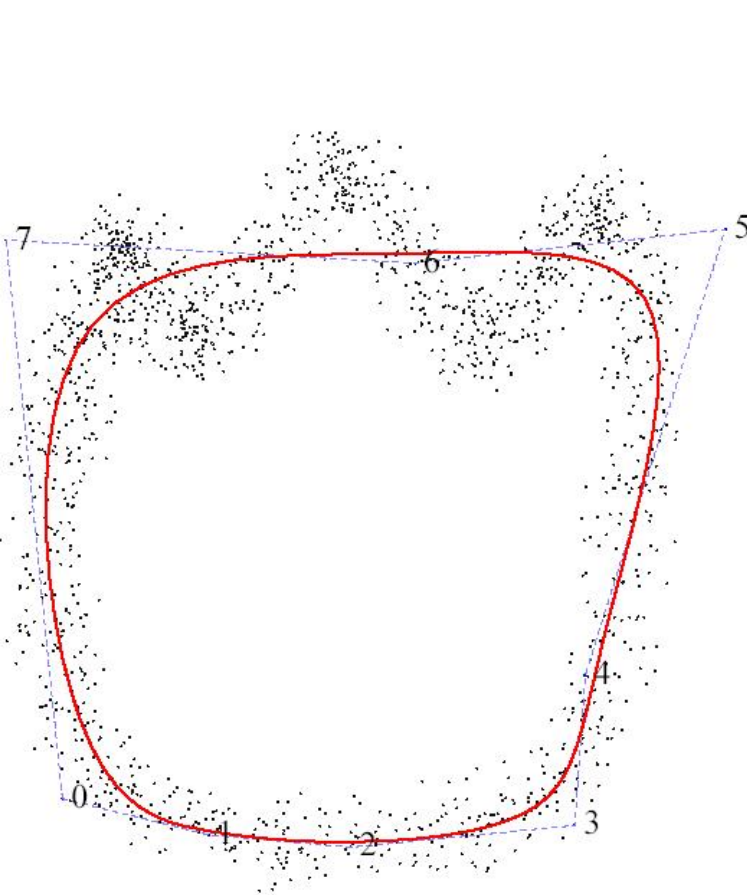


SDM after 31 iterations

Noisy data set; 50 iterations



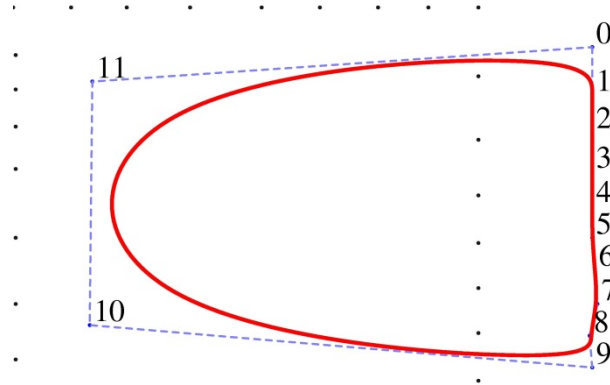
Comparison of PDM and SDM after 50 iteration steps



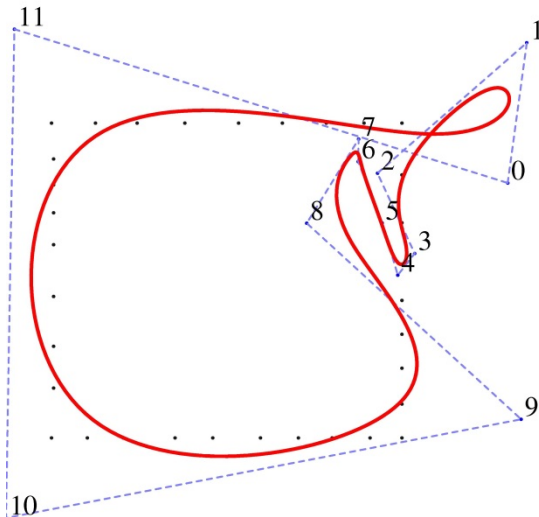
■ Point Distance Min. (PDM)

■ SDM

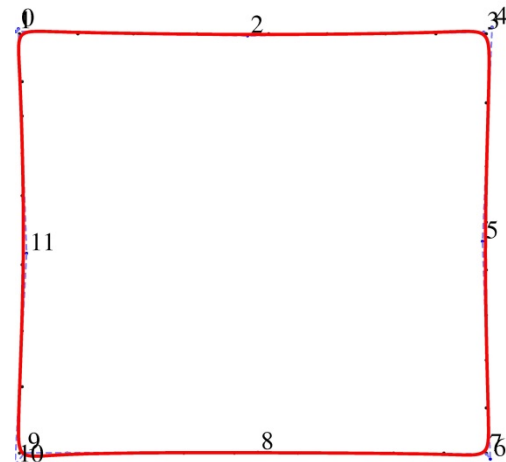
Comparison: PDM vs. SDM



Target shape and Initial position

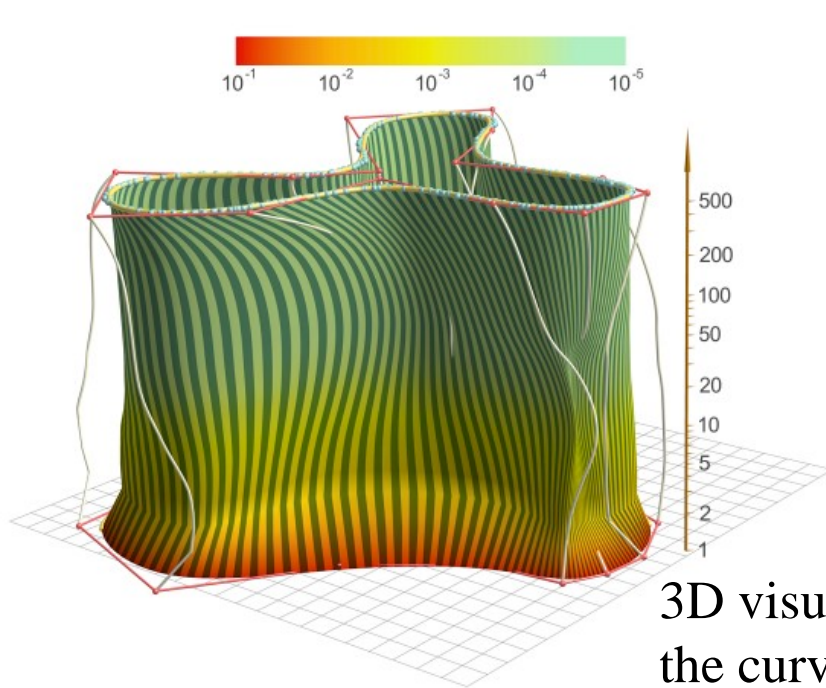


PDM after 20 iterations



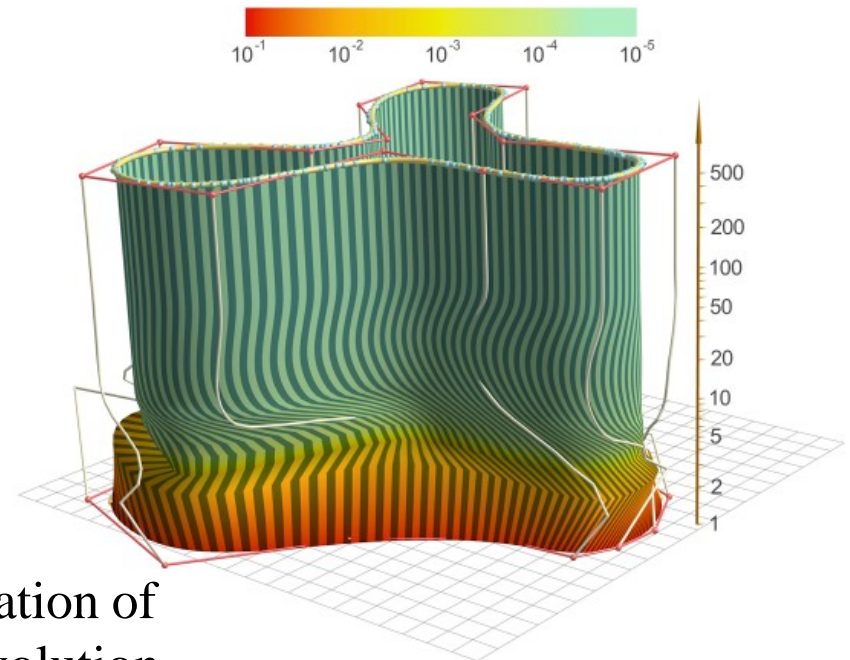
SDM after 20 iterations

- Standard method (PDM): linear convergence
- SDM: quadratic convergence



PDM

3D visualization of
the curve evolution



SDM