



A general surface reconstruction method for post-processing of topology optimisation results

Giulia Bertolino, Giulio Costa, Marco Montemurro, Nicolas Perry and Franck Pourroy

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Context and scientific objectives

Context

Topology optimisation:

- Optimal distribution of material in a prescribed domain
- Minimise an objective/cost function + meet optimisation constraints

Results of topology optimisation strategy:

- Density field described by element-wise format
- Need to obtain smooth surfaces
- How is it possible to obtain CAD compatible entity?







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Surface Reconstruction strategy for genus 0 open surfaces

Proposed strategy: main ingredients



Projection method [Piegl,1995] - open, genus 0, not folded

Mercator's projection method [Rahi,2007] - closed, genus 0





Global conformal method [Gu,2003] - closed, genus N

Shape preserving method [Floater, 1997] - open, genus 0, folded



Shape preserving method: capabilities and main features



Surface fitting: Optimisation strategy



Problem formulation and numerical aspects: genetic optimisation

Part A

Objective function

min
$$f(\mathbf{x}) = \left(\sum_{k=0}^{n_{tp}} ||S(u_k, v_k) - Q_k||^2 + \lambda J\right)^{\frac{1}{r_1 + r_2}}$$
 such that:

Distance between BSpline entity and target points

Thin-plate spline energy functional^{Floater,2000]}: smoothing term

$$J = \int_{a_1}^{b_1} \int_{a_2}^{b_2} S_{uu}^2 + 2S_{uv}^2 + S_{vv}^2 \, du \, dv$$

Constraint function:

Non singularity of Basis Functions (BF) matrix

 $g_1(\mathbf{x}) = \dim(BF) - \rho(BF) \rightarrow BF = [N_u N_v]^T [N_u N_v] + \lambda E$

Dimension of the basic functions matrix

Rank of the matrix of the basic functions matrix

Smoothing matrix

Design variables



Continuous variables: $U_{p1+2,...,}U_{p1+r1+2}, V_{p2+2,...,}V_{p2+r2+2} \rightarrow Knot vector$ components

Design space dimension = $4 + r_1 + r_2$

Discrete variables values affect the dimension of the Continuous variables module.

Problem formulation and numerical aspects: genetic optimisation

Part A

GA ERASMUS^[Montemurro,2018] capabilities:

- <u>Reproduction among individuals</u>: crossover and mutation operations
- <u>Reproduction among species</u>: on individuals with different number of chromosomes
- <u>Penalisation</u>: Automatic Dynamic Penalisation (ADP)
 - Automatically and adaptively updating the coefficients of penalisation
 - Preventing infeasible solutions
 - Efficient exploration of the boundary of the feasible domain



Part B

Design variables (only continuous)

$$\mathbf{X} \begin{cases} \mathbf{U} = \{0, \dots, 0, u_{p+1}, \dots u_n, 1, \dots, 1\} \\ \mathbf{V} = \{0, \dots, 0, v_{q+1}, \dots v_m, 1, \dots, 1\} \\ \mathbf{W} = \begin{pmatrix} w_{11} & \cdots & w_{1n_2} \\ \vdots & \ddots & \vdots \\ w_{n_11} & \cdots & w_{n_1n_2} \end{pmatrix} \end{cases}$$

Objective function

$$\min f(\mathbf{x}) = \sum_{k=0}^{n_{tp}} \|S(u_k, v_k) - Q_k\|^2 + \lambda J \text{ such that:}$$

Constraint function

$$g_1(\mathbf{x}) = \dim(BF) - \rho(BF) \rightarrow BF = [N_u N_v]^T [N_u N_v] + \lambda E$$

Non trivial KV components

• Numerical evaluation of $abla f(\mathbf{x})$ respect to KV

Weights

• Analytical evaluation of $\nabla f(\mathbf{x})$ respect to weights

Focus on the analytical CPs evaluation



Numerical results: 1st benchmark

	Design variables	Lower - Upper bounds
GA optim	Degree (p)	2 – 6
	N° of KV's components (r)	1 – 17
	KV's components values (U, V)	0.001 - 0.999
Grad optim	Alpha, Beta (KV's components values)	0.001 - 0.999
	Weights (w)	0-10



Results of the Surface Reconstruction strategy

Design space

Optimised values		Objective function value	
Degree (p ₁ -p ₂)	5 – 5	GA phase	0.0944
N° of KV's components ($r_1 - r_2$)	1-1	Grad KV	0.0860
KV's components values (U, V)	0.374 – 0.599	Grad KV + Weights	0.0844

Optimised design variables at the end of the Surface Reconstruction algorithm and objective function value evolution along the different phases

Numerical results: 2nd benchmark

	Design variables	Lower - Upper bounds
GA optim	Degree (p)	2 – 6
	N° of KV's components (r)	16 – 35
	KV's components values (U, V)	0.001 – 0.999
Grad optim	Alpha, Beta (KV's components values)	0.001 - 0.999
	Weights (w)	0-10

Design space

Optimised values		Objective function value	
Degree (p ₁ -p ₂)	2 – 2	GA phase	0.963576
N° of KV's components $(r_1 - r_2)$	19 – 19	Grad KV	0.954653
		Grad KV + Weights	0.892313

Optimised design variables at the end of the fitting algorithm and objective function value evolution along the different phases



Results of the Surface Reconstruction strategy

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Poly-patches strategy for genus N surfaces (open and closed)

Poly-patches strategy for genus N surfaces (open and closed)

Aim: Application of the Surface Reconstruction strategy to surfaces (open and closed) with holes (genus > 0)

Parameterisation

Strategy:



Adjacent patches have same parameters along boundary

Proper roto-translation of patches according to the global reference system



Automatic calculation of NURBS parameters Automatic imposition of CO and C1 continuity condition between patches





Numerical result

	Design variables	Lower - Upper bounds
GA optim	Degree (p)	2 – 6
	N° of KV's components (r)	4 – 20
	KV's components values (U, V)	0.001 - 0.999
Grad optim	Alpha, Beta (KV's components values)	0.001 - 0.999
	Weights (w)	0 - 10

Design space

Optimised values	
Degree (p ₁ -p ₂)	2 – 2
N° of KV's components $(r_1 - r_2)$	8-8

Optimised design variables at the end of the fitting algorithm.

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Conclusions

> Automatic optimisation of approximation surface parameters

Reduction of tasks delegated to the user

Perspectives

- Mapping methods for genus > 0 surfaces
- > Automatic segmentation of the triangulation (STL file)
- Integration of Tspline entities in the surface fitting



Thank you for your attention

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Shape preserving method^[Floater,1997]



Thin-plate spline energy^[Floater,2000]

Adding a smoothing term in the surface approximation of unstructured data aims to find a unique solution.

$$f(\mathbf{x}) = \sum_{k=0}^{n_{tp}} \|S(u_k, v_k) - Q_k\|^2 + \lambda J \qquad \begin{cases} J = \int_{a_1}^{b_1} \int_{a_2}^{b_2} S_{uu}^2 + 2S_{uv}^2 + S_{vv}^2 \, du \, dv \rightarrow \text{simple thin plate energy functional} \\ \lambda \rightarrow \text{constant measuring the trade off between approximation and smoothing} \end{cases}$$

Find the minimum \rightarrow normal equations

$$\frac{\partial f(\mathbf{x})}{\partial P_{ij}} = \left([N_u N_v]^T [N_u N_v] + \lambda E \right) [P] - [N_u N_v]^T [Q] = 0$$

Where *E* is a $(n_1n_2) \ge (n_1n_2)$ matrix whose elements are:

$$E_{ijrs} = A_{ijrs} + 2B_{ijrs} + C_{ijrs} \qquad \begin{cases} A_{ijrs} = \int_{a_1}^{b_1} N_i''(u) N_j''(u) du \int_{a_2}^{b_2} N_j(v) N_s(v) dv \\ B_{ijrs} = \int_{a_1}^{b_1} N_i'(u) N_j'(u) du \int_{a_2}^{b_2} N_j'(v) N_s''(v) dv \\ C_{ijrs} = \int_{a_1}^{b_1} N_i(u) N_j(u) du \int_{a_2}^{b_2} N_j''(v) N_s''(v) dv \end{cases}$$

And λ is:

$$\lambda = \frac{\|([N_u N_v]^T [N_u N_v])^2\|}{\|E^2\|}$$