



# Numerical Algorithms

Winter semester 2018/2019  
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**Exercise sheet 10.** Submission on **Tuesday, 2019-01-15, after the lecture.**

**Exercise 31.** (B-spline derivatives)

We have already looked into B-splines but not yet at their derivatives.

a) Show that

$$\frac{d}{d\xi} N_{i,p}(\xi) = p \left( \frac{N_{i,p-1}(\xi)}{\xi_{i+p} - \xi_i} - \frac{N_{i+1,p-1}(\xi)}{\xi_{i+p+1} - \xi_{i+1}} \right).$$

*Hint:* Recursion formula and induction starting with  $p = 1$ .

b) Using a), compute the derivative of the B-spline curve

$$\frac{d}{d\xi} C(\xi) = \frac{d}{d\xi} \sum_{i=1}^n N_{i,p}(\xi) B_i.$$

(2 Points)

**Exercise 32.** (Minimal degree condition)

A not yet posed or answered is how  $p_T, p_e$  should be related in a higher order FEM with  $V-E-C$  shape functions. One strategy for this is the minimal degree condition.

Let  $T$  be an element of some triangulation  $\tau$  of the domain  $\Omega$  for some PDE. For each edge  $e$  of the triangulation it should hold that

$$p_e = \min\{p_T | e \text{ is edge of } T\}.$$

For the depicted mesh, choose the degrees  $p_e$  of each edge  $e$  such that the minimal degree condition is fulfilled

$p = 1$	$p = 3$	$p = 1$
$p = 2$	$p = 4$	$p = 2$
$p = 1$	$p = 3$	$p = 1$

(2 Points)

**Exercise 33.** (sup – inf and  $n$ -width)

In the lecture the notion of classification of best approximation of a set  $Y \subset X$  with some normed space  $(X, \|\cdot\|_X)$  was introduced with the Kolmogorov  $n$ -width

$$d_n := \inf_{E_n \subset X, \dim E_n \leq n} \sup_{u \in Y} \inf_{v_n \in E_n} \|u - v_n\|_X .$$

If  $X, (\cdot, \cdot)_X$  is a Hilbert space and  $Y = TH$  with  $H, (\cdot, \cdot)_H$  being another Hilbert space that can be compactly embedded in  $Y$  via the operator  $T$

$$H \xrightarrow{C} X , \quad T : H \rightarrow X , \quad \|Tx\|_X \leq C\|x\|_H , \quad T \text{ compact} .$$

these quantity allows for a more practical expression. With  $H$  being a linear space it is reasonable to exclude multiplicative constants leading to

$$\begin{aligned} Y &:= \{x \in H : \|x\|_H = 1\} = \{Tx : x \in H, \|x\|_H = 1\} \subset X , \\ \Psi(E_n) &:= \sup_{u \in Y} \inf_{v_n \in E_n} \|u - v_n\|_X , \\ d_n &= \inf_{E_n \subset X, \dim E_n \leq n} \Psi(E_n) . \end{aligned}$$

$\Psi$  is called a sup – inf of  $E_n$  with respect to the  $\|\cdot\|_X$  approximation of  $Y$ . For convenience the embedding  $T$  is dropped from notation.

In this special case of a compact embedding these quantities are related to the generalized eigenvalue problem

$$(u_k, v)_X = \lambda_k (u_k, v)_H , \quad \forall v \in H$$

with eigenpairs  $(\lambda_k, u_k), k = 0, \dots$  and eigenvalues

$$\lambda_0 \geq \lambda_1 \geq \lambda_2 \dots > 0 , \quad d_n = \sqrt{\lambda_n}$$

There exists a similar generalized eigenvalue problem for the computation of  $\Psi(E_n)$ .

a) Show that for  $u \in H, E_n \subset X$  and  $\dim E_n = n$  we have

$$\inf_{v_n \in E_n} \|u - v_n\|_X \leq \|u\|_H \Psi(E_n)$$

b) Assume that  $H \subset Z$  with  $Z$  being another Hilbert space with a basis  $\varphi_i, i \in I = \{1, \dots, N\}$ . Using the coefficients of a  $H$  basis with respect to  $\varphi_i$ , give the matrix representation of the generalized eigenvalue problem for the computation of  $d_n$ .

(4 Points)

**Exercise 34.** (Moving Least Squares)

Given data  $x_i, f_i, i = 1, \dots, N$  and an approximation space  $P = \text{span}\langle \varphi_i \rangle$  the natural extension of least squares is moving least squares (MLS) approximation.

For a locally supported non-negative function  $W$ , often referred to as *window function* or *weight function*, the pointwise *moving least squares energy* of an approximation  $\pi$  is given as

$$J_x(\pi) = \sum_{i=1}^N W(x - x_i)(f_i - \pi(x_i))^2 = \sum_{W(x-x_i)>0} W(x - x_i)(f_i - \pi(x_i))^2 .$$

The MLS approximation  $\pi$  of the data  $x_i, f_i$  is given pointwise via the minimizers  $\pi_x \in P$  of  $J_x$

$$\pi(x) = \pi_x(x) .$$

Using the basis  $\varphi_i$  and a representation  $\pi_x = \sum_i u_{x,i} \varphi_i, u_x = (u_{x,i})_i$  we can compute the solutions  $\pi_x$  as

$$\begin{aligned} G_x u_x &= f_x , \\ (G_x)_{k,l} &:= \sum_{W(x-x_i)>0} \varphi_k(x_i) W(x - x_i) \varphi_l(x_i) , \\ f_x &:= \sum_{W(x-x_i)>0} f_i W(x - x_i) \varphi_l(x_i) . \end{aligned}$$

a) Expand and compute

$$\frac{d}{dx} \pi(x) .$$

b) The case  $P = \text{span}\langle 1 \rangle$  is called *Shepard* approximation. For this case, compute  $\pi, \frac{d}{dx} \pi$ .

(4 Points )

**Programming exercise 11.** (Refined Spinoraptor)

Consider the B-Spline curve defined through the knot vector and control points given in programming exercise 10. Since the resolution of the model is rough, a further analysis requires the refinement of the model.

The analogue of *h-refinement* from finite elements for B-Splines (and B-Spline curves and surfaces and volumes) is *knot* insertion. Similarly to *h-refinement*, this can be done without changing the curve geometrically or parametrically.

Given a knot vector

$$\Xi = (\xi_1, \dots, \xi_{n+p+1})$$

and let  $\hat{\xi} \in [\xi_k, \xi_{k+1})$  be a desired new knot. The  $n + 1$  basis functions  $\hat{N}_{i,p}$  are formed recursively using the new knot vector

$$\hat{\Xi} = (\xi_1, \dots, \xi_k, \hat{\xi}, \xi_{k+1}, \dots, \xi_{n+p+1})$$

of length  $n + p + 2$ .

The new  $n + 1$  control points  $\hat{B}_i, 1 \leq i \leq n + 1$  are formed from the original control points  $B_i, 1 \leq i \leq n$  by

$$\hat{B}_i = \alpha_i B_i + (1 - \alpha_i) B_{i-1}, \quad \alpha_i = \begin{cases} 1, & 1 \leq i \leq k - p, \\ \frac{\hat{\xi} - \xi_i}{\xi_{i+p} - \xi_i}, & k - p + 1 \leq i \leq k, \\ 0, & k + 1 \leq i \leq n + p + 2. \end{cases}$$

- a) Write a function using the described algorithm for insertion of valid knots. The function should compute the modified control points accordingly.
- b) Using appropriate knots and control points, show an example of internal knots appearing more than  $p$  times leading to discontinuous functions.
- c) Compute the midpoints of the unique knots

$$\eta_j = \frac{1}{2}(\xi_i + \xi_{i+1}), \quad \xi_i < \xi_{i+1}.$$

- d) Use a) to insert the knots  $\eta_j$  into the knot vector from programming exercise 10.
- e) Use the modified knots and control points from d) to plot  $\hat{N}_{i,p}$ .
- f) Use the modified knots and control points from d) to plot the corresponding B-Spline curve.

(4 Points )

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