

## B-SPLINE CURVE AS A MINIMUM OF QUADRATIC FORM AND ITS DERIVATIVES, I

SVETOSLAV I. NENOV

<sup>1</sup>Department of Mathematics, University of Chemical Technology and Metallurgy, Sofia 1756, Bulgaria  
*E-mail:* nenov@uctm.edu

**ABSTRACT.** The goal of this short note is to show a new approach for differentiation of b-spline curve.

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### 1. INTRODUCTION

The b-spline curves and surfaces are an essential tool in many engineering software for design and visualization – for example ANSYS ICEM CFD meshing software. Usually such software supports two primary geometry types, faceted and b-spline.

Faceted geometry is made up of lots of triangles (usually the data comes from 2d or 3d scanners or some low end geometry creation tools). Some types here are: STL, VRML and various mesh types, used as geometry.

Higher end CAD tools produce b-spline geometry. The sources would include low end options such as IGES/STEP and high end tools like Solidworks, etc. These are usually represented as large surfaces, each with their own UV-space.

Our goal in the article is to prove that any b-spline curve or surface minimizes positive quadratic operator: appropriate moving least-square error. As a result it is proven a formula for differentiation of b-spline curve.

We refer to Appendix A for some remarks on moving least-square method.

### 2. PRELIMINARIES: B-SPLINES

In this section we will remind the definition of b-splines generated by control points in  $\mathbb{R}^{d+1}$  and definition of moving least-squares approximation for a given data set  $\{(\mathbf{x}_i, f(\mathbf{x}_i)) : \mathbf{x}_i \in \mathbb{R}^d\} \subset \mathbb{R}^{d+1}$ .

Let  $\{\mathbf{p}_i \in \mathbb{R}^{d+1} : i = 0, \dots, n\}$  be a set of  $n + 1$  (control) points.

Let  $r$  be an integer,  $1 \leq r \leq n + 1$  (the order of spline).

We will use uniform knots, without multiplicity:  $t_i = i$ ,  $i = 0, \dots, n + r$ .

Using Cox-de Boor recursion formula (see [3], [4]), let us define the following basis functions:

$$B_{i,1}(t) = \begin{cases} 1, & \text{if } t_i \leq t < t_{i+1}, \\ 0, & \text{otherwise,} \end{cases} \quad (2.1)$$

for  $0 \leq i \leq n + r - 1$ ; and

$$\begin{aligned} B_{i,j}(t) &= \frac{t - t_i}{t_{i+j-1} - t_i} B_{i,j-1}(t) + \frac{t_{i+j} - t}{t_{i+j} - t_{i+1}} B_{i+1,j-1}(t) \\ &= \frac{t - i}{j - 1} B_{i,j-1}(t) + \frac{i + j - t}{j - 1} B_{i+1,j-1}(t), \end{aligned} \quad (2.2)$$

for  $2 \leq j \leq r$ ,  $0 \leq i \leq n + r - j$ .

The b-spline curve of order  $r$  is defined as a linear combination of control points  $\mathbf{p}_i$ :

$$\gamma(t) = \sum_{i=0}^n B_{i,r}(t) \mathbf{p}_i, \quad t \in [t_{r-1}, t_{n+1}]. \quad (2.3)$$

### 3. B-SPLINE CURVE AS A MINIMUM OF MOVING LEAST-SQUARES ERROR

Using the definitions and notations introduced in Section 2, our goal is to prove the following theorem.

**Theorem 3.1.** *Let:*

1.  $d = 1$ ;  $n, r \in \mathbb{N} = \{1, 2, \dots\}$ ,  $r \leq n + 1$ ;  $f : [0, n] \rightarrow \mathbb{R}$  be a continuous function.
2.  $\mathbf{P} = \left\{ \mathbf{p}_i = \begin{pmatrix} i \\ f(i) \end{pmatrix} : i = 0, \dots, n \right\}$  be a set of control points.
3. Let  $\gamma(t) = \begin{pmatrix} \gamma_1(t) \\ \gamma_2(t) \end{pmatrix}$  be the b-spline of order  $r$  and knot vector  $\mathbf{T} = \{i : i = 0, \dots, n + r\}$ .

*Then:*

1.  $\gamma_1(t) = t - \frac{r}{2}$ ,  $t \in [r, n + r]$
2. There exists a weight function  $W$ , such that

$$\gamma_2(x) = \hat{L}(f)(x), \quad x \in [r - 1, n + 1],$$

where  $\hat{L}(f)(x)$  is the approximation defined by the moving least-squares method for the data  $\{\mathbf{p}_i : i = 0, \dots, n + r\}$ .

*Proof.* We will prove the theorem for the cubic splines, i.e.  $r = 4$ . The proof for the different orders is similar.

$\vdots$	$\vdots$	$\vdots$	$\vdots$
$t_{i_0-4} < t_{i_0-3}$	$B_{i_0-4,1}(t) = 0$	$B_{i_0-4,2}(t) = 0$	$B_{i_0-4,3}(t) = 0$
$t_{i_0-3} < t_{i_0-2}$	$B_{i_0-3,1}(t) = 0$	$B_{i_0-3,2}(t) = 0$	$B_{i_0-4,4}(t) = 0$
$t_{i_0-2} < t_{i_0-1}$	$B_{i_0-2,1}(t) = 0$	$B_{i_0-2,2}(t) = 0$	$B_{i_0-3,4}(t) > 0$
$t_{i_0-1} < t_{i_0}$	$B_{i_0-1,1}(t) = 0$	$B_{i_0-1,2}(t) > 0$	$B_{i_0-2,4}(t) > 0$
$t_{i_0} < t < t_{i_0+1}$	$B_{i_0,1}(t) = 1$	$B_{i_0,2}(t) > 0$	$B_{i_0-1,4}(t) > 0$
$t_{i_0+1} < t_{i_0+2}$	$B_{i_0+1,1}(t) = 0$	$B_{i_0+1,2}(t) = 0$	$B_{i_0,3}(t) > 0$
$t_{i_0+2} < t_{i_0+3}$	$B_{i_0+2,1}(t) = 0$	$B_{i_0+2,2}(t) = 0$	$B_{i_0,4}(t) > 0$
$t_{i_0+3} < t_{i_0+4}$	$B_{i_0+3,1}(t) = 0$	$B_{i_0+3,2}(t) = 0$	$B_{i_0+1,4}(t) = 0$
$t_{i_0+4} < t_{i_0+5}$	$B_{i_0+4,1}(t) = 0$	$B_{i_0+2,3}(t) = 0$	$B_{i_0+2,4}(t) = 0$
$\vdots$	$\vdots$	$\vdots$	$\vdots$

TABLE 1. Cox-de Boor recursion algorithm

The b-spline curve of degree 3 (i.e. order  $r = 4$ ), defined using knots  $\mathbf{T}$  and control points  $\mathbf{P}$ , is

$$\begin{aligned} \gamma(t) &= \sum_{i=0}^n B_{i,4}(t) \mathbf{p}_i \\ &= \begin{pmatrix} \sum_{i=0}^n B_{i,4}(t) i \\ \sum_{i=0}^n B_{i,4}(t) f(i) \end{pmatrix}, \quad t \in [t_{r-1}, t_{n+1}] \equiv [3, n+1], \end{aligned}$$

Let us set

$$\gamma_1(t) = \sum_{i=0}^n B_{i,4}(t) i, \quad t \in [3, n+1]$$

and

$$\gamma_2(t) = \sum_{i=0}^n B_{i,4}(t) f(x_i), \quad x \in [3, n+1].$$

The following properties of b-spline basis functions  $B_{i,j}(t)$  are well known:

(BS-0) By the direct calculation (see formulas (2.1), (2.2), and the schema illustrated in Table 1,  $r = 4$ ):

$$B_{i,4}(t) = \begin{cases} 0, & \text{if } t < i, \\ \frac{1}{6}(t-i)^3, & \text{if } i \leq t < i+1, \\ -\frac{2}{3}(t-i-1)^3 + \frac{1}{6}(t-i)^3, & \text{if } i+1 \leq t < i+2, \\ (t-i-2)^3 - \frac{2}{3}(t-i-1)^3 \\ + \frac{1}{6}(t-i)^3, & \text{if } i+2 \leq t < i+3, \\ -\frac{2}{3}(t-i-3)^3 + (t-i-2)^3 \\ - \frac{2}{3}(t-i-1)^3 + \frac{1}{6}(t-i)^3, & \text{if } i+3 \leq t < i+4, \\ 0, & \text{if } i+4 \leq t. \end{cases}$$

(BS-1) If  $t \in (i, i+j)$ , then  $B_{i,j}(t) > 0$ .

(BS-2) If  $t \in [0, i] \cup [i+j, n+j]$ , then  $B_{i,j}(t) = 0$ .

(BS-3)  $\sum_{i=0}^n B_{i,j}(t) = 1$ , for any  $t \in (j-1, n+1)$ .

(BS-4)  $B_{i,j}(t)$  has  $C^{j-2}$  continuity at each knot.

(BS-5) By the simple substitutions in the formulas in (BS-0):

$$\begin{aligned} B_{i,j}(t+i) &= B_{k,j}(t+k), & t \in (0, j), \\ B_{i-2,j}(t) &= B_{i,j}(t+2), & t \in (i-2, i+j-2). \end{aligned}$$

Let us prove the first statement of Theorem 3.1.

From conditions (2) and (3), if we set  $x_i = i$ ,  $i = 0, \dots, n$ ,  $t_i = i$ ,  $i = 0, \dots, n+4$ , then  $x_{i-2} = i-2 = \gamma_1(t_i)$ ,  $i = 2, \dots, n+2$ .

Indeed, using the definition of b-spline function, let  $t$  be a fixed point in  $[3, n+1]$  and let  $i_0$  be an integer such that

$$3 \leq i_0 \leq t < i_0 + 1 < n + 1.$$

Then  $B_{i,4}(t) = 0$ , if  $i \leq i_0 - 4$  or  $i \geq i_0 + 1$ . Therefore

$$\begin{aligned} \gamma_1(t) &= \sum_{i=0}^n B_{i,4}(t)i \\ &= \sum_{i=i_0-3}^{i_0} B_{i,4}(t)i \end{aligned}$$

$$\begin{aligned}
 &= (i_0 - 3) \left\{ \begin{array}{ll} 0, & \text{if } t < i_0 - 3, \\ \frac{1}{6} (t + 3 - i_0)^3, & \text{if } t < -2 + i_0, \\ -\frac{2}{3} (t + 2 - i_0)^3 + \frac{1}{6} (t + 3 - i_0)^3, & \text{if } t < -1 + i_0, \\ (t + 1 - i_0)^3 - \frac{2}{3} (t + 2 - i_0)^3 + \frac{1}{6} (t + 3 - i_0)^3, & \text{if } t < i_0, \\ -\frac{2}{3} (t - i_0)^3 + (t + 1 - i_0)^3 & \\ -\frac{2}{3} (t + 2 - i_0)^3 + \frac{1}{6} (t + 3 - i_0)^3, & \text{if } t < 1 + i_0, \\ 0, & \text{if } 1 + i_0 \leq t, \end{array} \right. \\
 &+ (i_0 - 2) \left\{ \begin{array}{ll} 0, & \text{if } t < -2 + i_0, \\ \frac{1}{6} (t + 2 - i_0)^3, & \text{if } t < -1 + i_0, \\ -\frac{2}{3} (t + 1 - i_0)^3 + \frac{1}{6} (t + 2 - i_0)^3, & \text{if } t < i_0, \\ (t - i_0)^3 - \frac{2}{3} (t + 1 - i_0)^3 & \\ + \frac{1}{6} (t + 2 - i_0)^3, & \text{if } t < 1 + i_0, \\ -\frac{2}{3} (t - 1 - i_0)^3 + (t - i_0)^3 & \\ -\frac{2}{3} (t + 1 - i_0)^3 + \frac{1}{6} (t + 2 - i_0)^3, & \text{if } t < 2 + i_0, \\ 0, & \text{if } 2 + i_0 \leq t, \end{array} \right. \\
 &+ (i_0 - 1) \left\{ \begin{array}{ll} 0, & \text{if } t < -1 + i_0, \\ \frac{1}{6} (t + 1 - i_0)^3, & \text{if } t < i_0, \\ -\frac{2}{3} (t - i_0)^3 + \frac{1}{6} (t + 1 - i_0)^3, & \text{if } t < 1 + i_0, \\ (t - 1 - i_0)^3 - \frac{2}{3} (t - i_0)^3 & \\ + \frac{1}{6} (t + 1 - i_0)^3, & \text{if } t < 2 + i_0, \\ -\frac{2}{3} (t - 2 - i_0)^3 + (t - 1 - i_0)^3 & \\ -\frac{2}{3} (t - i_0)^3 + \frac{1}{6} (t + 1 - i_0)^3, & \text{if } t < 3 + i_0, \\ 0, & \text{if } 3 + i_0 \leq t, \end{array} \right. \\
 &+ i_0 \left\{ \begin{array}{ll} 0, & \text{if } t < i_0, \\ \frac{1}{6} (t - i_0)^3, & \text{if } t < 1 + i_0, \\ -\frac{2}{3} (t - 1 - i_0)^3 + \frac{1}{6} (t - i_0)^3, & \text{if } t < 2 + i_0, \\ (t - 2 - i_0)^3 - \frac{2}{3} (t - 1 - i_0)^3 & \\ + \frac{1}{6} (t - i_0)^3, & \text{if } t < 3 + i_0, \\ -\frac{2}{3} (t - 3 - i_0)^3 + (t - 2 - i_0)^3 & \\ -\frac{2}{3} (t - 1 - i_0)^3 + \frac{1}{6} (t - i_0)^3, & \text{if } t < 4 + i_0, \\ 0, & \text{if } 4 + i_0 \leq t, \end{array} \right.
 \end{aligned}$$

$$\begin{aligned}
&= (i_0 - 3) \left( -\frac{2}{3} (t - i_0)^3 + (t + 1 - i_0)^3 - \frac{2}{3} (t + 2 - i_0)^3 + \frac{1}{6} (t + 3 - i_0)^3 \right) \\
&\quad + (i_0 - 2) \left( (t - i_0)^3 - \frac{2}{3} (t + 1 - i_0)^3 + \frac{1}{6} (t + 2 - i_0)^3 \right) \\
&\quad + (i_0 - 1) \left( -\frac{2}{3} (t - i_0)^3 + \frac{1}{6} (t + 1 - i_0)^3 \right) \\
&\quad + i_0 \frac{1}{6} (t - i_0)^3 \\
&= t - 2.
\end{aligned}$$

The statement (1) of Theorem 3.1 has been proven.

To prove the second statement of Theorem 3.1, let again  $t$  be a fixed point in the interval  $(3, n + 1)$  and  $i_0$  be an integer such that  $3 \leq i_0 < t < i_0 + 1 \leq n + 1$ . Then

$$\gamma_2(t) = \sum_{i=0}^n B_{i,4}(t) f(i) = \sum_{i=i_0-3}^{i_0} B_{i,4}(t) f(i). \quad (3.1)$$

because if  $i = 1, \dots, i_0 - 4, i_0 + 1, \dots, n$ , then  $B_{i,4}(x) = 0$ , see (BS-1) and (BS-2).

On the other hand, let us consider the moving least-squares problem for the given data  $\{(i, f(i)) : i = 0, \dots, n\}$ . Let us set  $l = 1$ , and

$$W(x) = \begin{cases} 0, & \text{if } x < -2, \\ \frac{1}{6}(x+2)^3, & \text{if } -2 \leq x < -1, \\ -\frac{1}{2}x^3 - x^2 + \frac{2}{3}, & \text{if } -1 \leq x < 0, \\ \frac{1}{2}x^3 - x^2 + \frac{2}{3}, & \text{if } 0 \leq x < 1, \\ -\frac{1}{6}(x-2)^3, & \text{if } 1 \leq x < 2, \\ 0, & \text{if } 2 \leq x, \end{cases}$$

see Figure 1.

Then for any  $i = 0, \dots, n$ , we have (see also (BS-5)):

$$W(|x|) = W(x) = B_{i,4}(x + i + 2) = B_{i-2,4}(x + i), \quad x \in [-2, 2].$$

Hence  $W(|x - i|) = B_{i-2,4}(x)$ ,  $x \in [i - 2, i + 2]$ .

The least-squares error is (see also conditions (H1):  $p_1(x) = 1$ , so  $p(x) = p$  has to be a constant)

$$\sum_{i=1}^{n+r} W(x - i) (p - f(i))^2 = \sum_{i=i_0-1}^{i_0+2} W(x - i) (p - f(i))^2,$$

because  $W(x - i) > 0$ , iff  $i_0 - 1 \leq i \leq i_0 + 2$ .

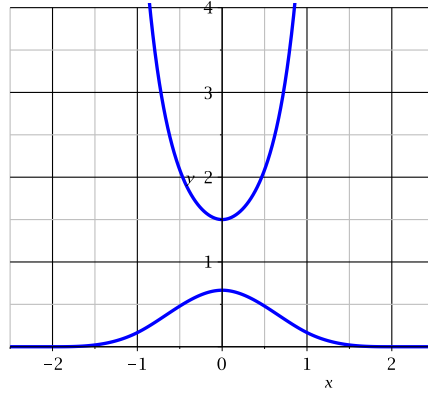


FIGURE 1. The graphics of  $W(x)$ ,  $x \in (-2.5, 2.5)$  and  $w(x)$ ,  $x \in (-1, 1)$

It is not hard to compute (see Theorem 3.2 from Appendix A)

$$E = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}, \quad \mathbf{c} = (1),$$

$$D = 2 \begin{pmatrix} w(|x - (i_0 - 1)|) & 0 & 0 & 0 \\ 0 & w(|x - i_0|) & 0 & 0 \\ 0 & 0 & w(|x - (i_0 + 1)|) & 0 \\ 0 & 0 & 0 & w(|x - (i_0 + 2)|) \end{pmatrix},$$

where  $w(x) = \frac{1}{W(x)}$ , and

$$\begin{aligned} E^t D^{-1} E &= \frac{1}{2} \left( \frac{1}{w(|x - (i_0 - 1)|)} + \frac{1}{w(|x - i_0|)} + \frac{1}{w(|x - (i_0 + 1)|)} \right. \\ &\quad \left. + \frac{1}{w(|x - (i_0 + 2)|)} \right) \\ &= \frac{1}{2} (B_{i_0-3,4}(x) + B_{i_0-2,4}(x) + B_{i_0-1,4}(x) + B_{i_0,4}(x)) \\ &= \frac{1}{2}, \quad \text{because } x \in [i_0, i_0 + 1], \text{ see (BS-3),} \end{aligned}$$

$$\mathbf{a} = D^{-1} E (E^t D^{-1} E)^{-1} \mathbf{c} = \begin{pmatrix} B_{i_0-3,4}(x) \\ B_{i_0-2,4}(x) \\ B_{i_0-1,4}(x) \\ B_{i_0,4}(x) \end{pmatrix}.$$

Hence, by Theorem 3.2 from Appendix A, we have

$$\hat{L}(f)(x) = \sum_{i=1}^4 a_i f(x_i) = \sum_{i=i_0-3}^{i_0} B_{i,4}(x) f(i),$$

i.e. we received b-spline (3.1).  $\square$

Our goal below is to calculate the first derivative of  $\gamma_2$ . Again, we will consider cubic b-splines.

Let  $t_0$  be a fixed point in the interval  $(0, n + 4)$  and  $i_0$  be an integer such that  $3 \leq i_0 < t < i_0 + 1 \leq n + 1$ ;  $x_0 = t_0 - 2 \in (3, n + 1)$ . Then

$$\hat{L}(f)(x) = \sum_{i=1}^4 a_i f(x_i),$$

where:

$$\mathbf{a} = \begin{pmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{pmatrix}, \quad E = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}, \quad \mathbf{c} = \begin{pmatrix} 1 \end{pmatrix},$$

$$D = 2 \begin{pmatrix} w(|x - (i_0 - 1)|) & 0 & 0 & 0 \\ 0 & w(|x - i_0|) & 0 & 0 \\ 0 & 0 & w(|x - (i_0 + 1)|) & 0 \\ 0 & 0 & 0 & w(|x - (i_0 + 2)|) \end{pmatrix},$$

$$D^{-1} = \frac{1}{2} \begin{pmatrix} W(x - (i_0 - 1)) & 0 & 0 & 0 \\ 0 & W(x - i_0) & 0 & 0 \\ 0 & 0 & W(x - (i_0 + 1)) & 0 \\ 0 & 0 & 0 & W(x - (i_0 + 2)) \end{pmatrix},$$

$w(x) = \frac{1}{W(x)}$ , and (see Theorem 3.2)

$$\mathbf{a} = B_0 \mathbf{c} \quad \text{and} \quad B_0 = D^{-1} E (E^t D^{-1} E)^{-1}.$$

Let us set:

$$A_0 = D^{-1} E (E^t D^{-1} E)^{-1} E^t, \quad A_1 = A_0 - I,$$

and let  $H$  be a diagonal matrix, such that  $\frac{dD^{-1}}{dx} = -HD^{-1}$ , see Remark 3.3 from the appendix. In fact, we may set

$$H = -\frac{dD^{-1}}{dx} D.$$

Using Corollary 3.5 from the appendix, we have

$$\frac{d\mathbf{a}(x)}{dx} = A_1 H \mathbf{a}(x) = -A_1 \frac{dD^{-1}}{dx} D \mathbf{a}(x). \quad (3.2)$$

But

$$A_1 = D^{-1} E (E^t D^{-1} E)^{-1} E^t - I$$



$$\begin{aligned}
&= D^{-1}E \left(\frac{1}{2}\right)^{-1} E^t - I \\
&= 2D^{-1}EE^t - I.
\end{aligned}$$

Therefore

$$\frac{d\mathbf{a}(x)}{dx} = - (2D^{-1}EE^t - I) \frac{dD^{-1}}{dx} D\mathbf{a}(x). \quad (3.3)$$

## APPENDIX A: MOVING LEAST-SQUARES METHOD

Let:

1.  $\mathcal{D}$  be a bounded domain in  $\mathbb{R}^d$ .
2.  $\mathbf{x}_i \in \mathcal{D}$ ,  $i = 0, \dots, m$ ;  $\mathbf{x}_i \neq \mathbf{x}_j$ , if  $i \neq j$ .
3.  $f : \mathcal{D} \rightarrow \mathbb{R}$  be a continuous function.
4.  $p_i : \mathcal{D} \rightarrow \mathbb{R}$  be continuous functions,  $i = 1, \dots, l$ . The functions  $\{p_1, \dots, p_l\}$  are linearly independent in  $\mathcal{D}$  and let  $\mathcal{P}_l$  be their linear span.
5.  $W : (0, \infty) \rightarrow (0, \infty)$  is a strictly positive functions.

Usually the basis in  $\mathcal{P}_l$  is constructed by monomials. For example:  $p_l(\mathbf{x}) = x_1^{k_1} \dots x_d^{k_d}$ , where  $\mathbf{x} = (x_1, \dots, x_d)$ ,  $k_1, \dots, k_d \in \mathbb{N}$ ,  $k_1 + \dots + k_d \leq l - 1$ . In the case  $d = 1$ , the standard basis is  $\{1, x, \dots, x^{l-1}\}$ .

Following [5], [6], [7], [8], we use the following definition: The *moving least-squares approximation* of order  $l$  at a fixed point  $\mathbf{x}$  is the value of  $p^*(\mathbf{x})$ , where  $p^* \in \mathcal{P}_l$  is minimizing the least-squares error

$$\sum_{i=1}^m W(\|\mathbf{x} - \mathbf{x}_i\|) (p(\mathbf{x}) - f(\mathbf{x}_i))^2 \quad (3.4)$$

among all  $p \in \mathcal{P}_l$ .

The approximation is “local” if weight function  $W(s)$  is fast decreasing as  $s$  tends to infinity. Interpolation is achieved if  $W(0) = \infty$ . We define additional function  $w : [0, \infty) \rightarrow [0, \infty)$ , such that:

$$w(s) = \begin{cases} \frac{1}{W(s)}, & \text{if } (s > 0) \text{ or } (s = 0 \text{ and } W(0) < \infty), \\ 0, & \text{if } (s = 0 \text{ and } W(0) = \infty). \end{cases}$$

Some examples of  $W(s)$  and  $w(s)$ ,  $s \geq 0$ :

$$\begin{aligned}
W(s) &= e^{-\alpha^2 s^2} && \text{exp-weight,} \\
W(s) &= s^{-\alpha^2} && \text{Shepard weights,} \\
w(s) &= s^2 e^{-\alpha^2 s^2} && \text{McLain weight,} \\
w(s) &= e^{\alpha^2 s^2} - 1 && \text{see Levin's works.}
\end{aligned}$$

Here and below: the superscript  $t$  denotes transpose of matrix;  $I$  is the identity matrix.

Let us introduce the matrices:

$$E = \begin{pmatrix} p_1(\mathbf{x}_1) & p_2(\mathbf{x}_1) & \cdots & p_l(\mathbf{x}_1) \\ p_1(\mathbf{x}_2) & p_2(\mathbf{x}_2) & \cdots & p_l(\mathbf{x}_2) \\ \vdots & \vdots & & \vdots \\ p_1(\mathbf{x}_m) & p_2(\mathbf{x}_m) & \cdots & p_l(\mathbf{x}_m) \end{pmatrix}, \quad \mathbf{a} = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_m \end{pmatrix}, \quad \mathbf{c} = \begin{pmatrix} p_1(\mathbf{x}) \\ p_2(\mathbf{x}) \\ \vdots \\ p_l(\mathbf{x}) \end{pmatrix}$$

$$D = 2 \begin{pmatrix} w(\|\mathbf{x} - \mathbf{x}_1\|) & 0 & \cdots & 0 \\ 0 & w(\|\mathbf{x} - \mathbf{x}_2\|) & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & w(\|\mathbf{x} - \mathbf{x}_m\|) \end{pmatrix}.$$

We assume the following conditions (H1):

- (H1.1)  $1 \in \mathcal{P}_l$ .
- (H1.2)  $1 \leq l \leq m$ .
- (H1.3)  $\text{rank}(E^t) = l$ .
- (H1.4)  $w$  is a smooth function.

**Theorem 3.2** (see [6]). *Let the conditions (H1) hold true.*

*Then:*

1. *The matrix  $E^t D^{-1} E$  is non-singular.*
2. *The approximation defined by the moving least-squares method is*

$$\hat{L}(f) = \sum_{i=1}^m a_i f(\mathbf{x}_i), \quad (3.5)$$

where

$$\mathbf{a} = B_0 \mathbf{c} \quad \text{and} \quad B_0 = D^{-1} E (E^t D^{-1} E)^{-1}. \quad (3.6)$$

3. *If  $w(0) = 0$ , then the approximation is interpolatory.*

Let us set

$$A_0 = D^{-1} E (E^t D^{-1} E)^{-1} E^t, \quad A_1 = A_0 - I,$$

and let us introduce the following hypothesis (H2):

(H2) There exist matrix  $H$  such that

$$\frac{dD}{dx} = HD.$$

**Remark 3.3.** *If the matrix  $D$  is non-singular, the condition (H2) is equivalent to*

$$\frac{dD^{-1}}{dx} = -HD^{-1}.$$

Indeed

$$\begin{aligned}\frac{dD^{-1}}{dx} &= -D^{-1}\frac{dD}{dx}D^{-1} \\ &= -D^{-1}HDD^{-1} \\ &= -D^{-1}H = -HD^{-1}.\end{aligned}$$

**Remark 3.4.** It is not difficult to construct the matrix  $H$  in the case of exp-weight. For example if  $m = 4$ , then

$$D = \begin{pmatrix} e^{-\alpha^2(x-x_1)^2} & 0 & 0 & 0 \\ 0 & e^{-\alpha^2(x-x_2)^2} & 0 & 0 \\ 0 & 0 & e^{-\alpha^2(x-x_3)^2} & 0 \\ 0 & 0 & 0 & e^{-\alpha^2(x-x_4)^2} \end{pmatrix}$$

and

$$\frac{dD}{dx} = HD,$$

where

$$H = -2\alpha^2 \begin{pmatrix} x - x_1 & 0 & 0 & 0 \\ 0 & x - x_2 & 0 & 0 \\ 0 & 0 & x - x_3 & 0 \\ 0 & 0 & 0 & x - x_4 \end{pmatrix}.$$

**Corollary 3.5.** Let the conditions (H1) and (H2) hold true and let  $d = 1$ .

Then

$$\frac{d\mathbf{a}(x)}{dx} = A_1 H \mathbf{a}(x) + B_0 \frac{d}{dx} \mathbf{c}(x), \quad (3.7)$$

and

$$E^t \frac{d\mathbf{a}(x)}{dx} = \frac{d}{dx} \mathbf{c}. \quad (3.8)$$

*Proof.* We have

$$\begin{aligned}\frac{d\mathbf{a}(x)}{dx} &= \frac{d}{dx} \left( D^{-1} E (E^t D^{-1} E)^{-1} \mathbf{c} \right) \\ &= \left( \frac{d}{dx} D^{-1} \right) E (E^t D^{-1} E)^{-1} \mathbf{c} + D^{-1} E \left( \frac{d}{dx} (E^t D^{-1} E)^{-1} \right) \mathbf{c} \\ &\quad + D^{-1} E (E^t D^{-1} E)^{-1} \frac{d}{dx} \mathbf{c} \\ &= -HD^{-1} E (E^t D^{-1} E)^{-1} \mathbf{c} \\ &\quad + D^{-1} E \left( - (E^t D^{-1} E)^{-1} \left( \frac{d}{d\alpha} E^t D^{-1} E \right) (E^t D^{-1} E)^{-1} \right) \mathbf{c} \\ &\quad + D^{-1} E (E^t D^{-1} E)^{-1} \frac{d}{dx} \mathbf{c} \\ &= -H\mathbf{a}\end{aligned}$$

$$\begin{aligned}
& + D^{-1}E (E^t D^{-1}E)^{-1} (E^t H D^{-1}E) (E^t D^{-1}E)^{-1} \mathbf{c} \\
& + D^{-1}E (E^t D^{-1}E)^{-1} \frac{d}{dx} \mathbf{c} \\
= & D^{-1}E (E^t D^{-1}E)^{-1} (E^t H) \left( D^{-1}E (E^t D^{-1}E)^{-1} \right) \mathbf{c} \\
& - H \mathbf{a} \\
& + D^{-1}E (E^t D^{-1}E)^{-1} \frac{d}{dx} \mathbf{c} \\
= & \left( D^{-1}E (E^t D^{-1}E)^{-1} E^t - I \right) H \mathbf{a} \\
& + D^{-1}E (E^t D^{-1}E)^{-1} \frac{d}{dx} \mathbf{c} \\
= & A_1 H \mathbf{a} + B_0 \frac{d}{dx} \mathbf{c}.
\end{aligned}$$

Multiplying the equation (3.7) by  $E^t$  on the left, yields

$$\begin{aligned}
E^t \frac{d\mathbf{a}(x)}{dx} & = E^t \left( D^{-1}E (E^t D^{-1}E)^{-1} E^t - I \right) H \mathbf{a} \\
& + E^t D^{-1}E (E^t D^{-1}E)^{-1} \frac{d}{dx} \mathbf{c} \\
& = \left( (E^t D^{-1}E) (E^t D^{-1}E)^{-1} E^t - E^t \right) H \mathbf{a} \\
& + (E^t D^{-1}E) (E^t D^{-1}E)^{-1} \frac{d}{dx} \mathbf{c} \\
& = \frac{d}{dx} \mathbf{c}.
\end{aligned}$$

Corollary 3.5 has been proven. □

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