

# Presenting Univariate, Linear and Stationary Subdivision Schemes for Refining Noisy Data

Sombir\*

M.Sc. in Mathematics, Kurukshetra University, Kurukshetra

**Abstract – This is the principal endeavour to plan subdivision schemes for noisy data. We present and dissect univariate, linear, and stationary subdivision schemes for refining noisy data, by fitting nearby least squares polynomials. We present primal schemes, with refinement rules dependent on locally fitting linear polynomials to the data, and concentrate their convergence, smoothness, and basic limit functions. Primal schemes and schemes identified with noisy data are first talked about, in light of fitting linear polynomials to the data, and concentrate their convergence, smoothness, and basic limit functions. In this investigation we manage the issue of-"how to surmised a function from its noisy examples by subdivision schemes".**

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## I. INTRODUCTION

As of late, subdivision schemes have turned into a significant device in numerous applications and research zones, including movement, PC illustrations, and PC supported geometric design, just to give some examples. A subdivision plan creates values related with the vertices of a succession of settled meshes, with a thick association, by rehashed use of a lot of neighbourhood refinement rules.

These rules decide the values related with a refined mesh from the values related with the coarser mesh. The subdivision plan is focalized if the produced values combine consistently to the values of a constant function, for any arrangement of introductory values. The specific class of interpolatory schemes comprises of schemes with refinement rules that keep the values related with the coarse mesh and just create new values identified with the extra vertices of the refined mesh. A significant group of interpolatory schemes is the group of Dubuc-Deslauriers (DD) schemes.

As of late, subdivision schemes have turned into a vital tool in numerous applications and research territories, including animation, computer graphics, and computer supported geometric design, just to give some examples. Subdivision schemes, and explicitly binary subdivision schemes (BSS), are methods used to create and speak to smooth curves and surfaces by an iterative procedure of refinement. Subdivision schemes are multi-resolution methods utilized in computer-supported geometric design to produce smooth curves or surfaces. This paper proposes new methods for abusing the subdivision procedure for data examination and compression.

Subdivision is a procedure of recursively refining discrete data utilizing a lot of subdivision rules (called subdivision plot) which creates a persistent or even smooth limit. It has various applications, for example, picture recreation, the design of curves and surfaces, shape protection in data and geometric items, the estimation of self-assertive functions, and so on.

The subdivision schemes are distinctive because of the diverse qualities of  $\beta$ 's. A few systems have been utilized to discover  $\beta$ 's. A few analysts utilized Lagrange polynomials to figure  $\beta$ 's while others utilized wavelet strategies for the calculations of these qualities. A few mathematicians utilized Hermite polynomials to produce the estimations of  $\beta$ 's. One can get 2-point, 3-point,... n- point schemes from the mask (i.e. the values of  $\beta$ 's) got from the above systems. A class of subdivision schemes by making the variation of existing algorithms was presented. Later a group of schemes by convolving the current schemes. A noteworthy favorable position of the subdivision schemes is that they can be effectively connected to virtually any data type. Be that as it may, by early work in the subdivision schemes don't manage noisy data with impulsive noises. It was brought up that minimum squares based subdivision schemes are better decisions to deal with these kinds of issues. In this manner, in this paper, we want to utilize slightest squares strategy rather than different procedures.

The strategy for slightest squares is one of the brilliant procedures in measurements for curve fitting. In this cutting edge period strategy for minimum squares is habitually used to discover numerical estimations of the parameters to fit a

function to set of data. It implies that the general arrangement limits the whole of the squares of the mistakes made in the aftereffects of each and every equation i.e. the best fitted curve by the minimum squares methods limits the entirety of squared distinction between a watched esteem and the fitted esteem given by a model.

Suppose that the data points are  $(r_1, f_{r_1}), (r_2, f_{r_2}), (r_3, f_{r_3}), \dots, (r_m, f_{r_m})$ , where  $r_i$  the independent variable is, and  $f_{r_i}$  is the dependent variable. The fitting curve has the deviation (error) from each data point, i.e.

$$d_1 = f_{r_1} - r_1, d_2 = f_{r_2} - r_2, d_3 = f_{r_3} - r_3, \dots, d_m = f_{r_m} - r_m$$

According to the techniques of least squares, the best fit curve has the property that  $R = \sum_{r=1}^m d_i^2$ , is minimum.

## II. DIFFERENT SUBDIVISION SCHEMES

### i. Linear Schemes

We review some results from the theory of linear schemes. We let  $n = 1$ , i.e. the points  $p_\gamma$  are elements of  $\mathbb{R}$ . This is no restriction, since convergence and smoothness properties can be treated component wise in the case of scalar linear schemes. In the univariate setting, the (unique) difference operator plays an important role. On grids we look at all the difference operators in the different directions on the grid simultaneously. This allows us to define derived schemes as discrete analogues of the left derivatives.

Let  $\Delta_i$  be the operator that maps grid function  $(p_\gamma)_{\gamma \in \mathbb{Z}^s}$ , to the grid function  $(p_\gamma - p_{\gamma - \delta_i})_{\gamma \in \mathbb{Z}^s}$ , where  $\delta_i$  denotes the  $i$ -th canonical basis vector in  $\mathbb{Z}^s$ .

Definition 4.1 For  $p \in l^\infty(\mathbb{Z}^s)$  we define  $\Delta_p \in (l^\infty(\mathbb{Z}^s))^s$  as

$$\Delta_p: (\Delta_{1p}, \Delta_{2p}, \dots, \Delta_{sp})^T.$$

For a subdivision scheme to converge, we need that the distance between consecutive points becomes small.

### ii. Stationary Schemes

Here we talk about just stationary subdivision schemes, and subsequently, we won't compose "stationary" any longer. We signify stationary scheme with letter S, the subdivision procedure ought to be ordinary.

### Some basic definition

Definition 2.1 A subdivision scheme S is **finite** if there exists an integer  $B \geq 0$ , such that for every  $\alpha \in \mathbb{Z}^s, (v^1)_\alpha$ , is a function of at most  $B$  elements of  $v^0$ .

Definition 1.1 A subdivision scheme S **reproduces constants** if  $S^1 = 1$ , where  $1 \in l^\infty(\mathbb{Z}^s)$  is the constant sequence 1.

Definition 1.2 A subdivision scheme S is **affine invariant** if  $S(av + b1) = aSv + b1$  for every  $a, b \in \mathbb{R}$  and  $v \in l^\infty(\mathbb{Z}^s)$ .

Since we are managing stationary subdivision schemes, our dynamical framework is self-governing! A subdivision scheme S which repeats constants, can't be asymptotically stable, in such a case that we take  $v^0 = 1$  and  $\tilde{v}^0 = 0$ , at that point for each  $j \geq 0$  the separation between  $\{(v^j, j)\}$  and  $\{(\tilde{v}^j, j)\}$  is 1. It looks like Liapunov soundness yet the primary downside of this elucidation is that, not normal for the Liapunov dependability, the trajectories of "close" initial points remain in a cylinder around the initial trajectory simply after some limited minute, that relies upon the initial sequence and isn't consistently limited. Once more, similarly as with the uniform convergence, if S is move invariant it suffices to take  $\Omega = [0, 1]^s$ , and if S is neighbourhood - we can work with limited sub sequences of  $v^0$  and  $\tilde{v}^0$ .

## III. PRIMAL LEAST SQUARES SCHEMES

We begin by considering the simplest least squares subdivision scheme  $S_n$  for  $n \geq 1$ , which produces the data at level  $k + 1$  as pursues. From one perspective, the esteem  $f_{2i}^{k+1}$ , which replaces  $f_i^k$ , is controlled by fitting a linear polynomial to the  $2n - 1$  data esteems in a symmetric neighborhood around  $t_i^k$  and assessing it at the related dyadic point  $t_i^k = t_{2i}^{k+1}$ . On the other, the scheme processes the new esteem  $f_{2i+1}^{k+1}$  between  $f_i^k$  and  $f_{i+1}^k$  by assessing the linear slightest squares polynomial regarding the closest  $2n$  data esteems somewhere between the comparing dyadic points  $t_i^k$  and  $t_{i+1}^k$ , to be specific at  $t_{2i+1}^{k+1}$ . In this development the parameter  $n$  controls the locality of the scheme.

In particular, we use polynomials that fit the data best in the least squares sense. That is, for given data  $y_1, \dots, y_m$  at nodes  $x_1, \dots, x_m$ , we are interested in the polynomial pd of some degree  $d$  that minimizes the sum of squared residuals,

$$\sum_{i=1}^m (p_d(x_i) - y_1)^2 \quad (1)$$

Following (1), the refinement rules of  $S_n$  turn out to be

$$f_{2i}^{k+1} = \frac{1}{2n-1} \sum_{j=-n+1}^{n-1} f_{i+j}^k \quad \text{and} \quad f_{2i+1}^{k+1} = \frac{1}{2n} \sum_{j=-n+1}^n f_{i+j}^k. \quad (2)$$

Consequently, the symbol [7] of the scheme is

$$a_n(z) = \frac{1}{2n} \sum_{j=-n+1}^n z^{2j-1} + \frac{1}{2n-1} z^{2j} \quad (3)$$

It pursues from the symmetry of the points deciding the linear least squares polynomials, that  $a_n(z) = a_n(1/z)$ , thus the scheme is odd symmetric [11]. As the data at level  $k+1$  relies upon at most  $2n$  values at level  $k$ , we infer that  $S_n$  is a primal  $2n$ -point scheme. A case of the refinement conditions for  $S_3$  is appeared in Figure 1, and the masks of the initial three schemes are

$$\begin{aligned} a_1 &= [1,2,1] & /2, \\ a_2 &= [3,4,3,4,3,4,3] & /12 \\ a_3 &= [5,6,5,6,5,6,5,6,5,6,5] & /30 \end{aligned}$$

Note that the scheme  $S_1$  is the interpolating 2-point scheme, which generates piecewise linear functions in the limit.

#### IV. CONVERGENCE AND SMOOTHNESS

Following the usual definition of convergence, we denote the limit of a convergent subdivision scheme  $S$  for initial data  $f^0$  by  $S^\infty f^0$ . The explicit form of the symbol in (3) implies that  $a_n(1) = 2$  and  $a_n(-1) = 0$ , which are necessary conditions for  $S_n$  to be convergent. Following the analysis in, we define

$$q_n(z) = \frac{a_n(z)}{1+z} = \frac{1}{2n(2n-1)} \left( \sum_{j=-n+1}^{n-1} (n-j)z^{2j-1} + \sum_{j=-n+1}^{n-1} (n+j)z^{2j} \right) \quad (4)$$

and conclude the convergence of  $S_n$  by analysing the norm of the subdivision scheme with symbol  $q_n$ .

Theorem 1 The least squares subdivision scheme  $S_n$  is convergent for  $n \geq 1$ .

Proof It follows from (4) that the norm of the subdivision scheme with symbol  $q_n$  is

$$\begin{aligned} \|S_{[q_n]}\|_\infty &= \max \left\{ \frac{1}{2n(2n-1)} \sum_{j=-n+1}^{n-1} |n-j|, \frac{1}{2n(2n-1)} \sum_{j=-n+1}^{n-1} |n+j| \right\} \\ &= \frac{1}{2n(2n-1)} \sum_{j=1}^{2n-1} j = \frac{1}{2}. \end{aligned}$$

According to the scheme  $S_n$  is therefore convergent

Note that the norm of the scheme  $S_{[q_n]}$  is the least possible value as is the case for the uniform B-spline schemes, indicating "quickest" possible convergence. The structure of  $q_n$  further reveals that the limit functions generated by  $S_n$  are  $C^1$ .

#### 1. The Basic Limit Function

Give us a chance to mean by  $\delta$  the sequence which is zero wherever aside from at 0, where it is 1. The basic limit function of the convergent subdivision scheme  $S_n$  is then characterized as

$$\phi_n = S_n^\infty \delta. \quad (5)$$

A few instances of  $\phi_n$  for small values of  $n$  are appeared in Figure 2.

Numerous properties of a linear subdivision scheme can be gotten from its basic limit function. Specifically, because of linearity, the limit function produced from the initial data  $f^0 = (f_i^0)_{i \in \mathbb{Z}}$  by the scheme  $S_n$  has the frame

$$(S_n^\infty f^0)(x) = \sum_{j \in \mathbb{Z}} f_j^0 \phi_n(x-j). \quad (6)$$

Our first observation is that the help of  $\phi_n$  is  $[-2n+1, 2n-1]$ , in light of the fact that  $S_n$  is a primal  $2n$ -point scheme [6]. Additionally,  $\phi_n$  is sure inside its help, in light of the fact that the coefficients of the mask  $a_n$  are sure in the mask's help, and  $\phi_n$  has the segment of unity property

$$\sum_{j \in \mathbb{Z}} \phi_n(x-j) = 1 \quad (7)$$

because of the multiplication of consistent polynomials by  $S_n$ .

## V. CONCLUSION

In this paper we presented application of univariate schemes for subdivision of noisy data. Different schemes which are using geometrical applications for refining data which are univariate, linear and stationary schemes. Univariate polynomials are taken for the study. Convergence, smoothness, and basic limit functions of linear polynomials are analyzed by using primal schemes.

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## Corresponding Author

### Sombir\*

M.Sc. in Mathematics, Kurukshetra University,  
Kurukshetra

[sombirlamba2@gmail.com](mailto:sombirlamba2@gmail.com)