

# Alpha-rooting and Correlation Method of Image Enhancement



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# Abstract

- This paper analyzes the method of the Fourier transforms-based alpha-rooting in image enhancement and describes a new generalized method of alpha-rooting, by using the autocorrelation function of the image.
- The alpha-rooting can be described by the Taylor series in the frequency-domain, as well as in the spatial domain. In such a series, the alpha-rooting is the convolution of the image with the series of the autocorrelation functions.
- The application of the Taylor series in alpha-rooting allows us to use the parameterized filters even for alpha parameter in a much larger interval than  $[0,1]$ .
- New correlation alpha-rooting filters are presented.
- Examples of application of these two filters for enhancement of the grayscale 'jetplane' image are given.

# Alpha-rooting and approximation

In the  $\alpha$ -rooting method of image enhancement, the magnitude of the 2-D discrete Fourier transform (DFT) of the discrete image of  $N \times M$  pixel is modified as follows:

$$|F_{p,s}| \rightarrow |F_{p,s}|^\alpha, \quad p = 0:(N-1), \quad s = 0:(M-1), \quad (1)$$

where the parameter  $\alpha$  is from the interval (0,1). The phase  $\varphi(p,s)$  of the transform does not change. The coefficients of the Fourier transform are multiplied by the coefficients  $|F_{p,s}|^{\alpha-1}$ ,

$$F_{p,s} = |F_{p,s}|e^{i\varphi(p,s)} \rightarrow Y_{p,s} = |F_{p,s}|^\alpha e^{i\varphi(p,s)} = |F_{p,s}|^{\alpha-1}|F_{p,s}|e^{i\varphi(p,s)} = |F_{p,s}|^{\alpha-1}F_{p,s}. \quad (2)$$

Here,  $\varphi(p,s)$  is the phase of the transform at frequency-point  $(p,s)$ .

The enhanced image is the inverse 2-D DFT of the modified transform

$$y_{n,m} = \frac{1}{NM} \sum_{p=0}^{N-1} \sum_{s=0}^{M-1} F_{p,s} |F_{p,s}|^{\alpha-1} W_N^{-np} W_M^{-ms}, \quad n = 0:(N-1), m = 0:(M-1). \quad (3)$$

where the exponential coefficients are  $W_N^t = \exp\left(-\frac{i2\pi t}{N}\right)$  and  $W_M^t = \exp\left(-\frac{i2\pi t}{M}\right)$ .

For simplicity of calculation, we consider the 1-D case of alpha-rooting of a signal  $f_n$ , whose  $N$ -point discrete Fourier transform  $F_p$  is modified as

$$F_p \rightarrow F_p |F_p|^{\alpha-1}, \quad p = 0:(N-1). \quad (4)$$

The inverse discrete Fourier transform of  $|F_p|^2$  is equal to the autocorrelation function

$$K(n) = \sum_{k=0}^{N-1} f_{n+k} f_k, \quad n = 0:(N-1). \quad (5)$$

The signal is considered periodic and the subscripts in this sum are calculated by modulo  $N$ .

For this autocorrelation function, we will use the notation  $K = f \circ f$ . Note that  $K(-n) = K(n)$ .

If we consider the inverse DFT of the spectrum  $|F_p|^4$ , we obtain the second order correlation, i.e.,  $K \circ K = f \circ f \circ f \circ f$ , or simply  $K^{(2)} = f^{(4)}$ .

The Taylor series

$$(1 + x)^a = 1 + \sum_{n=1}^{\infty} \binom{a}{n} x^n, \quad |x| < 1,$$

can be written as

$$x^a = \sum_{n=0}^{\infty} \binom{a}{n} (x - 1)^n = 1 + \sum_{n=1}^{\infty} \binom{a}{n} (x - 1)^n, \quad 0 < x < 2. \quad (6)$$

Here, the binomial coefficients are

$$\binom{a}{n} = \frac{a(a - 1) \cdots (a - n + 1)}{n!}.$$

We consider the spectrum  $|F_p|^{\alpha-1}$  in the form

$$|F_p|^{\alpha-1} = \left(|F_p|^2\right)^{\frac{\alpha-1}{2}}, \quad p = 0, 1, \dots, (N - 1). \quad (7)$$

To use the Taylor series for this power, we perform the normalization by the constant  $A_0$ , i.e.,

$$F_p = F_p/A, \quad p = 1, 2, \dots, (N - 1), \quad (8)$$

such that  $|F_p| < 1$ . For instance, the normalization coefficient  $A = N \times \max|F_p|$  can be used.

Denoting  $b = (a - 1)/2$  and  $G_p = |F_p|^2$ , we write the following Taylor series:

$$|F_p|^{\alpha-1} = \sum_{n=0}^{\infty} \binom{b}{n} (G_p - 1)^n, \quad 0 < G_p < 1. \quad (9)$$

Here, the cases when  $G_p = 0$  and 1 are not considered, since the alpha-rooting does not change the values of such components. The inverse DFT of this series can be written as

$$H(x) = \frac{1}{N} \sum_{p=0}^{N-1} |F_p|^{\alpha-1} W_N^{-px} = \sum_{n=0}^{\infty} \binom{b}{n} (K - \delta)^n(x), \quad x = 0, 1, \dots, (N - 1). \quad (10)$$

In this sum,  $\delta(x)$  is the discrete unit sample function, and the polynomials are

$$(K - \delta)^n(x) = \sum_{m=0}^n \binom{n}{m} K^{(m)}(x) * (-\delta(x))^{n-m}, \quad (11)$$

and  $K^{(m)}(x) = \underbrace{(K \circ K \circ \dots \circ K)}_{m \text{ times}}(x)$  and  $\delta^{(k)}(x) = \underbrace{(\delta \circ \delta \circ \dots \circ \delta)}_{k \text{ times}}(x) = \delta(x)$ , when integer  $k > 0$ .

Thus, the function  $H(x)$ , that represents the filter of the alpha-rooting in time domain, can be written in Eq. 10 as

$$\begin{aligned} H(x) = & \delta(x) + b(K(x) - \delta(x)) + \frac{b(b-1)}{2} (K^{(2)}(x) - 2K(x) + \delta(x)) + \\ & + \frac{b(b-1)(b-2)}{3!} (K^{(3)}(x) - 3K^{(2)}(x) + 3K(x) + \delta(x)) + \dots . \end{aligned} \quad (12)$$



In the time-domain, the alpha-rooting  $F_p \rightarrow F_p |F_p|^{\alpha-1}$  means the convolution of the signal  $f_n$  with the obtained filter  $H(x)$ . This filter in the zeroth approximation is equal to

$$H_0(x) = \delta(x),$$

and in the first approximation, it is equal to

$$H_1(x) = \delta(x) + b(K(x) - \delta(x)) = (1 - b)\delta(x) + bK(x), \quad (13)$$

and in the 2<sup>nd</sup> approximation, it is equal to

$$\begin{aligned} H_2(x) &= \delta(x) + b(K(x) - \delta(x)) + \frac{b(b-1)}{2} (K^{(2)}(x) - 2K(x) + \delta(x)) \\ &= \frac{(1-b)(2-b)}{2} \delta(x) + b(2-b)K(x) + \frac{b(b-1)}{2} K^{(2)}(x). \end{aligned} \quad (14)$$

The results of the alpha-rooting of the signal  $f_n$  when using these approximations are

$$(y_0)_n = (f * H_0)_n = f_n,$$

$$(y_1)_n = (f * H_1)_n = (1 - b)f_n + bf_n * K(n), \quad (15)$$

$$(y_2)_n = (f * H_2)_n = \frac{(1 - b)(2 - b)}{2}f_n + b(2 - b)f_n * K(n) + \frac{b(b - 1)}{2}f_n * K^{(2)}(n). \quad (16)$$

In Eq. 15, the coefficient  $(1 - b) > 0$  when  $0 < a < 3$ . Using the coefficients

$$c_1 = \frac{(1 - b)(2 - b)}{2}, \quad c_2 = b(2 - b), \quad \text{and} \quad c_3 = \frac{b(b - 1)}{2},$$

The signal in the 2<sup>nd</sup> approximation can be written as

$$(y_2)_n = (f * H_2)_n = c_1f_n + c_2f_n * K(n) + c_3f_n * K^{(2)}(n).$$

In the first approximation, the convolution is the mean of the signal and its convolution with autocorrelation function.

It also can be written as the signal plus a gradient,

$$(y_1)_n = f_n - b[f_n - f_n * K(n)], \quad (-b > 0).$$

Thus, the correlation function is used to calculate a gradient of the signal (image).

It is interesting to note that the sum of the coefficients in Eq. 16 is equals to 1, too. Indeed,

$$c_1 + c_2 + c_3 = \frac{(1-b)(2-b)}{2} + b(2-b) + \frac{b(b-1)}{2} = \frac{2-3b+b^2+4b-2b^2+b^2-b}{2} = 1.$$

The same is for the higher than the 2<sup>nd</sup> order approximation of the alpha-rooting filter  $H(x)$  in Eq. 12.

### ***Example 1:***

Consider the  $\alpha = \frac{3}{4}$  case, for which  $b = \frac{\alpha-1}{2} = -\frac{1}{8}$ .

Then, the 1<sup>st</sup> and 2<sup>nd</sup> approximations of the alpha-rooting are the signals

$$(y_1)_n = (f * H_1)_n = \frac{9}{8}f_n - \frac{1}{8}f_n * K(n) = (y_1)_n = f_n + \frac{1}{8}[f_n - f_n * K(n)], \quad (18)$$

and

$$\begin{aligned} (y_2)_n &= (f * H_2)_n = \frac{1}{64}[76.5f_n - 17f_n * K(n) + 4.5f_n * K^{(2)}(n)]. \\ &= 1.1953f_n - 0.2656f_n * K(n) + 0.0703f_n * K^{(2)}(n). \end{aligned} \quad (19)$$

For the 0.9-rooting approximations, we obtain the signals

$$(y_1)_n = (f * H_1)_n = \frac{21}{20}f_n - \frac{1}{20}f_n * K(n) = (y_1)_n = f_n + \frac{1}{20}[f_n - f_n * K(n)], \quad (20)$$

and

$$(y_2)_n = (f * H_2)_n = 1.0762f_n - 0.1025f_n * K(n) + 0.0262f_n * K^{(2)}(n). \quad (21)$$

One can see from these two examples that the coefficients  $c_1, c_2$ , and  $c_3$  for the 0.9-rooting became smaller, when comparing with the  $\alpha = 0.75$  case.

The graphs of these coefficients as functions of  $\alpha$ , i.e.,  $c_1(\alpha)$ ,  $c_2(\alpha)$ , and  $c_3(\alpha)$  are shown in Fig. 1, when  $\alpha$  runs the interval  $[0.2, 1]$ . A significant contribution of the correlation function in alpha-rooting is when the parameter  $\alpha$  takes small values. When  $\alpha$  is close to 1, the coefficients  $c_2(\alpha)$  and  $c_3(\alpha)$  are very small.

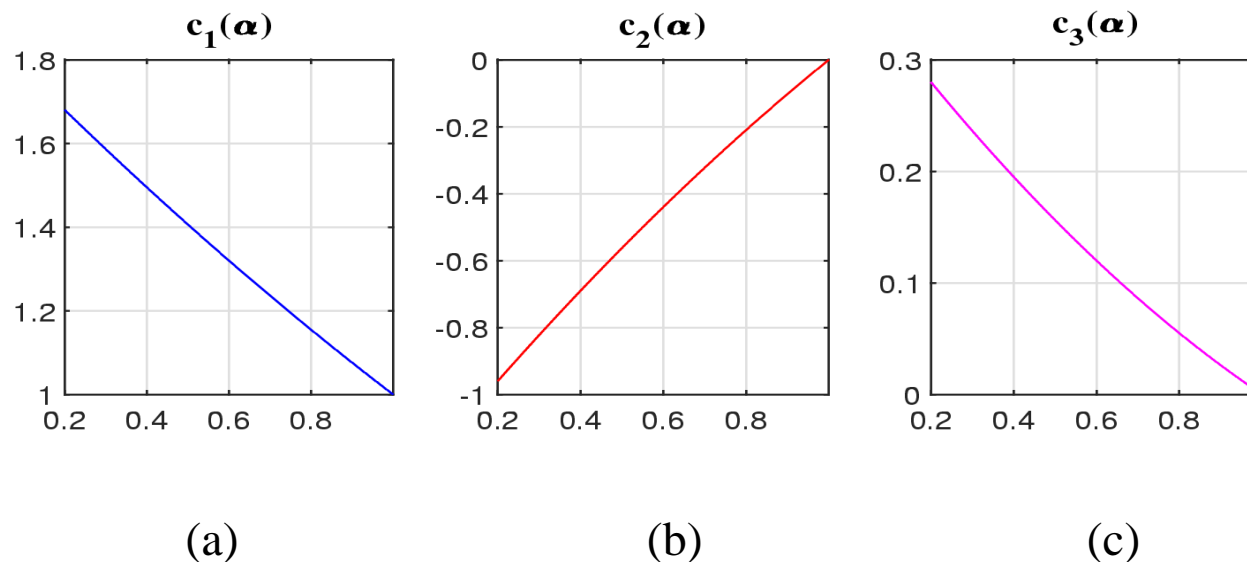


Figure 1. (a) The coefficients of the 2<sup>nd</sup> approximation of the alpha-rooting.

The filter  $H(x)$  in Eq. 12 was derived for the normalized signal  $\check{f}_n = f_n/A$  and not for  $f_n$ . For instance, in the first approximation

$$(y_1)_n = \frac{f_n}{A} * H_1(n), \quad \text{or} \quad (y_1)_n A = f_n * H_1(n).$$

Here, the filter  $H_1(n) = (1 - b)\delta(n) + bK(n)$  is defined by the autocorrelation function  $K(n)$  of  $\check{f}_n$ , i.e., up to the constant  $1/A^2$ , it is the autocorrelation function  $R(n)$  of the original signal. Thus,

$$A(y_1)_n = (1 - b)f_n + b \frac{1}{A^2} f_n * R(n). \quad (22)$$

Similarly, the 2<sup>nd</sup> approximation of the alpha-rooting can be written as

$$A(y_2)_n = c_1 f_n + c_2 \frac{1}{A^2} f_n * R(n) + c_3 \frac{1}{A^4} f_n * R^{(2)}(n). \quad (23)$$

The output of the alpha-rooting will be scaled to the original range, let say [1,255]. Therefore, the constant A in the left parts of Eqs. 22 and 23, can be omitted. The alpha-rooting is not linear transformation.

1. The first modification of the alpha-rooting of the signal is defined as

$$(y_1)_n = (1 - b)f_n + (bA_1)f_n * K(n). \quad (24)$$

2. The 2<sup>nd</sup> modification of the alpha-rooting of the original signal is defined as

$$(y_2)_n = c_1f_n + (c_2A_2)f_n * K(n) + (c_3B_2)f_n * K^{(2)}(n). \quad (25)$$

We will call the corresponding filters

$$H_1(n) = (1 - b)\delta(n) + (bA_1)K(n) \quad (26)$$

and

$$H_2(n) = c_1\delta(n) + (c_2A_2)K(n) + (c_3B_2)K^{(2)}(n) \quad (27)$$

the 1<sup>st</sup> and 2<sup>nd</sup> order correlation alpha-rooting filters, respectively.

The coefficients  $c_2$ , and  $c_3$  at the autocorrelation function in the above approximations are small. Therefore, in these new filters, the parameters  $A_1$ ,  $A_2$ , and  $B_2$  will be chosen with large values.

Figure 2 show the “jetplane” image of  $256 \times 256$  pixels in part (a), the first and second approximation of the 0.75-rooting in parts (b) and (c), respectively. The parameters of the filters are  $A_1 = 10$ , and  $A_2 = 5, B_2 = 10$ . In part d, the result of the traditional 0.75-rooting. The last image is shown with twice large amplitude (i.e.,  $2y_2$ ); otherwise, the displayed image will be very dark.

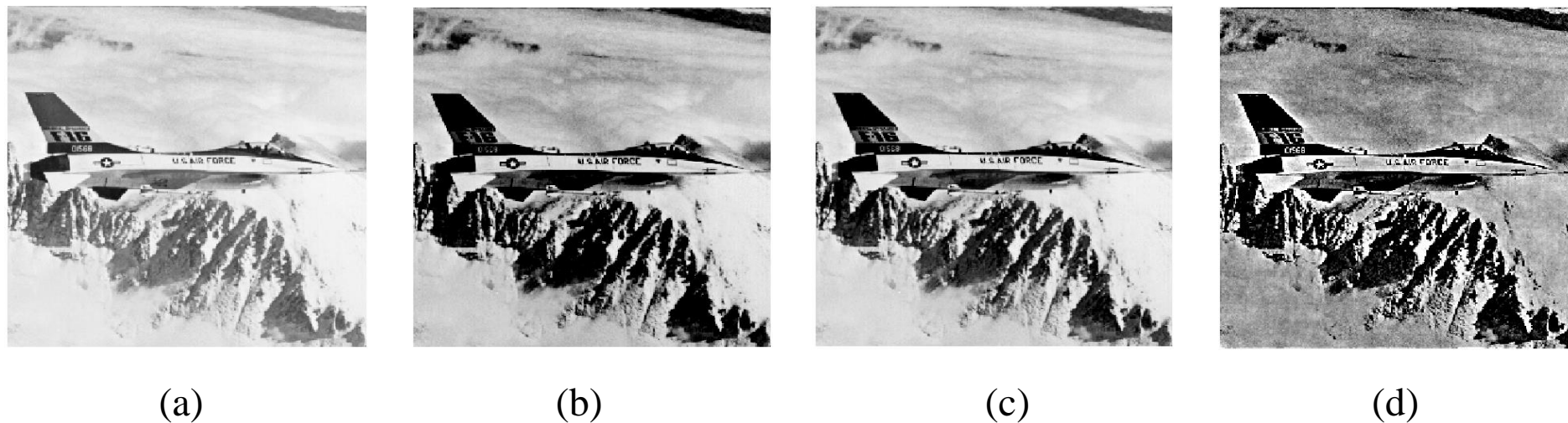


Figure 2. (a) The original image, (b) the 1<sup>st</sup> and (c) 2<sup>nd</sup> approximations, and (d) the alpha-rooting with  $\alpha = 0.75$ .



Figure 3 shows the same image in part (a) and the first and second approximations of the 0.90-rooting in parts (b) and (c), respectively. The parameters of the filters are  $A_1 = 10$  and  $A_2 = 10$ ,  $B_2 = 20$ .



(a)



(b)



(c)

Figure 3. (a) The “jetplane” image, and (b) the 1<sup>st</sup> and (c) 2<sup>nd</sup> approximations of the alpha-rooting with  $\alpha = 0.90$ .

The traditional alpha-rooting works for the parameter alpha from the interval (0,1). The above Eqs. 12-27 can be used for the values of alpha parameter outside this interval, too. Figure 4 shows the 1<sup>st</sup> and 2<sup>nd</sup> approximations, when  $\alpha = 1.5$ . The parameters of the filters are  $A_1 = -2$  and  $A_2 = 1$ ,  $B_2 = 20$ .



(a)



(b)

Figure 4. (a) The 1<sup>st</sup> and (b) 2<sup>nd</sup> approximations of the image when  $\alpha = 1.5$ .

Figure 5 shows the 1<sup>st</sup> approximation of the alpha-rooting, when  $\alpha = -0.75$ . The parameters of the filter  $A_1 = -2$ .



Figure 5. The 1<sup>st</sup> approximations of the image when  $\alpha = -0.75$ .

The best parameter of alpha for the proposed method can be found, by using a measure of enhancement. For instance, the EME enhancement measure can be used, as in the traditional method of alpha-rooting

## SUMMARY

This paper analyzes the traditional method of the Fourier transforms-based alpha-rooting in image enhancement and presents the approximations of the alpha-rooting, by using the autocorrelation function of the image. In the spatial domain, the alpha-rooting can be described by the Taylor series. For this, the inverse 2-D DFT is used. In such a series, the alpha-rooting is the convolution of the image with the series of the autocorrelation functions. The application of the Taylor series in the method of alpha-rooting allows to use the new correlation alpha-rooting filters even for values of alpha in an interval larger than  $[0,1]$ . Examples of application of correlation alpha-rooting filters for enhancement of the grayscale image 'jetplane' are given. We believe that the proposed methods can also be used to enhance color images.

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