

# Sliding signal processing in telecommunication networks based on two-dimensional discrete Fourier transform

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**Abstract.** A method of vertical sliding processing of two-dimensional discrete signals in the spatial frequency domain is proposed — a method of fast vertically sliding two-dimensional discrete Fourier transform. The mathematical representation of the two-dimensional discrete Fourier transform in algebraic and matrix form is considered. An effective method of vertically sliding two-dimensional discrete Fourier transform is proposed. The algorithm developed in the framework of the proposed method allows calculating the coefficients (bins) of this transformation in real time.

## 1 Introduction

It is difficult to overestimate the role and place of digital spectral processing of discrete one-dimensional (1-D) and two-dimensional (2-D) signals in telecommunication networks (telephone, computer, television networks and radio networks). The importance and relevance of the development and improvement of the methods of digital spectral processing of 1-D and 2-D signals is increasing due to the creation of telecommunication universal multiservice networks that can equally effectively transmit any type of information: data, sound and video.

The classical method of spatial-frequency processing of two-dimensional discrete signals is the 1-D and 2-D discrete Fourier transform (DFT), which allows to obtain a 1-D and 2-D frequency spectrum [1-13]. At the same time, there are a number of applications [9-12] where it is necessary to find the values of the frequency spectrum not at all 1-D and 2-D frequencies, but at a subset of them. In this case, the application of the full version of 1-D and 2-D DFT, even on the basis of the fast Fourier transform (FFT), becomes ineffective, since most of the obtained 1-D and 2-D DFT coefficients are not used. The solution to this problem for 1-D signals is given in [9].

The article discusses the solution of this problem for 2-D signals, introduces the concept of moving spatial-frequency processing based on 2-D DFT. Specifically, the vertical spatial-frequency processing of 2-D signals is considered.

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## 2 Direct two-dimensional discrete Fourier transform

Suppose we are given a discrete two-dimensional signal  $x(n_1, n_2)$  in the form of a two-dimensional sequence of finite length  $0 \leq n_1 \leq (N_1 - 1)$  and  $0 \leq n_2 \leq (N_2 - 1)$  or with a matrix of size  $n_1 \times n_2$  in a rectangular reference zone (specifically on a plane).

The direct two-dimensional discrete Fourier transform (2-D DFT) of a two-dimensional signal  $x(n_1, n_2)$  is a special case of a direct two-dimensional z-transform:

$$S_{N_1, N_2}(k_1, k_2) = X(z_1, z_2) \Big|_{z_1=W_{N_1}^{k_1}, z_2=W_{N_2}^{k_2}} ; \tag{1}$$

and can be specified both in algebraic and in matrix form.

**Algebraic form:**

$$\begin{aligned} S_{N_1, N_2}(k_1, k_2) &= \frac{1}{N_1 \cdot N_2} \sum_{n_1=0}^{N_1-1} \sum_{n_2=0}^{N_2-1} x(n_1, n_2) \exp \left[ -j2\pi \left( \frac{k_1 n_1}{N_1} + \frac{k_2 n_2}{N_2} \right) \right] = \\ &= \frac{1}{N_1 \cdot N_2} \sum_{n_1=0}^{N_1-1} \sum_{n_2=0}^{N_2-1} x(n_1, n_2) \cdot W_{N_1}^{k_1 n_1} \cdot W_{N_2}^{k_2 n_2} ; \end{aligned} \tag{2}$$

where  $k_1 = \overline{0, (N_1 - 1)}$ ,  $k_2 = \overline{0, (N_2 - 1)}$  are spatial frequencies;  $x(n_1, n_2)$  – two-dimensional signal;  $n_1 = \overline{0, N_1 - 1}$ ,  $n_2 = \overline{0, N_2 - 1}$  ;

$W_{N_2}^{k_2 n_2} = \exp(-j \frac{2\pi}{N_2} (k_2 n_2))$ ;  $S_{N_1, N_2}(k_1, k_2)$  – coefficients (bins) of 2-D DFT (two-dimensional vector spatial-frequency spectrum).

**Matrix form:**

$$S_{N_1 \times N_2} = \frac{1}{N_1 \cdot N_2} F_{N_1 \times N_1}^{(2)} \cdot X_{N_1 \times N_2} \cdot F_{N_2 \times N_2}^{(1)} ; \tag{3}$$

where

$$X_{N_1 \times N_2} = \begin{matrix} & & & & & n_2 \\ & & & & & 0 \\ & & & & & 1 \\ & & & & & \dots \\ & & & & & (N_2 - 1) \\ & & & & & \dots \\ & & & & & n_2 \\ & & & & & 0 \\ & & & & & 1 \\ & & & & & \dots \\ & & & & & (N_2 - 1) \\ & & & & & \dots \\ & & & & & n_2 \\ & & & & & 0 \\ & & & & & 1 \\ & & & & & \dots \\ & & & & & (N_2 - 1) \\ & & & & & \dots \\ & & & & & n_2 \end{matrix} \begin{bmatrix} x(0,0) & x(0,1) & \dots & x(0, N_2 - 1) \\ x(1,0) & x(1,1) & \dots & x(1, N_2 - 1) \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ x(N_1 - 1, 0) & x(N_1 - 1, 1) & \dots & x(N_1 - 1, N_2 - 1) \end{bmatrix} ; \tag{4}$$

$$F_{N_2 \times N_2}^{(1)} = \begin{matrix} & & & & & k_2 \\ & & & & & 0 \\ & & & & & 1 \\ & & & & & \dots \\ & & & & & (N_2 - 1) \\ & & & & & \dots \\ & & & & & k_2 \\ & & & & & 0 \\ & & & & & 1 \\ & & & & & \dots \\ & & & & & (N_2 - 1) \\ & & & & & \dots \\ & & & & & k_2 \\ & & & & & 0 \\ & & & & & 1 \\ & & & & & \dots \\ & & & & & (N_2 - 1) \\ & & & & & \dots \\ & & & & & k_2 \end{matrix} \begin{bmatrix} W_{N_2}^{0 \cdot 0} & W_{N_2}^{0 \cdot 1} & \dots & W_{N_2}^{0 \cdot (N_2 - 1)} \\ W_{N_2}^{1 \cdot 0} & W_{N_2}^{1 \cdot 1} & \dots & W_{N_2}^{1 \cdot (N_2 - 1)} \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ W_{N_2}^{(N_2 - 1) \cdot 0} & W_{N_2}^{(N_2 - 1) \cdot 1} & \dots & W_{N_2}^{(N_2 - 1) \cdot (N_2 - 1)} \end{bmatrix} ; \tag{5}$$

$$F_{N_1 \times N_1}^{(2)} = \begin{matrix} & & 0 & 1 & \dots & (N_1 - 1) & \\ & 0 & \left[ \begin{array}{cccc} W_{N_1}^{0,0} & W_{N_1}^{0,1} & \dots & W_{N_1}^{0,(N_1-1)} \\ W_{N_1}^{1,0} & W_{N_1}^{1,1} & \dots & W_{N_1}^{1,(N_1-1)} \\ \dots & \dots & \dots & \dots \\ W_{N_1}^{(N_1-1),0} & W_{N_1}^{(N_1-1),1} & \dots & W_{N_1}^{(N_1-1),(N_1-1)} \end{array} \right] & & & \\ & 1 & & & & & \\ & \dots & & & & & \\ & (N_1 - 1) & & & & & \\ & k_1 & & & & & \end{matrix} \cdot n_1 \quad (6)$$

Hereinafter there is no loss of generality in omitting the multiplier  $1/(N_1 \cdot N_2)$  in (3).

Due to the fact that for the product of the matrices (3) holds the associative property:

$$S_{N_1 \times N_2} = F_{N_1 \times N_1}^{(2)} \cdot \left[ X_{N_1 \times N_2} \cdot F_{N_2 \times N_2}^{(1)} \right] = \left[ F_{N_1 \times N_1}^{(2)} \cdot X_{N_1 \times N_2} \right] \cdot F_{N_2 \times N_2}^{(1)} \quad (7)$$

and the 2-D DFT core is separable, then, according to (7), you can get the 2-D DFT bins  $S_{N_1 \times N_2}$ ,  $k_1 = 0, (N_1 - 1)$ ,  $k_2 = 0, (N_2 - 1)$ , in two ways, each consists of two stages.

It is easy to find out that for obtaining  $S_{N_1, N_2}(k_1, k_2)$   $k_1 = 0, (N_1 - 1)$ ,  $k_2 = 0, (N_2 - 1)$  it is necessary to perform  $N_1 \cdot N_2$  one-dimensional 2-D DFTs, for calculating which the algorithms of the fast Fourier transform - FFT can be effectively applied.

## 2. Sliding spatial-frequency processing of discrete signals

Let us consider the spatial-frequency processing of two-dimensional discrete signals in a sliding spatial analysis window. In contrast to the one-dimensional case for the two-dimensional case, there are 4 possible ways of sliding of the spatial analysis window on the original two-dimensional discrete signal:

- 1<sup>st</sup> way consists of horizontal right shift (HS +) and horizontal left shift (HS-);
- 2<sup>nd</sup> way consists of vertical shift up (VS +) and vertical shift down (VS-);
- 3<sup>rd</sup> way consists of right diagonal shift up (RDS +) and right diagonal shift down (RDS-);
- 4<sup>th</sup> way consists of left diagonal shift up (LDS +) and left diagonal shift down (LDS-).

Fig. 1 shows a star diagram illustrating 4 types of sliding of a spatial analysis window using a discrete two-dimensional signal and examples of its shift.

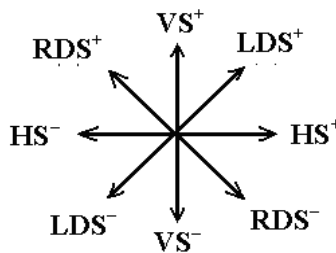


Fig. 1. The star chart illustrating 4 kinds of sliding of a spatial window.

Suppose we need to find the coefficient (bin) of a two-dimensional discrete transform  $S_{N_1, N_2}(k_1, k_2)$  at the spatial frequency  $(k_1, k_2)$  from the samples of the input signal  $x(n_1, n_2)$ . In this case, the matrix equation (3) is converted into the form:

$$S(k_1, k_2) = [W_{N_1}^{k_1 \cdot 0}, W_{N_1}^{k_1 \cdot 1}, \dots, W_{N_1}^{k_1 \cdot (N_1 - 1)}] \cdot \left\{ X_{N_1 \times N_2} \cdot \begin{bmatrix} W_{N_2}^{0 \cdot k_2} \\ W_{N_2}^{1 \cdot k_2} \\ \vdots \\ W_{N_2}^{(N_2 - 1) \cdot k_2} \end{bmatrix} \right\}. \quad (8)$$

At the first stage, according to (8), we multiply the basis function of frequency  $k_2$  and duration  $N_2$  by a matrix of a discrete two-dimensional signal  $x(n_1, n_2)$ . As a result, we obtain a columned matrix  $S_{N_2}(n_1, k_2)$  of size  $N_2$ , having spent on this procedure  $N_2 \cdot N_1$  complex multiplications and  $(N_2 - 1) \cdot N_1$  complex additions. Further, at the second stage, we multiply the basis function of frequency  $k_1$  and duration  $N_1$  by a columned matrix of size obtained at the first stage, having spent on this procedure  $N_1$  complex multiplications and  $(N_1 - 1)$  complex additions.

Thus, it is necessary to expend  $N_2 \cdot (N_1 + 1)$  complex multiplications and  $(N_2 - 1) \cdot (N_1 + 1)$  complex additions to obtain one coefficient of a two-dimensional discrete transform  $S_{N_1, N_2}(k_1, k_2)$  at spatial frequency  $(k_1, k_2)$ . Considering that performing one complex multiplication requires four real multiplications and two real additions, and one complex addition of two real additions, it is necessary to spend  $4 \cdot N_1 \cdot (N_2 + 1)$  real multiplications and  $4 \cdot N_1 \cdot N_2 + 2 \cdot (N_1 - 1)$  real additions to obtain the value of one coefficient of a two-dimensional discrete transformation  $S_{N_1, N_2}(k_1, k_2)$ .

Note that this amount of computation needs to be performed at each shift of a two-dimensional spatial analysis window using a two-dimensional signal. At the same time, it is easy to see that for any kind of shift of a two-dimensional signal a large number of values  $X_{N_1 \times N_2}$  of the complex matrix in the spatial analysis window remains unchanged.

Note that the shift of the spatial window along a two-dimensional discrete signal can be considered as a shift of a two-dimensional discrete signal in the spatial analysis window in the opposite direction to the movement of the spatial window.

### 3. The algorithm of vertical sliding processing of 2-D signals based on 2-D DFT

Let us need to find one coefficient (bin) of a two-dimensional discrete transformation  $S_{N_1, N_2}(k_1, k_2)$  at a spatial frequency  $(m_1, m_2)$  from the samples of the input signal  $x(n_1, n_2)$ . To obtain a bin, matrix equation (8) is converted to the form:

$$S_{N_1, N_2}(m_1, m_2) = [W_{N_1}^{m_1 \cdot 0}, W_{N_1}^{m_1 \cdot 1}, \dots, W_{N_1}^{m_1 \cdot (N_1 - 1)}] \cdot \{ X_{N_1 \times N_2} \cdot [W_{N_2}^{m_2 \cdot 0}, W_{N_2}^{m_2 \cdot 1}, \dots, W_{N_2}^{m_2 \cdot (N_2 - 1)}]^T \}; \quad (9)$$

where  $T$  is the transpose symbol.

The sequence of operations in the algorithm of vertical sliding processing of 2-D signals:

1. Find the column matrix  $S_{N_2}(n_1, m_2)$  by size  $N_1$ , by multiplying the basis function duration  $N_2$  and frequency  $m_2$  on a matrix of a discrete two-dimensional signal  $x(n_1, n_2)$ .
2. Remember the column matrix  $S_{N_2}(n_1, m_2)$  as a column matrix  $S_{N_2}^{(n_1, m_2)}(0)$ .
3. Calculate the 2-D coefficient of the discrete transform  $S_{N_1, N_2}(m_1, m_2)$  by multiplying the column matrix  $S_{N_2}(n_1, m_2)$  by a basis function of duration  $N_1$  and frequency  $m_1$ . This stage ends the output of the algorithm to the operating mode.
4. Carry out the VS – shift of the discrete spatial window by one sample down the two-dimensional signal  $x(n_1, n_2)$  and obtain the matrix of the discrete two-dimensional signal  $x[(n_1 - 1), n_2]$ .
5. Form a column matrix according to the ratio;

$$S_{N_2}^{(n_1, m_2)}(-1) = S_{N_2}^{((n_1+1), m_2)}(0); n_1 = \overline{0, N_1 - 2}$$

6. Calculate the coefficient value of a two-dimensional discrete transformation  $S_{N_1, N_2}(m_1, m_2)$  by multiplying the column matrix  $S_{N_2}^{(n_1, m_2)}(-1)$  by a basis function of duration  $N_1$  and frequency  $m_1$ .
7. Go to the implementation of paragraph No. 4.

## 4. Conclusion

The effectiveness of the proposed algorithm for vertical sliding processing of two-dimensional discrete signals in the spatial frequency domain in comparison with the standard method for obtaining the coefficient of two-dimensional discrete transformation  $S_{N_1, N_2}(m_1, m_2)$ :

- the standard algorithm (8) of the sliding processing of two-dimensional discrete signals in the space-frequency domain requires to obtain a coefficient  $S_{N_1, N_2}(m_1, m_2)$  of two-dimensional discrete conversion of  $4 \cdot N_1 \cdot (N_2 + 1)$  real multiplications  $4 \cdot N_1 \cdot N_2 + 2 \cdot (N_1 - 1)$  of real additions;
- the developed algorithm for the sliding processing of two-dimensional discrete signals in the space-frequency region after reaching the operating mode requires  $4 \cdot (N_1 + N_2)$  real multiplications and  $4 \cdot (N_1 + N_2 - 1)$  real additions to obtain a two-dimensional discrete transform coefficient  $S_{N_1, N_2}(m_1, m_2)$ .

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