Digital Image Processing

Filtering in the Frequency Domain (Circulant Matrices and Convolution)

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Toeplitz matrices

- Elements with constant value along the main diagonal and sub-diagonals.
- For a NxN matrix, its elements are determined by a (2N-1)-length sequence $\{t_n \mid -(N-1) \le n \le N-1\}$

$$\mathbf{T}(m,n) = t_{m-n}$$

$$\mathbf{T} = \begin{bmatrix} t_0 & t_{-1} & t_{-2} & \dots & t_{-(N-1)} \\ t_1 & t_0 & t_{-1} & \ddots & \vdots \\ t_2 & \ddots & \ddots & t_{-2} \\ \vdots & \ddots & \ddots & \vdots \\ t_{N-1} & \dots & t_2 & t_1 & t_0 \end{bmatrix}_{N \times N}$$

Toeplitz matrices (cont.)

- Each row (column) is generated by a shift of the previous row (column).
 - The last element disappears.
 - A new element appears.

$$\mathbf{T}(m,n) = t_{m-n}$$

$$\mathbf{T} = \begin{bmatrix} t_0 & t_{-1} & t_{-2} & \dots & t_{-(N-1)} \\ t_1 & t_0 & t_{-1} & \ddots & \vdots \\ t_2 & \ddots & \ddots & \ddots & t_{-2} \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ t_{N+1} & \dots & t_2 & t_1 & t_0 \end{bmatrix}_{N \times N}$$

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Circulant matrices

- Elements with constant value along the main diagonal and sub-diagonals.
- For a NxN matrix, its elements are determined by a N-length sequence $\{c_n \mid 0 \le n \le N-1\}$

$$\mathbf{C}(m,n) = c_{(m-n) \bmod N}$$

$$\mathbf{C} = \begin{bmatrix} c_0 & c_{N-1} & c_{N-2} & \dots & c_1 \\ c_1 & c_0 & c_{N-1} & \dots & \vdots \\ c_2 & c_1 & c_0 & \dots & c_{N-2} \\ \vdots & \vdots & \vdots & \dots & c_{N-1} \\ c_{N-1} & c_{N-2} & \dots & c_1 & c_0 \end{bmatrix}_{N \times N}$$

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Circulant matrices (cont.)

- Special case of a Toeplitz matrix.
- Each row (column) is generated by a circular shift (modulo N) of the previous row (column).

$$\mathbf{C}(m,n) = c_{(m-n) \bmod N}$$

$$\mathbf{C} = \begin{bmatrix} c_0 & c_{N-1} & c_{N-2} & \dots & c_1 \\ c_1 & c_0 & c_{N-1} & \dots & \vdots \\ c_2 & c_1 & c_0 & \dots & c_{N-2} \\ \vdots & \vdots & \vdots & \dots & c_{N-1} \\ c_{N-1} & c_{N-2} & \dots & c_1 & c_0 \end{bmatrix}_{N \times N}$$

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Convolution by matrix-vector operations

- 1-D linear convolution between two discrete signals may be expressed as the product of a Toeplitz matrix constructed by the elements of one of the signals and a vector constructed by the elements of the other signal.
- 1-D circular convolution between two discrete signals may be expressed as the product of a circulant matrix constructed by the elements of one of the signals and a vector constructed by the elements of the other signal.
- Extension to 2D signals.

1D linear convolution using Toeplitz matrices

$$f[n] = \{\underline{1}, 2, 2\}, h[n] = \{\underline{1}, -1\}, N_1 = 3, N_2 = 2$$

- The linear convolution g[n]=f[n]*h[n] will be of length $N=N_1+N_2-1=3+2-1=4$.
- We create a Toeplitz matrix \mathbf{H} from the elements of h[n] (zero-padded if needed) with
 - N=4 lines (the length of the result).
 - N_1 =3 columns (the length of f(n)).
 - The two signals may be interchanged.

1D linear convolution using Toeplitz matrices (cont.)

$$f[n] = \{\underline{1}, 2, 2\}, h[n] = \{\underline{1}, -1\}, N_1 = 3, N_2 = 2$$

$$\mathbf{H} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \\ 0 & 0 & -1 \end{bmatrix}_{4\times 3}$$
Length of $f[n] = 3$

Length of the result =4

Notice that *H* is not circulant (e.g. a -1 appears in the second line which is not present in the first line.

Zero-padded h[n] in the first column

1D linear convolution using Toeplitz matrices (cont.)

$$f[n] = \{\underline{1}, 2, 2\}, h[n] = \{\underline{1}, -1\}, N_1 = 3, N_2 = 2$$

$$\mathbf{g} = \mathbf{Hf} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \\ -2 \end{bmatrix}$$

$$g[n] = \{\underline{1}, 1, 0, -2\}$$

1D circular convolution using circulant matrices

$$f[n] = \{\underline{1}, 2, 2\}, h[n] = \{\underline{1}, -1\}, N_1 = 3, N_2 = 2$$

- The circular convolution $g[n]=f[n] \circ h[n]$ will be of length $N=\max\{N_1,N_2\}=3$.
- We create a circulant matrix H from the elements of h [n] (zero-padded if needed) of size NxN.
 - The two signals may be interchanged.

1D circular convolution using circulant matrices (cont.)

$$f[n] = \{\underline{1}, 2, 2\}, h[n] = \{\underline{1}, -1\}, N_1 = 3, N_2 = 2$$

$$\mathbf{H} = \begin{bmatrix} 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix}_{3\times 3}$$

Zero-padded h[n] in the first column

1D circular convolution using circulant matrices (cont.)

$$f[n] = \{\underline{1}, 2, 2\}, h[n] = \{\underline{1}, -1\}, N_1 = 3, N_2 = 2$$

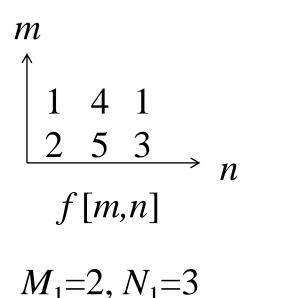
$$\mathbf{g} = \mathbf{Hf} = \begin{bmatrix} 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

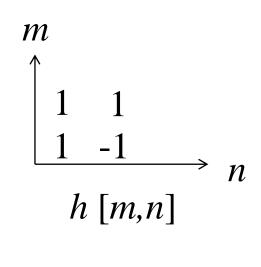
$$g[n] = \{-1, 1, 0\}$$

Block matrices

$$A = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2N} \\ \vdots & \vdots & \vdots & \vdots \\ A_{M1} & A_{M2} & \dots & A_{MN} \end{bmatrix}$$

- A_{ii} are matrices.
- If the structure of A, with respect to its sub-matrices, is Toeplitz (circulant) then matrix A is called **block-Toeplitz** (**block-circulant**).
- If each individual A_{ij} is also a Toeplitz (circulant) matrix then A is called **doubly block-Toeplitz** (**doubly block-circulant**).

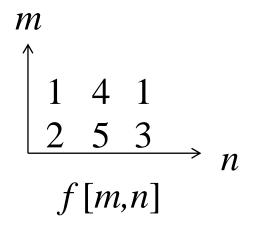


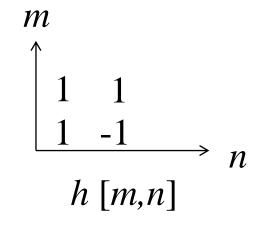


$$M_2=2, N_2=2$$

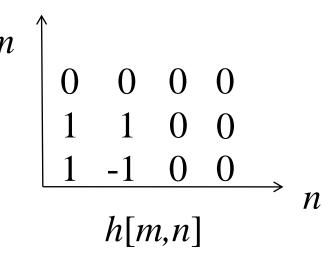
The result will be of size $(M_1+M_2-1) \times (N_1+N_2-1) = 3 \times 4$

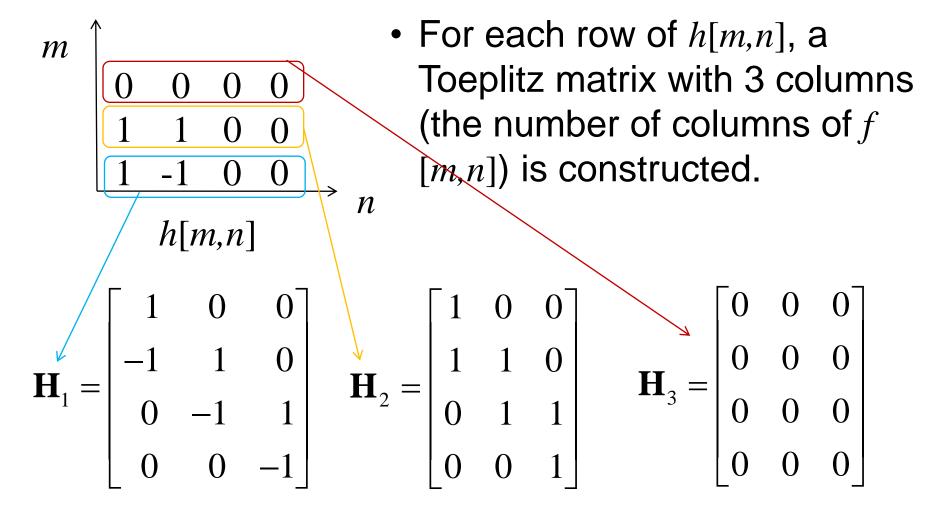
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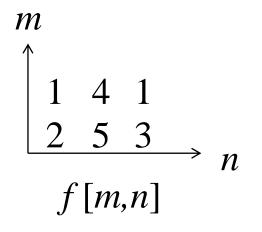


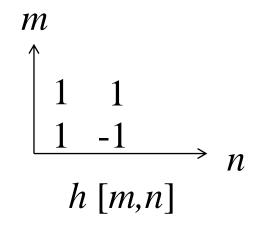


- At first, h[m,n] is zero-padded to 3×4 (the size of the result).
- Then, for each row of h[m,n], a Toeplitz matrix with 3 columns (the number of **columns** of f[m,n]) is constructed.



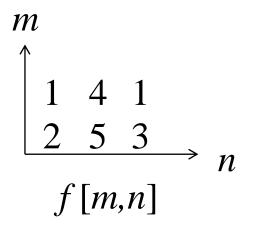






Using matrices H₁, H₂ and H₃ as elements, a doubly block Toeplitz matrix H is then constructed with 2 columns (the number of rows of f[m,n]).

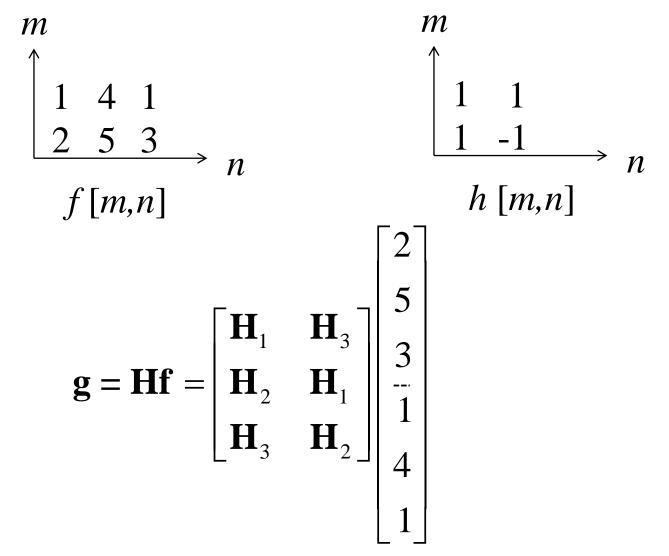
$$\mathbf{H} = \begin{bmatrix} \mathbf{H}_1 & \mathbf{H}_3 \\ \mathbf{H}_2 & \mathbf{H}_1 \\ \mathbf{H}_3 & \mathbf{H}_2 \end{bmatrix}_{12 \times 6}$$



 $\begin{array}{c|c}
 & 1 & 1 \\
\hline
 & 1 & -1 \\
\hline
 & h [m,n]
\end{array}$

We now construct
 a vector from the
 elements of f [m,n].

$$\mathbf{f} = \begin{bmatrix} 2 \\ 5 \\ 3 \\ -1 \\ 1 \\ 4 \\ 1 \end{bmatrix} = \begin{bmatrix} (2 & 5 & 3)^T \\ (1 & 4 & 1)^T \end{bmatrix}$$



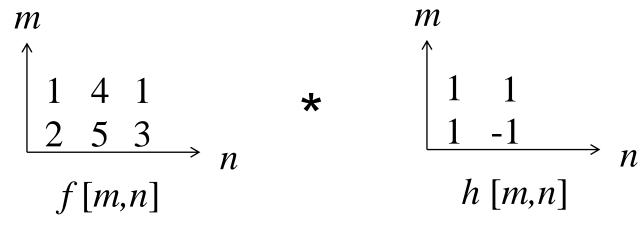
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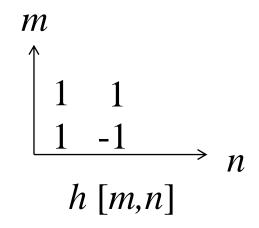
$$\mathbf{g} = \mathbf{H}\mathbf{f} = \begin{bmatrix} -1 & 1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 1 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 & 0 \\ \hline 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & -1 & 1 & 0 \\ 0 & 1 & 1 & 0 & -1 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & -1 \\ \hline 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 5 \\ 3 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} (2 & 3 & -2 & 3)^T \\ 3 \\ 10 \\ 5 \\ 2 \\ -1 \\ 1 \end{bmatrix} = \begin{bmatrix} (2 & 3 & -2 & 3)^T \\ \hline (3 & 10 & 5 & 2)^T \\ \hline (1 & 5 & 5 & 1)^T \end{bmatrix}$$

0

0

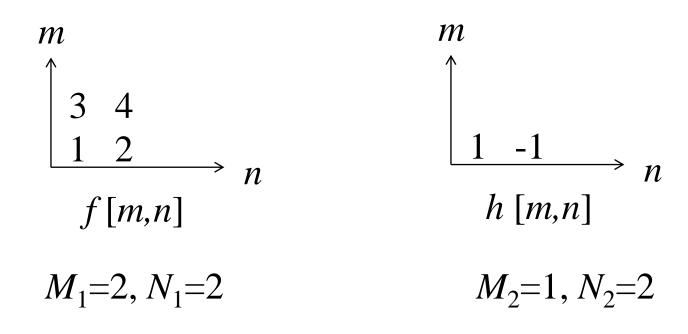
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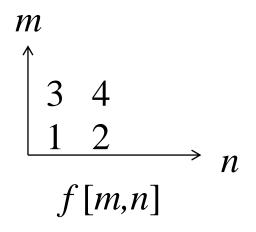


$$\mathbf{g} = \begin{bmatrix} (2 & 3 & -2 & 3)^T \\ \hline (3 & 10 & 5 & 2)^T \\ \hline (1 & 5 & 5 & 1)^T \end{bmatrix}$$

Another example



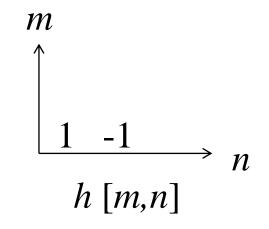
The result will be of size $(M_1+M_2-1) \times (N_1+N_2-1) = 2 \times 3$

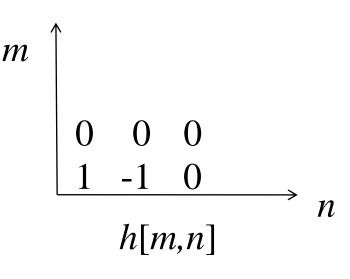


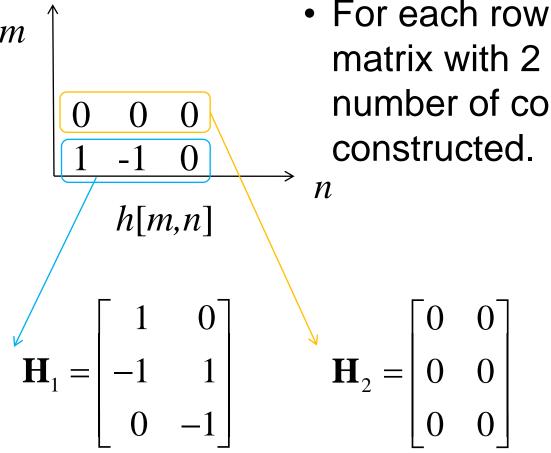
- At first, h[m,n] is zero-padded to 2 x 3 (the size of the result).
- Then, for each line of h[m,n], a Toeplitz matrix with 2 columns (the number of **columns** of f[m,n]) is constructed.

 Then, for each line of h[m,n], a Toeplitz matrix with 2

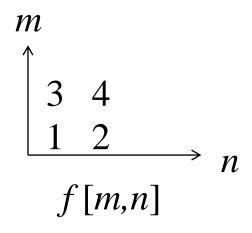
 Columns of f[m,n] is constructed.



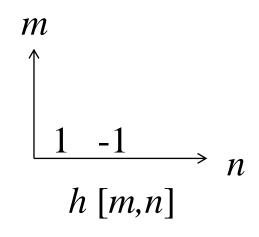




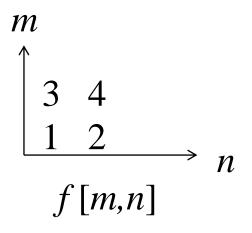
• For each row of h[m,n], a Toeplitz matrix with 2 columns (the number of columns of f[m,n]) is constructed.

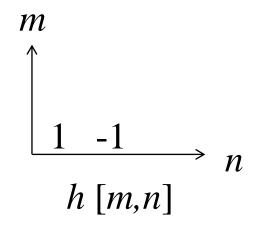


Using matrices H₁ and H₂ as elements, a doubly block Toeplitz matrix H is then constructed with 2 columns (the number of rows of f[m,n]).



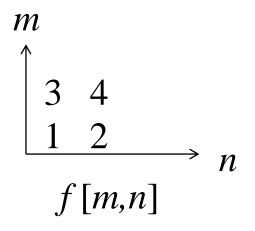
$$\mathbf{H} = \begin{bmatrix} \mathbf{H}_1 & \mathbf{H}_2 \\ \mathbf{H}_2 & \mathbf{H}_1 \end{bmatrix}_{6 \times 4}$$





 We now construct a vector from the elements of f [m,n].

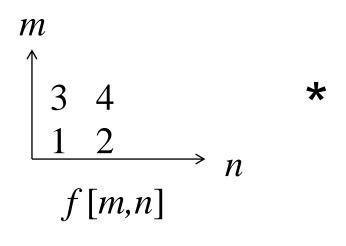
$$\mathbf{f} = \begin{bmatrix} 1 \\ 2 \\ -3 \\ 4 \end{bmatrix} = \begin{bmatrix} (1 & 2)^T \\ -(3 & 4)^T \end{bmatrix}$$

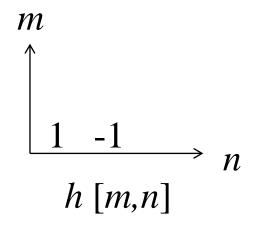


$$\begin{array}{c}
m \\
\uparrow \\
1 & -1 \\
h & [m,n]
\end{array}$$

$$\mathbf{g} = \mathbf{H}\mathbf{f} = \begin{bmatrix} \mathbf{H}_1 & \mathbf{H}_2 \\ \mathbf{H}_2 & \mathbf{H}_1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ \vdots \\ 3 \end{bmatrix}$$

$$\mathbf{g} = \mathbf{H}\mathbf{f} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ \hline 0 & 0 & 1 & 0 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ -2 \\ 3 \\ 1 \\ -4 \end{bmatrix} = \begin{bmatrix} (1 & 1 & -2)^T \\ (3 & 1 & -4)^T \end{bmatrix}$$





$$= \begin{array}{c} m & \uparrow \\ \hline & 3 & 1 & 4 \\ \hline & 1 & 1 & -2 \\ \hline & g & [m,n] \end{array}$$

$$\mathbf{g} = \begin{bmatrix} (1 & 1 & -2)^T \\ ---- & (3 & 1 & -4)^T \end{bmatrix}$$

The circular convolution g[m,n]=f[m,n] $\bigcirc h[m,n]$ with $0 \le m \le M-1, 0 \le n \le N-1,$

may be expressed in matrix-vector form as:

$$g = Hf$$

where **H** is a doubly block circulant matrix generated by h[m,n] and **f** is a vectorized form of f[m,n].

$$\mathbf{H} = \begin{bmatrix} \mathbf{H}_0 & \mathbf{H}_{M-1} & \mathbf{H}_{M-2} & \dots & \mathbf{H}_1 \\ \mathbf{H}_1 & \mathbf{H}_0 & \mathbf{H}_{M-1} & \dots & \mathbf{H}_2 \\ \mathbf{H}_2 & \mathbf{H}_1 & \mathbf{H}_0 & \dots & \mathbf{H}_3 \\ \vdots & \vdots & \vdots & \vdots & \dots & \vdots \\ \mathbf{H}_{M-1} & \mathbf{H}_{M-2} & \mathbf{H}_{M-3} & \dots & \mathbf{H}_0 \end{bmatrix}$$

Each \mathbf{H}_{j} , for j=1,..M, is a circulant matrix with N columns (the number of **columns** of f[m,n]) generated from the elements of the j-th **row** of h[m,n].

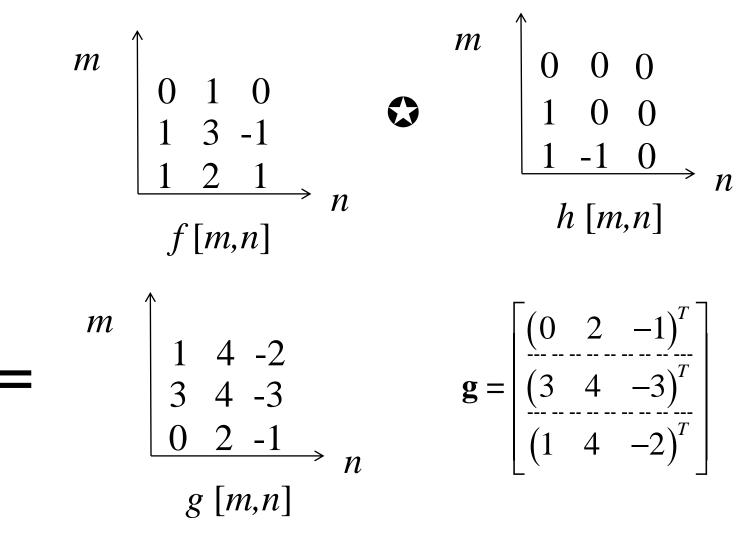
$$\mathbf{H}_{j} = \begin{bmatrix} h[j,0] & h[j,N-1] & \dots & h[j,1] \\ h[j,1] & h[j,0] & \dots & h[j,2] \\ \vdots & \vdots & \vdots & \vdots \\ h[j,N-1] & h[j,N-2] & \dots & h[j,0] \end{bmatrix}_{N\times N}$$

Each \mathbf{H}_{j} , for j=1,..M, is a $N\mathbf{x}N$ circulant matrix generated from the elements of the j-th **row** of h [m,n].

$$\mathbf{H}_{0} = \begin{bmatrix} 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \quad \mathbf{H}_{1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \mathbf{H}_{2} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\mathbf{g} = \mathbf{H}\mathbf{f} = \begin{bmatrix} \mathbf{H}_0 & \mathbf{H}_2 & \mathbf{H}_1 \\ \mathbf{H}_1 & \mathbf{H}_0 & \mathbf{H}_2 \\ \mathbf{H}_2 & \mathbf{H}_1 & \mathbf{H}_0 \end{bmatrix} \begin{bmatrix} (1 & 2 & 1)^T \\ (1 & 3 & -1)^T \\ (0 & 1 & 0)^T \end{bmatrix}$$

$$\mathbf{g} = \mathbf{H}\mathbf{f} = \begin{bmatrix} 1 & 0 & -1 & 0 & 0 & 0 & 1 & 0 & 0 \\ -1 & 1 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ \hline 1 & 0 & 0 & 1 & 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & -1 & 1 & 1 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 1 & 0 & -1 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 1 & 0 & -1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & -1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & -1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \\ 1 \\ 3 \\ 3 \\ 4 \\ -1 \\ 0 \end{bmatrix}$$



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Theorem: The columns of the inverse DFT matrix are eigenvectors of any circulant matrix. The corresponding eigenvalues are the DFT values of the signal generating the circulant matrix.

Proof: Let

$$w_N = e^{-j\frac{2\pi}{N}} \iff w_N^{nk} = e^{-j\frac{2\pi n}{N}k}$$

be the DFT basis elements of length N with:

$$0 \le k \le N-1, \ 0 \le n \le N-1,$$

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We know that the DFT F[k] of a 1D signal f[n] may be expressed in matrix-vector form:

$$\mathbf{F} = \mathbf{Af}$$

where

$$\mathbf{f} = [f[0], f[1], ..., f[N-1]]^T, \mathbf{F} = [F[0], F[1], ..., F[N-1]]^T$$

$$\mathbf{A} = \begin{bmatrix} \left(w_{N}^{0}\right)^{0} & \left(w_{N}^{0}\right)^{1} & \left(w_{N}^{0}\right)^{2} & \dots & \left(w_{N}^{0}\right)^{N-1} \\ \left(w_{N}^{1}\right)^{0} & \left(w_{N}^{1}\right)^{1} & \left(w_{N}^{1}\right)^{2} & \dots & \left(w_{N}^{1}\right)^{N-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \left(w_{N}^{N-1}\right)^{0} & \left(w_{N}^{N-1}\right)^{1} & \left(w_{N}^{N-1}\right)^{2} & \dots & \left(w_{N}^{N-1}\right)^{N-1} \end{bmatrix}$$

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The inverse DFT is then expressed by:

$$f = A^{-1}F$$

where
$$\mathbf{A}^{-1} = \frac{1}{N} (\mathbf{A}^*)^T = \frac{1}{N} \begin{bmatrix} \left(w_N^0 \right)^0 & \left(w_N^0 \right)^1 & \left(w_N^0 \right)^2 & \dots & \left(w_N^0 \right)^{N-1} \\ \left(w_N^1 \right)^0 & \left(w_N^1 \right)^1 & \left(w_N^1 \right)^2 & \dots & \left(w_N^1 \right)^{N-1} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \left(w_N^{N-1} \right)^0 & \left(w_N^{N-1} \right)^1 & \left(w_N^{N-1} \right)^2 & \dots & \left(w_N^{N-1} \right)^{N-1} \end{bmatrix}^* \end{bmatrix}^T$$

The theorem implies that any circulant matrix has eigenvectors the columns of \mathbf{A}^{-1} .

Let **H** be a $N \times N$ circulant matrix generated by the 1D N-length signal h[n], that is:

$$\mathbf{H}(m,n) = h[(m-n)_{\text{mod }N}] \triangleq h[m-n]_N$$

Let also α_k be the k-th column of the inverse DFT matrix \mathbf{A}^{-1} .

We will prove that α_k , for any k, is an eigenvector of \mathbf{H} .

The *m*-th element of the vector $\mathbf{H}\boldsymbol{\alpha}_{k}$, denoted by $\left[\mathbf{H}\boldsymbol{\alpha}_{k}\right]_{m}$

is the result of the circular convolution of the signal h[n] with α_k .

$$\left[\mathbf{H}\alpha_{k}\right]_{m} = \sum_{n=0}^{N-1} h[m-n]_{N} \alpha_{k}[n] = \frac{1}{N} \sum_{n=0}^{N-1} h[m-n]_{N} w_{N}^{-kn}$$

$$\stackrel{l=m-n}{=} \frac{1}{N} \sum_{l=m}^{m-(N-1)} h[l]_N w_N^{-k(m-l)} = \frac{1}{N} w_N^{-km} \sum_{l=m}^{m-(N-1)} h[l]_N w_N^{kl}$$

$$= \frac{1}{N} w_N^{-km} \left[\sum_{l=m-(N-1)}^{-1} h[l]_N w_N^{kl} + \sum_{l=0}^{m} h[l]_N w_N^{kl} \right]$$

We will break it into two parts

$$= \frac{1}{N} w_N^{-km} \left[\sum_{l=m-(N-1)}^{-1} h[l]_N w_N^{kl} + \sum_{l=0}^{N-1} h[l]_N w_N^{kl} - \sum_{l=m+1}^{N-1} h[l]_N w_N^{kl} \right]$$

Periodic with respect to N.

$$= \frac{1}{N} w_N^{-km} \left[\sum_{l=N-m-(N-1)}^{N-1} h[l]_N w_N^{kl} + \sum_{l=0}^{N-1} h[l]_N w_N^{kl} - \sum_{l=m+1}^{N-1} h[l]_N w_N^{kl} \right]$$

$$=\frac{1}{N}w_{N}^{-km}\left[\sum_{l=m+1}^{N-1}h[U]_{N}w_{N}^{kl}+\sum_{l=0}^{N-1}h[l]_{N}w_{N}^{kl}-\sum_{l=m+1}^{N-1}h[U]_{N}w_{N}^{kl}\right]\iff$$

$$\left[\mathbf{H}\boldsymbol{\alpha}_{k}\right]_{m} = \frac{1}{N} w_{N}^{-km} \left[\sum_{l=0}^{N-1} h[l]_{N} w_{N}^{kl}\right] = H[k] \left[\boldsymbol{\alpha}_{k}\right]_{m}$$
DFT of $h[n]$ at k .

This holds for any value of m. Therefore:

$$\mathbf{H}\mathbf{\alpha}_{k} = H[k]\mathbf{\alpha}_{k}$$

which means that α_k , for any k, is an eigenvector of \mathbf{H} with corresponding eigenvalue the k-th element of H[k], the DFT of the signal generating \mathbf{H} .

The above expression may be written in terms of the inverse DFT matrix:

$$\mathbf{H}\mathbf{A}^{-1} = \mathbf{A}^{-1}\mathbf{\Lambda} \iff \mathbf{H} = \mathbf{A}^{-1}\mathbf{\Lambda}\mathbf{A}$$

$$\mathbf{\Lambda} = \operatorname{diag}\{H[0], H[1], ..., H[N-1]\}$$

or equivalently: $\Lambda = AHA^{-1}$

Based on this diagonalization, we can prove the property between circular convolution and DFT.

$$\mathbf{g} = \mathbf{H}\mathbf{f} \Leftrightarrow \mathbf{g} = \mathbf{H}\mathbf{A}^{-1}\mathbf{A}\mathbf{f} \quad \Leftrightarrow \mathbf{A}\mathbf{g} = \mathbf{A}\mathbf{H}\mathbf{A}^{-1}\mathbf{A}\mathbf{f} \quad \Leftrightarrow \mathbf{G} = \mathbf{A}\mathbf{F}$$

$$\Leftrightarrow \begin{bmatrix} G[0] \\ G[1] \\ \vdots \\ G[N-1] \end{bmatrix} = \begin{bmatrix} H[0] & 0 & \dots & 0 \\ 0 & H[1] & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & H[N-1] \end{bmatrix} \begin{bmatrix} F[0] \\ F[1] \\ \vdots \\ F[N-1] \end{bmatrix}$$

$$\downarrow \qquad \qquad \downarrow$$
DFT of $g[n]$ DFT of $h[n]$ DFT of $f[n]$

Diagonalization of doubly block circulant matrices

- These properties may be generalized in 2D.
- We need to define the Kronecker product:

$$\mathbf{A} \in \mathbb{R}^{M \times N}$$
, $\mathbf{B} \in \mathbb{R}^{K \times L}$

$$\mathbf{A} \otimes \mathbf{B} = \begin{bmatrix} a_{11}\mathbf{B} & a_{12}\mathbf{B} & \dots & a_{1N}\mathbf{B} \\ a_{21}\mathbf{B} & a_{22}\mathbf{B} & \dots & a_{2N}\mathbf{B} \\ \vdots & \vdots & \vdots & \vdots \\ a_{M1}\mathbf{B} & a_{M2}\mathbf{B} & \dots & a_{MN}\mathbf{B} \end{bmatrix}_{MK \times NL}$$

Diagonalization of doubly block circulant matrices (cont.)

• The 2D signal f[m,n], $0 \le m \le M-1$, $0 \le n \le N-1$, may be vectorized in lexicographic ordering (stacking one column after the other) to a vector:

$$\mathbf{f} \in \mathbb{R}^{MN imes 1}$$

• The DFT of f[m,n], may be obtained in matrix-vector form:

$$F = (\mathbf{A} \otimes \mathbf{A})\mathbf{f}$$

Diagonalization of doubly block circulant matrices (cont.)

Theorem: The columns of the inverse 2D DFT matrix

$$(\mathbf{A} \otimes \mathbf{A})^{-1}$$

are eigenvectors of any doubly block circulant matrix. The corresponding eigenvalues are the 2D DFT values of the 2D signal generating the doubly block circulant matrix:

$$\mathbf{\Lambda} = (\mathbf{A} \otimes \mathbf{A}) \mathbf{H} (\mathbf{A} \otimes \mathbf{A})^{-1}$$

Diagonal, containing the 2D DFT Doubly block circulant of h[m,n] generating **H**

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