

Introduction to Data Processing and representation

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HW3

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Theory

1. Circulant Matrices

a. Considering the matrix $J = \begin{pmatrix} 0 & \dots & \dots & 0 & 1 \\ 1 & 0 & \ddots & \ddots & 0 \\ 0 & 1 & 0 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & 1 & 0 \end{pmatrix} \in \mathbb{R}^{n \times n}$, we want to compute J^k ,

for $k \in \mathbb{N}$:

Let's check the behavior of J^k for $k = 2$

$$J^2 = J^1 \cdot J = \begin{pmatrix} 0 & \dots & \dots & 0 & 1 \\ 1 & 0 & \ddots & \ddots & 0 \\ 0 & 1 & 0 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & \dots & \dots & 0 & 1 \\ 1 & 0 & \ddots & \ddots & 0 \\ 0 & 1 & 0 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & \dots & \dots & 1 & 0 \\ 0 & 0 & \ddots & \ddots & 1 \\ 1 & 0 & 0 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 1 & 0 & 0 \end{pmatrix}$$

As can be seen from the outcome above, we got matrix J with one cyclic shift of the rows downward or $n - 1$ cyclic shifts of the rows upward. Thus for $k \in \mathbb{N}$ we get $J^k = J^{k-1}J$, meaning the result is matrix J with $k - 1$ cyclic shifts of the rows downward. For $k = n$ we get $J^n = J^{n-1}J$, meaning the result is matrix J with $n - 1$ cyclic shifts of the rows downward or one cyclic shift of the rows upwards. The first row of J would be the n^{th} row of the outcome matrix now, after shifting all the rows we get

$$J^n = \begin{pmatrix} 1 & \dots & \dots & 0 & 0 \\ 0 & 1 & \ddots & \ddots & 0 \\ 0 & 0 & 1 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & 0 & 1 \end{pmatrix} = I_{n \times n}$$

This means, for $k = n$ the result is the identity matrix of order n .

J is called permutation matrix which means a square binary matrix that has exactly one entry of 1 in each row and each column and 0s elsewhere.

b. We want to compute the eigenvalues of J

Using the definition, if there is a vector $V \in \mathbb{R}^n \neq 0$ such that

$$JV = \lambda V$$

For some scalar λ , then λ is called the eigenvalues of J with corresponding eigenvector V .

This is equivalent to $(J - \lambda I)V = 0$, and nontrivial solution is available iff the determinant vanishes, so the solutions of $(J - \lambda I)V = 0$ are given by

$$\det(J - \lambda I) = 0$$

By applying the determinant of upper or lower triangular matrix is the product of all the diagonal elements of the matrix, we get

$$\det(J - \lambda I) = 0$$

$$\begin{aligned} \det(J - \lambda I) &= \det \begin{pmatrix} -\lambda & \cdots & \cdots & 0 & 1 \\ 1 & -\lambda & \ddots & \ddots & 0 \\ 0 & 1 & -\lambda & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & 1 & -\lambda \end{pmatrix} \\ &= -\lambda \cdot \det \underbrace{\begin{pmatrix} -\lambda & \cdots & 0 & 0 \\ 1 & -\lambda & \ddots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 1 & -\lambda \end{pmatrix}}_{\text{lower triangular matrix}} + (-1)^{n+1} \cdot 1 \cdot \det \underbrace{\begin{pmatrix} 1 & -\lambda & \cdots & 0 \\ 0 & 1 & -\lambda & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & 1 \end{pmatrix}}_{\text{upper triangular matrix}} \\ &= (-\lambda)^n + (-1)^{n+1} \\ &\Rightarrow (-\lambda)^n + (-1)^{n+1} = 0 \\ &\Rightarrow \lambda^n = -(-1) = 1 \\ &\lambda^n = 1e^{i0} \end{aligned}$$

Using the polar form, we denote λ by $\lambda = re^{i\theta}$, then we get

$$r^n e^{in\theta} = 1e^{i0}$$

$$r^n = 1 \Rightarrow r = 1$$

$$n\theta = 0 + 2\pi k \text{ for } k = 0, 1, 2, \dots, n-1$$

$$\theta = +\frac{2\pi k}{n} \text{ for } k = 0, 1, 2, \dots, n-1$$

Since every $n \times n$ matrix has exactly n complex eigenvalues, we get that the n eigenvalues of J are given by:

$$\lambda_k = e^{i\frac{2\pi k}{n}} \text{ for } k = 0, 1, 2, \dots, n-1$$

- c. To complete the eigendecomposition of J , we need to compute the corresponding eigenvectors of the previously computed eigenvalues in (b):

We get

$$(J - \lambda_r I)V_r = 0 \text{ for } r = 0, \dots, n-1$$

$$\begin{pmatrix} -\lambda_r & \dots & \dots & 0 & 1 \\ 1 & -\lambda_r & \ddots & \ddots & 0 \\ 0 & 1 & -\lambda_r & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & 1 & -\lambda_r \end{pmatrix} \begin{pmatrix} v_{r_1} \\ v_{r_2} \\ v_{r_3} \\ \vdots \\ v_{r_n} \end{pmatrix} = 0$$

$$\begin{pmatrix} -\lambda_r v_{r_1} + v_{r_n} \\ -\lambda_r v_{r_2} + v_{r_1} \\ -\lambda_r v_{r_3} + v_{r_2} \\ \vdots \\ -\lambda_r v_{r_n} + v_{r_{n-1}} \end{pmatrix} = 0$$

For $v_{r_1} = 1 = \lambda_r^0$ (since eigenvectors are up to multiplication by a scalar), we get

$$\begin{cases} v_{r_1} = 1 \\ v_{r_2} = \lambda_r^{-1} \\ v_{r_3} = \lambda_r^{-2} \\ \vdots \\ v_{r_n} = \lambda_r^{-(n-1)} = \lambda_r^1 \end{cases}$$

Thus, we get

$$V_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix}, V_2 = \begin{pmatrix} 1 \\ e^{-i\frac{2\pi}{n}} \\ e^{-i\frac{2\pi \cdot 2}{n}} \\ \vdots \\ e^{-i\frac{2\pi(n-1)}{n}} \end{pmatrix}, \dots, V_n = \begin{pmatrix} 1 \\ e^{-i\frac{2\pi(n-1)}{n}} \\ e^{-i\frac{2\pi(n-1) \cdot 2}{n}} \\ \vdots \\ e^{-i\frac{2\pi(n-1) \cdot (n-1)}{n}} \end{pmatrix}$$

Therefore, J is diagonalizable and can be decomposed to the form

$$J = PDP^{-1}$$

Where P is a matrix composed of the eigenvectors of J , D is the diagonal matrix constructed from the corresponding eigenvalues and P^{-1} is the matrix inverse of P .

Thus, P is of the form

$$P = \begin{pmatrix} | & | & \cdots & | \\ V_1 & V_2 & \cdots & V_n \\ | & | & \cdots & | \end{pmatrix}$$

And D is of the form

$$D = \begin{pmatrix} \lambda_0 & \cdots & \cdots & 0 & 0 \\ 0 & \lambda_1 & \ddots & \ddots & 0 \\ 0 & 0 & \lambda_2 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & 0 & \lambda_{n-1} \end{pmatrix}$$

We can diagonalize J in a unitary basis, since we can get U by normalizing the matrix P such that $U = \frac{1}{\sqrt{n}} P$ (eigenvectors are up to multiplication by a scalar), U is symmetric and unitary since its rows and columns obey

$$\sum_{l=0}^{n-1} U^{kl} (U^{rl})^* = \sum_{l=0}^{n-1} U^{(k-r)l} = \frac{1}{n} \sum_{l=0}^{n-1} e^{-\frac{i2\pi}{n}(k-r)l} = \begin{cases} 1 & k = r \\ 0 & k \neq r \end{cases}$$

and which makes the following statement true

$$U^* J U = \begin{pmatrix} \lambda_0 & \cdots & \cdots & 0 & 0 \\ 0 & \lambda_1 & \ddots & \ddots & 0 \\ 0 & 0 & \lambda_2 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & 0 & \lambda_{n-1} \end{pmatrix}$$

d. Considering the general circulant matrix $H = \begin{pmatrix} h_0 & h_{n-1} & h_{n-2} & \cdots & h_1 \\ h_1 & h_0 & h_{n-1} & \ddots & h_2 \\ h_2 & h_1 & h_0 & \ddots & h_3 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ h_{n-1} & h_{n-2} & h_{n-3} & \cdots & h_0 \end{pmatrix}.$

And recalling that J is given by $J = \begin{pmatrix} 0 & \cdots & \cdots & 0 & 1 \\ 1 & 0 & \ddots & \ddots & 0 \\ 0 & 1 & 0 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & 1 & 0 \end{pmatrix} \in \mathbb{R}^{n \times n}.$

Using the results of (a) we get:

$$H = h_1 J + h_2 J^2 + \cdots + h_{n-1} J^{n-1} + h_0 J^n = h_0 I_n + h_1 J + h_2 J^2 + \cdots + h_{n-1} J^{n-1}$$

Thus, the polynomial P on matrix J is given by

$$P(J) = h_0 I_n + h_1 J + h_2 J^2 + \cdots + h_{n-1} J^{n-1}$$

Therefore, $H = P(J)$

- e. We got that $H = P(J)$, by applying this we get that for all eigenvalue of matrix J , λ_k when $k = 1, \dots, n$, $P(\lambda_k)$ is an eigenvalue of H with the same eigenvector. To see that we first show that if λ is an eigenvalue of matrix A then λ^k is the eigenvalue of the matrix A^k , since λ is the eigenvalue of A then $AV = \lambda V$ as V is the corresponding eigenvector. The eigenvalue of A^k is computed by

$$A^k V = A^{k-1}(AV) = A^{k-1}(\lambda V) = \lambda A^{k-1}V = \lambda A^{k-2}(AV) = \lambda^2 A^{k-2}V = \dots = \lambda^k V$$

This result indicates that λ^k is the eigenvalue of A^k with the same eigenvector of the eigenvalue λ of A .

Now, let's compute the eigenvalues of H

$$HV_r = \left(\sum_{k=0}^{n-1} h_k J^k \right) V_r = \sum_{k=0}^{n-1} h_k (J^k V_r)$$

Using the results above, we get

$$\begin{aligned} HV_r &= \left(\sum_{k=0}^{n-1} h_k J^k \right) V_r = \sum_{k=0}^{n-1} h_k (J^k V_r) = \sum_{k=0}^{n-1} h_k (\lambda_r^k V_r) = \left(\sum_{k=0}^{n-1} h_k \lambda_r^k \right) V_r = P(\lambda_r) V_r \\ &= \lambda_{H_r} V_r \end{aligned}$$

Therefore, the eigenvalue λ_{H_r} of H with V_r as corresponding eigenvector is given by

$$\begin{aligned} \lambda_{H_r} &= P(\lambda_r) = P\left(e^{i\frac{2\pi(r-1)}{n}}\right) = h_1 e^{i\frac{2\pi r}{n}} + h_2 e^{i\frac{2\pi r \cdot 2}{n}} + \dots + h_{n-1} e^{i\frac{2\pi r(n-1)}{n}} + h_0 \quad \text{for } r \\ &= 0, 1, 2, \dots, n-1 \end{aligned}$$

H is diagonalizable by the same unitary matrix $U = \frac{1}{\sqrt{n}} \begin{pmatrix} | & | & \dots & | \\ V_1 & V_2 & \dots & V_n \\ | & | & \dots & | \end{pmatrix}$ with the eigenvalues computed above, meaning it is diagonalizable in a unitary basis.

- f. We want to show that the diagonalization basis matrix B can be chosen as the DFT^* matrix:

We saw earlier that H is diagonalizable in unitary basis using the basis matrix U .

Since U is symmetric and each one of its elements is given by $W^{*kr} =$

$e^{-\frac{i2\pi}{n}kr}$ for $k, r = 0, \dots, n-1$, therefore its conjugate is the conjugate of the Discrete Fourier Transform matrix, given that H is diagonalizable using U and since $U = DFT$, we get

$$\begin{aligned} H &= B \Lambda B^* \\ \underbrace{H}_{H \text{ is real}} &\underbrace{=}_{\underbrace{DFT \text{ is symmetric}}} \underbrace{\overline{DFT^*} \Lambda \overline{DFT}}_{\underbrace{DFT \text{ is symmetric}}} \underbrace{=}_{\underbrace{DFT \text{ is symmetric}}} DFT \Lambda DFT^* \end{aligned}$$

In short, both the DFT and its complex conjugate can be chosen as the diagonalization basis matrix and it's up to the choice of the eigenvalues (we can choose them to be $\lambda_k = e^{-i\frac{2\pi k}{n}}$ for $k = 0, 1, 2, \dots, n-1$ or $\lambda_k = e^{i\frac{2\pi k}{n}}$ for $k = 0, 1, 2, \dots, n-1$).

- g. Let B be the diagonalization basis matrix and let's denote $\sqrt{n}B = U$, then we get $U^*HU = \Lambda$

By applying matrix transpose operation on both sides and considering the symmetry of U and U^* and the fact that Λ is diagonal matrix, we get

$$(U^*HU)^T = \Lambda^T \\ \Rightarrow UH^TU^* = \Lambda$$

After multiplying both side by U from the right, we get

$$UH^T = \Lambda U \\ U \begin{pmatrix} h_0 & h_1 & h_2 & \cdots & h_{n-1} \\ h_{n-1} & h_0 & h_1 & \cdots & h_{n-2} \\ h_{n-2} & h_{n-1} & h_0 & \cdots & h_{n-3} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ h_1 & h_2 & h_3 & \cdots & h_0 \end{pmatrix} \\ = \begin{pmatrix} \lambda_0 & 0 & 0 & \cdots & 0 \\ 0 & \lambda_1 & 0 & \cdots & 0 \\ 0 & 0 & \lambda_3 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & \lambda_{n-1} \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & \lambda_1 & \lambda_2 & \cdots & \lambda_n \\ 1 & \lambda_1^2 & \lambda_2^2 & \cdots & \lambda_n^2 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \lambda_1^{n-1} & \lambda_2^{n-1} & \cdots & \lambda_n^{n-1} \end{pmatrix}$$

Since both sides are equal, we get

$$\sqrt{n}B \begin{pmatrix} h_0 \\ h_{n-1} \\ h_{n-2} \\ \vdots \\ h_1 \end{pmatrix} = \begin{pmatrix} \lambda_0 & 0 & 0 & \cdots & 0 \\ 0 & \lambda_1 & 0 & \cdots & 0 \\ 0 & 0 & \lambda_3 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & \lambda_{n-1} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 1 \\ \vdots \\ 1 \end{pmatrix} = \begin{pmatrix} \lambda_0 \\ \lambda_1 \\ \lambda_2 \\ \vdots \\ \lambda_{n-1} \end{pmatrix}$$

Therefore, we get

$$\begin{pmatrix} \lambda_0 \\ \lambda_1 \\ \lambda_2 \\ \vdots \\ \lambda_{n-1} \end{pmatrix} = \sqrt{n}B \begin{pmatrix} h_0 \\ h_{n-1} \\ h_{n-2} \\ \vdots \\ h_1 \end{pmatrix}$$

- h. We want to show that given the two circulant matrices $H_1, H_2 \in \mathbb{R}^{n \times n}$, H_1 and H_2 commute, meaning $H_1H_2 = H_2H_1$:

Let B be the diagonalization basis matrix and let's denote $\sqrt{n}B = U$, hence $U^*HU = \Lambda$.

Let's assume the following

$$H_1 = U\Lambda_1U^* \\ H_2 = U\Lambda_2U^*$$

Then we get

$$H_1H_2 = U\Lambda_1U^*U\Lambda_2U^*$$

Since U is unitary matrix, we get $U^*U = I$, therefore

$$H_1H_2 = U\Lambda_1U^*U\Lambda_2U^* = U\Lambda_1\Lambda_2U^*$$

Considering that Λ_1 and Λ_2 are diagonal matrices meaning they commute, we get

$$H_1H_2 = U\Lambda_1\Lambda_2U^* = U\Lambda_2\Lambda_1U^*$$

By using the fact $U^*U = I$ again we get

$$H_1 H_2 = U \Lambda_2 \Lambda_1 U^* = U \Lambda_2 U^* U \Lambda_1 W^* = H_2 H_1$$

Therefore, H_1 and H_2 commute.

$H_1 H_2$ is a circulant matrix, we show this using the results of section (d).

We saw that H_1 and H_2 are given by polynomial expression of the matrix J , meaning

$$H_1 = P_1(J)$$

$$H_2 = P_2(J)$$

Hence,

$$H_1 H_2 = P_1(J) P_2(J) = \sum_{k=0}^{n-1} h_{1k} J^k \sum_{r=0}^{n-1} h_{2r} J^r = \sum_{k=0}^{n-1} \sum_{r=0}^{n-1} h_{1k} h_{2r} J^{k+r}$$

As we saw earlier, J^{k+r} is a circulant matrix, thus $h_{1k} h_{2r} J^{k+r}$ is a circulant matrix too, and since the outcome of adding two circulant matrices is also a circulant matrix we get that $H_1 H_2$ is a circulant matrix too.

i. We want to compute DFT^k for $k \in \mathbb{N}$:

Let's first check the behavior of DFT^k for $k = 2$, we get

$$DFT^2 = \frac{1}{\sqrt{n}} \begin{pmatrix} w^{*0 \cdot 0} & w^{*1 \cdot 0} & w^{*2 \cdot 0} & \dots & w^{*(n-1) \cdot 0} \\ w^{*0 \cdot 1} & w^{*1 \cdot 1} & w^{*2 \cdot 1} & \dots & w^{*(n-1) \cdot 1} \\ w^{*0 \cdot 2} & w^{*1 \cdot 2} & w^{*2 \cdot 2} & \dots & w^{*(n-1) \cdot 2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ w^{*0 \cdot (n-1)} & w^{*1 \cdot (n-1)} & w^{*2 \cdot (n-1)} & \dots & w^{*(n-1) \cdot (n-1)} \end{pmatrix}$$

$$\cdot \frac{1}{\sqrt{n}} \begin{pmatrix} w^{*0 \cdot 0} & w^{*1 \cdot 0} & w^{*2 \cdot 0} & \dots & w^{*(n-1) \cdot 0} \\ w^{*0 \cdot 1} & w^{*1 \cdot 1} & w^{*2 \cdot 1} & \dots & w^{*(n-1) \cdot 1} \\ w^{*0 \cdot 2} & w^{*1 \cdot 2} & w^{*2 \cdot 2} & \dots & w^{*(n-1) \cdot 2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ w^{*0 \cdot (n-1)} & w^{*1 \cdot (n-1)} & w^{*2 \cdot (n-1)} & \dots & w^{*(n-1) \cdot (n-1)} \end{pmatrix}$$

Hence,

$$DFT_{kr}^2 = \frac{1}{n} \sum_{l=0}^{n-1} w^{*lk} w^{*rl} = \frac{1}{n} \sum_{l=0}^{n-1} e^{-\frac{i2\pi lk}{n}} e^{-\frac{i2\pi lr}{n}} = \frac{1}{n} \sum_{l=0}^{n-1} e^{-\frac{i2\pi l(k+r)}{n}}$$

$$= \begin{cases} 1 & (k+r) \bmod n = 0 \\ 0 & \text{else} \end{cases}$$

And we get

$$DFT^2 = \begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & \vdots & \ddots & 1 \\ 0 & \vdots & 0 & \ddots & 0 \\ \vdots & 0 & 1 & \ddots & 0 \\ 0 & 1 & 0 & \dots & 0 \end{pmatrix}$$

We notice that this is the vertically flipped form of J , let's denote it by FJ .

Now for $k = 4$ we get

$$\begin{aligned}
DFT^4 &= DFT^2 + DFT^2 = \begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & \vdots & \ddots & 1 \\ 0 & \vdots & 0 & \ddots & 0 \\ \vdots & 0 & 1 & \ddots & 0 \\ 0 & 1 & 0 & \dots & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & \vdots & \ddots & 1 \\ 0 & \vdots & 0 & \ddots & 0 \\ \vdots & 0 & 1 & \ddots & 0 \\ 0 & 1 & 0 & \dots & 0 \end{pmatrix} \\
&= \begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix} = I_{n \times n}
\end{aligned}$$

To calculate DFT^3 we rely on the fact that $DFT \cdot DFT^* = I$, thus for $k = 3$ we get

$$DFT^3 = DFT^3 \cdot I = DFT^3 \cdot DFT \cdot DFT^* = DFT^4 \cdot DFT^* = I \cdot DFT^*$$

Relying on the results we got above, we can conclude that for $k \in \mathbb{N}$ DFT^k is given by:

$$DFT^k = \begin{cases} I_{n \times n} & k \bmod 4 = 0 \\ DFT & k \bmod 4 = 1 \\ FJ & k \bmod 4 = 2 \\ DFT^* & k \bmod 4 = 3 \end{cases}$$

- j. We want to prove that a convolution of n -dimensional signals can be computed by point-wise multiplication of the signals in the Fourier domain, up to a normalization.
The convolution of x and y can be computed using a circulant matrix X built from x (the same as H by with the elements of x) then multiplying by y as follows

$$z = x \otimes y = Xy$$

We get

$$(DFT)z = (DFT)Xy$$

Considering $(DFT^*)(DFT) = I$ we get

$$(DFT)z = (DFT)Xy = (DFT)X(DFT^*)(DFT)y$$

Considering that $(DFT)X(DFT^*)$ is a diagonal matrix and hence symmetric, we get

$$(DFT)z = (DFT)Xy = (DFT)X(DFT^*)(DFT)y = (DFT^*)X(DFT) \cdot (DFT)y$$

Relying on the results of (g), we get

$$\sqrt{n}(DFT)y = \begin{pmatrix} \lambda_0^y \\ \lambda_1^y \\ \lambda_2^y \\ \vdots \\ \lambda_{n-1}^y \end{pmatrix}$$

$$(DFT)y = \frac{1}{\sqrt{n}} \begin{pmatrix} \lambda_0^y \\ \lambda_1^y \\ \lambda_2^y \\ \vdots \\ \lambda_{n-1}^y \end{pmatrix}$$

Therefore, we get

$$(DFT)z = (DFT^*)X(DFT)(DFT)y = \begin{pmatrix} \lambda_0^x & 0 & 0 & \cdots & 0 \\ 0 & \lambda_1^x & 0 & \cdots & 0 \\ 0 & 0 & \lambda_2^x & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & \lambda_{n-1}^x \end{pmatrix} \begin{pmatrix} \lambda_0^y \\ \lambda_1^y \\ \lambda_2^y \\ \vdots \\ \lambda_{n-1}^y \end{pmatrix} =$$

$$\sqrt{n} \begin{pmatrix} \lambda_0^x \\ \lambda_1^x \\ \lambda_2^x \\ \vdots \\ \lambda_{n-1}^x \end{pmatrix} \odot \begin{pmatrix} \lambda_0^y \\ \lambda_1^y \\ \lambda_2^y \\ \vdots \\ \lambda_{n-1}^y \end{pmatrix} = \sqrt{n}(DFT)x \odot (DFT)y$$

2. Fourier Transform

- a. $f(t)$ and $g(t)$ are two given functions, with convolution denoted by $h(t)$ and given by

$$h(t) = f(t) * g(t)$$

We want to find $f(t-1) * g(t+1)$ in terms of $h(t)$

We start by denoting $f(t-1) = \tilde{f}(t)$ and $g(t+1) = \tilde{g}(t)$, then we get

$$\begin{aligned} f(t-1) * g(t+1) &= \tilde{f}(t) * \tilde{g}(t) = \int_{-\infty}^{\infty} \tilde{f}(\tilde{\tau}) \tilde{g}(t - \tilde{\tau}) d\tilde{\tau} \\ &= \int_{-\infty}^{\infty} f(\tilde{\tau} - 1) g(t - \tilde{\tau} + 1) d\tilde{\tau} = \int_{-\infty}^{\infty} f(\tilde{\tau} - 1) g(t - (\tilde{\tau} - 1)) d\tilde{\tau} \end{aligned}$$

By changing the integral variable $\tilde{\tau}$ by denoting $\tilde{\tau} - 1 = \tau$ and hence $d\tilde{\tau} = d\tau$, we get

$$\begin{aligned} f(t-1) * g(t+1) &= \int_{-\infty}^{\infty} f(\tilde{\tau} - 1) g(t - (\tilde{\tau} - 1)) d\tilde{\tau} = \int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau \\ &= f(t) * g(t) = h(t) \end{aligned}$$

- b. $f(t)$ and $g(t)$ are two given functions, we want to show that the following condition holds

$$\int_{-\infty}^{\infty} f(t) g(-t) dt = \int_{-\infty}^{\infty} \mathcal{F}(u) \mathcal{G}(u) du$$

Where $\mathcal{F}(u)$ and $\mathcal{G}(u)$ are the Fourier transform of $f(t)$ and $g(t)$ respectively.

We get

$$\int_{-\infty}^{\infty} \mathcal{F}(u) \mathcal{G}(u) du = \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} f(\tau) e^{-i2\pi u \tau} d\tau \int_{-\infty}^{\infty} g(\eta) e^{-i2\pi u \eta} d\eta \right) du$$

$$\begin{aligned}
&\stackrel{\text{Fubini's theorem}}{=} \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\tau) e^{-i2\pi u \tau} g(\eta) e^{-i2\pi u \eta} d\tau d\eta \right) du \\
&= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\tau) g(\eta) e^{-i2\pi u (\tau + \eta)} d\tau d\eta \right) du
\end{aligned}$$

By changing the variables and denoting $\eta + \tau = t$, hence $\eta = t - \tau$ and $d\eta = dt$. We get

$$\begin{aligned}
\int_{-\infty}^{\infty} \mathcal{F}(u) \mathcal{G}(u) du &= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\tau) g(\eta) e^{-i2\pi u (\tau + \eta)} d\tau d\eta \right) du \\
&= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\tau) g(t - \tau) e^{-i2\pi u t} d\tau dt \right) du \\
&= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} e^{-i2\pi u t} dt \int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau \right) du
\end{aligned}$$

$\int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau$ is the definition of convolution of the two functions $f(t)$ and $g(t)$, meaning

$$h(t) = f(t) * g(t) = \int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau$$

Now we get

$$\begin{aligned}
\int_{-\infty}^{\infty} \mathcal{F}(u) \mathcal{G}(u) du &= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} e^{-i2\pi u t} dt \int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau \right) du \\
&= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} h(t) e^{-i2\pi u t} dt \right) du
\end{aligned}$$

$\int_{-\infty}^{\infty} h(t) e^{-i2\pi u t} dt = H(u)$ is the Fourier transform of $h(t)$, therefore we get

$$\int_{-\infty}^{\infty} \mathcal{F}(u) \mathcal{G}(u) du = \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} h(t) e^{-i2\pi u t} dt \right) du = \int_{-\infty}^{\infty} H(u) du = \int_{-\infty}^{\infty} H(u) e^{2\pi i u \cdot 0} du$$

$\int_{-\infty}^{\infty} H(u) e^{2\pi i u \cdot 0} du$ is the projection of $h(t = 0)$ to the Fourier family, as seen in the lecture, or the inverse Fourier transform, also $h(t) = f(t) * g(t) = \int_{-\infty}^{\infty} f(\tau) g(t - \tau) d\tau$

Thus, $h(0) = \int_{-\infty}^{\infty} f(\tau) g(-\tau) d\tau = \int_{-\infty}^{\infty} f(t) g(-t) dt$

$$\int_{-\infty}^{\infty} \mathcal{F}(u) \mathcal{G}(u) du = \int_{-\infty}^{\infty} H(u) e^{2\pi i u \cdot 0} du = h(0) = \int_{-\infty}^{\infty} f(t) g(-t) dt$$

We got $\int_{-\infty}^{\infty} \mathcal{F}(u) \mathcal{G}(u) du = \int_{-\infty}^{\infty} f(t) g(-t) dt$ as required.

Question 3

$$\begin{aligned}
 \text{a. } \phi^F &= \sqrt{2N} \begin{bmatrix} \sum_{k=0}^{2N-1} W^{*k \cdot 0} \cdot \phi(k) \\ \sum_{k=0}^{2N-1} W^{*k \cdot 1} \cdot \phi(k) \\ \sum_{k=0}^{2N-1} W^{*k \cdot 2} \cdot \phi(k) \\ \vdots \\ \sum_{k=0}^{2N-1} W^{*k \cdot (2N-1)} \cdot \phi(k) \end{bmatrix} = \\
 &\sqrt{2N} \begin{bmatrix} W^{*0 \cdot 0} + \frac{W^{*0}}{2} + \frac{W^{*(2N-1) \cdot 0}}{2} \\ W^{*0 \cdot 1} + \frac{W^{*1}}{2} + \frac{W^{*(2N-1) \cdot 1}}{2} \\ W^{*0 \cdot 2} + \frac{W^{*2}}{2} + \frac{W^{*(2N-1) \cdot 2}}{2} \\ \vdots \\ W^{*0 \cdot (2N-1)} + \frac{W^{*(2N-1)}}{2} + \frac{W^{*(2N-1) \cdot (2N-1)}}{2} \end{bmatrix} = \\
 &\sqrt{2N} \begin{bmatrix} 1 + \frac{W^{*0}}{2} + \frac{W^0}{2} \\ 1 + \frac{W^{*1}}{2} + \frac{W^1}{2} \\ 1 + \frac{W^{*2}}{2} + \frac{W^2}{2} \\ \vdots \\ 1 + \frac{W^{*(2N-1)}}{2} + \frac{W^{(2N-1)}}{2} \end{bmatrix}
 \end{aligned}$$

b. For $k \in [0, \dots, 2N-1]$:

$$\begin{aligned}
 \gamma^F(k) &= \frac{1}{\sqrt{2N}} \cdot \sum_{n=0}^{2N-1} W_{2N}^{*n \cdot k} \cdot \gamma_n = \\
 &\frac{1}{\sqrt{2N}} \cdot \sum_{n=0}^{N-1} W_{2N}^{*(2n) \cdot k} \cdot \psi_n = \\
 &\frac{1}{\sqrt{2N}} \cdot \sum_{n=0}^{N-1} W_N^{*n \cdot k} \cdot \psi_n = \\
 &\frac{1}{\sqrt{2}} \cdot \psi^F(k \pmod{N})
 \end{aligned}$$

So:

$$\gamma^F = \frac{1}{\sqrt{2}} \cdot [\psi_0^F, \psi_1^F, \dots, \psi_{N-1}^F, \psi_0^F, \psi_1^F, \dots, \psi_{N-1}^F]^T$$

c. For $k \in [0, \dots, 2N - 1]$:

$$(\gamma * \phi)(k) =$$

$$\sum_{m=0}^{2N-1} \phi(m) \cdot \gamma(k - m) =$$

$$\phi(0) \cdot \gamma(k - 0) + \phi(1) \cdot \gamma(k - 1) + \phi(2N - 1) \cdot \gamma(k - 2N + 1) =$$

$$\begin{aligned} & \gamma(k) + \frac{\gamma(k - 1)}{2} + \frac{\gamma(k - 2N + 1)}{2} = \\ & \gamma(k) + \frac{\gamma(k - 1)}{2} + \frac{\gamma(k + 1)}{2} = \begin{cases} \psi\left(\frac{k}{2}\right), & k \text{ is even} \\ \frac{\psi\left(\frac{k-1}{2}\right)}{2} + \frac{\psi\left(\frac{k+1}{2}\right)}{2}, & k \text{ is odd} \end{cases} \end{aligned}$$

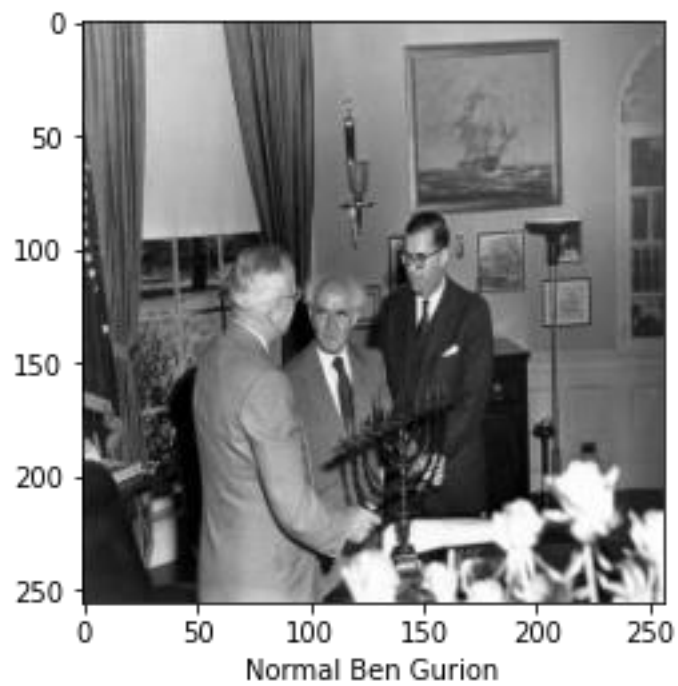
d. For $k \in [0, \dots, 2N - 1]$:

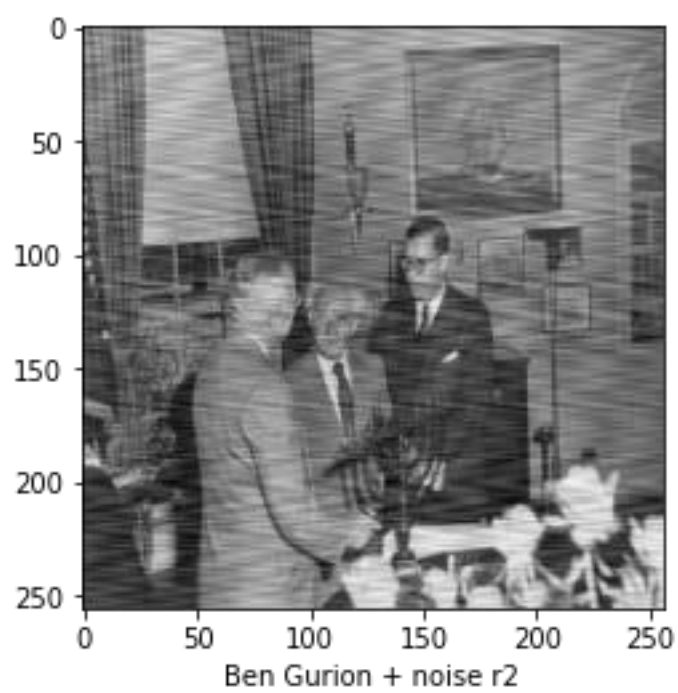
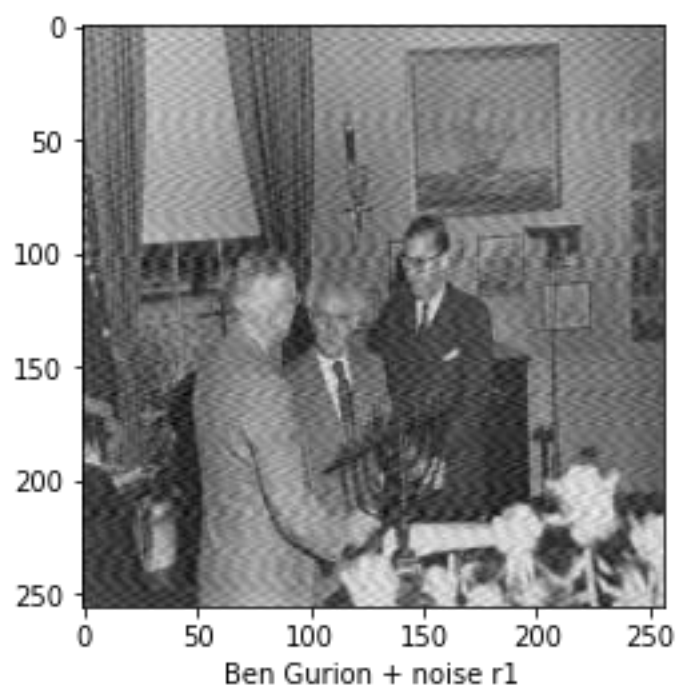
$$\begin{aligned} h^F(k) &= \frac{1}{\sqrt{2N}} \cdot \sum_{n=0}^{2N-1} W_{2N}^{*n \cdot k} \cdot h_n = \\ & \frac{1}{\sqrt{2N}} \cdot \sum_{n=0}^{N-1} W_{2N}^{*(2n) \cdot k} \cdot \psi_n + \frac{1}{\sqrt{2N}} \cdot \sum_{n=0}^{N-1} W_{2N}^{*(2n+1) \cdot k} \cdot \frac{\psi_n + \psi_{n+1}}{2} = \\ & \frac{1}{\sqrt{2}} \cdot \psi^F(k \pmod{N}) + \frac{1}{\sqrt{8N}} \cdot \left(\sum_{n=0}^{N-1} W_{2N}^{*(2n+1) \cdot k} \cdot \psi_n + \sum_{n=0}^{N-1} W_{2N}^{*(2n+1) \cdot k} \cdot \psi_{n+1} \right) = \\ & \frac{1}{\sqrt{2}} \cdot \psi^F(k \pmod{N}) + \frac{1}{\sqrt{8N}} \cdot \left(W_{2N}^{*k} \sum_{n=0}^{N-1} W_{2N}^{*(2n) \cdot k} \cdot \psi_n + \frac{1}{W_{2N}^{*k}} \sum_{n=0}^{N-1} W_{2N}^{*(2n) \cdot k} \cdot \psi_n \right) = \\ & \frac{1}{\sqrt{2}} \cdot \psi^F(k \pmod{N}) + \frac{W_{2N}^{*k} + \frac{1}{W_{2N}^{*k}}}{\sqrt{8}} \cdot \left(\frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} W_{2N}^{*(2n) \cdot k} \cdot \psi_n \right) = \\ & \frac{W_{2N}^{*k} + W_{2N}^k + 2}{\sqrt{8}} \cdot \psi^F(k \pmod{N}) \end{aligned}$$

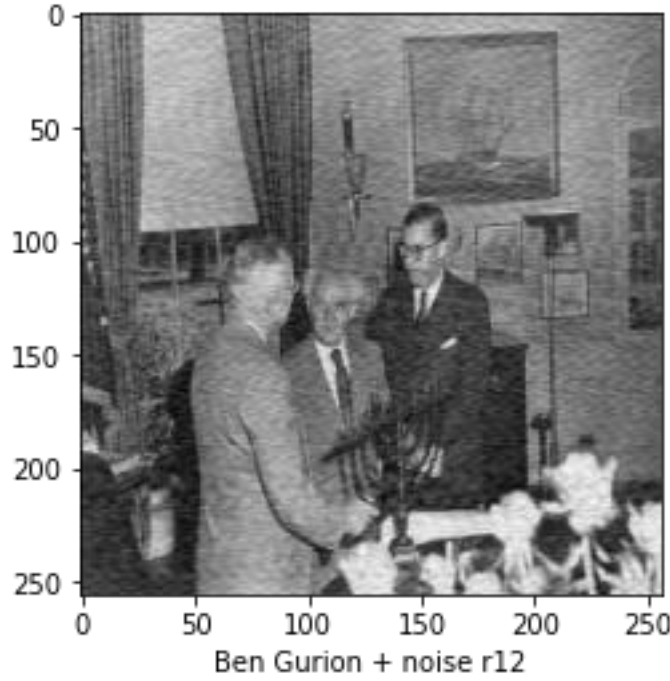


Implementation

a.







- b. We use the index t instead of i , so as not to confuse with the imaginary number i .
We will define: $u = f \cdot n$.

$$\begin{aligned}
 \phi_{I_{t,j}}^F &= \phi_{I_t}^F(j) = \frac{1}{\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot I_t^{\text{noisy}}(k) = \\
 &= \frac{1}{\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot (I_t(k) + A_t \cdot \cos(2\pi f k + \varphi_t)) = \\
 &= \frac{1}{\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot I_t(k) + \frac{1}{2\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot A_t \cdot \left(e^{\frac{i2\pi}{n}uk + i\varphi_t} + e^{-\frac{i2\pi}{n}uk - i\varphi_t} \right) = \\
 &= \frac{1}{\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot I_t(k) + \frac{1}{2\sqrt{n}} \sum_{k=0}^{n-1} A_t \cdot W^{*j \cdot k} \cdot (W^{u \cdot k} \cdot e^{i\varphi_t} + W^{-u \cdot k} \cdot e^{-i\varphi_t}) = \\
 &= \frac{1}{\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot I_t(k) + \frac{A_t e^{i\varphi_t}}{2\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot W^{u \cdot k} + \frac{A_t e^{-i\varphi_t}}{2\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot W^{-u \cdot k}
 \end{aligned}$$

For $\sum_{k=0}^{n-1} W^{*j \cdot k} \cdot W^{u \cdot k}$:

If $u = j$:

$$\sum_{k=0}^{n-1} W^{*j \cdot k} \cdot W^{u \cdot k} = \sum_{k=0}^{n-1} W^0 = n$$

Else:

$$\sum_{k=0}^{n-1} W^{*j \cdot k} \cdot W^{u \cdot k} = \sum_{k=0}^{n-1} (W^{*j-u})^k = \frac{(W^{*j-u})^n - 1}{(W^{*j-u}) - 1} = \frac{(W^{*n})^{j-u} - 1}{(W^{*j-u}) - 1} = \frac{(1)^{j-u} - 1}{(W^{*j-u}) - 1} = 0$$

For $\sum_{k=0}^{n-1} W^{*j \cdot k} \cdot W^{-u \cdot k}$:

If $u = -j$:

$$\sum_{k=0}^{n-1} W^{*j \cdot k} \cdot W^{-u \cdot k} = \sum_{k=0}^{n-1} W^0 = n$$

Else:

$$\sum_{k=0}^{n-1} W^{*j \cdot k} \cdot W^{-u \cdot k} = \sum_{k=0}^{n-1} (W^{*-j-u})^k = \frac{(W^{*-j-u})^n - 1}{(W^{*-j-u}) - 1} = \frac{(1)^{-j-u} - 1}{(W^{*-j-u}) - 1} = 0$$

So:

$$\phi_{I_{t,j}}^F =$$

$$\frac{1}{\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot I_t(k) + \frac{A_t e^{i\varphi_t \sqrt{n}}}{2} \cdot \delta_{j,f \cdot n} + \frac{A_t e^{-i\varphi_t \sqrt{n}}}{2} \cdot \delta_{j,-f \cdot n}$$

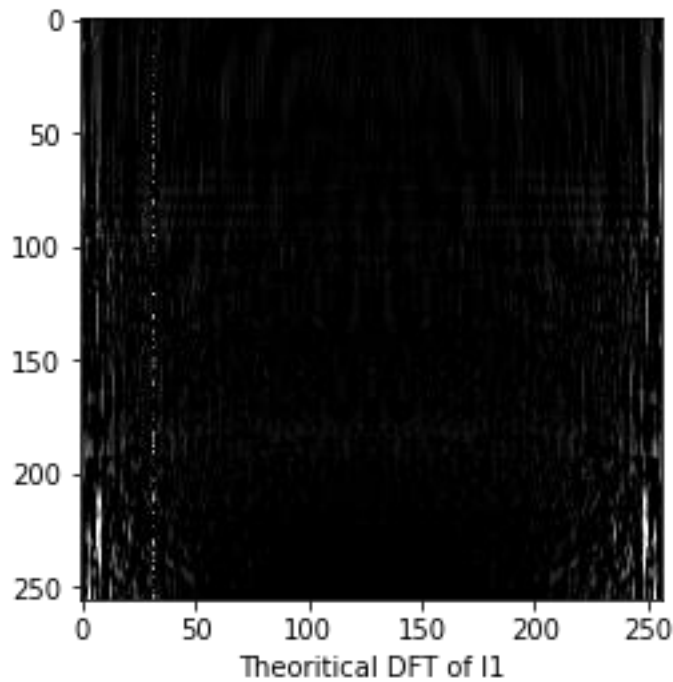
c. The weight of noise 1: w_1 ; The weight of noise 2: w_2

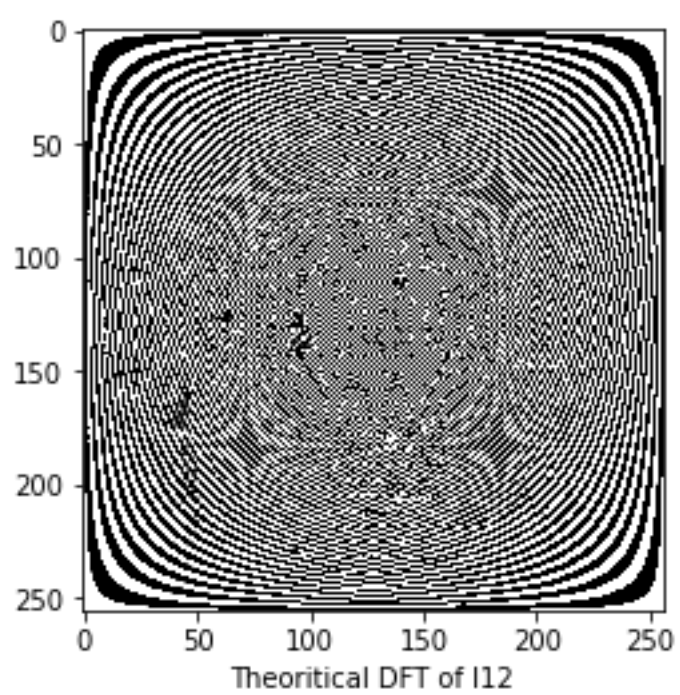
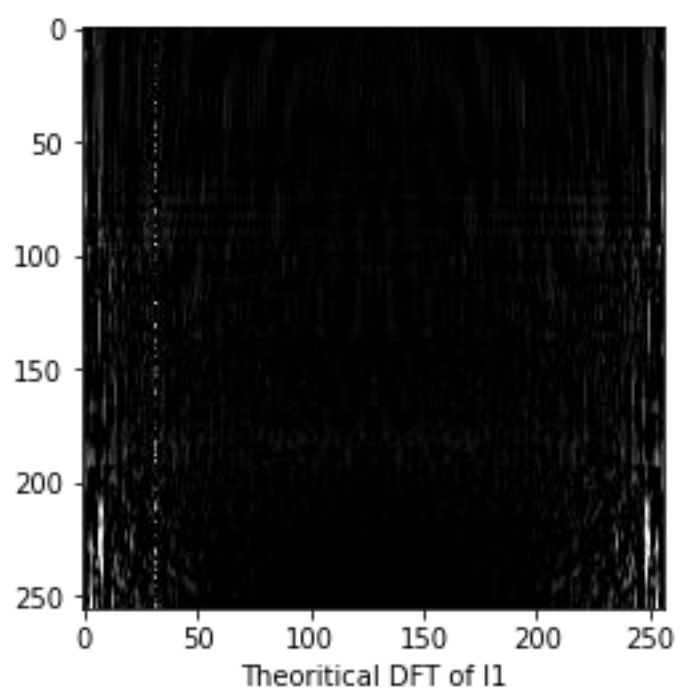
$$\phi_{I_{t,j}}^F =$$

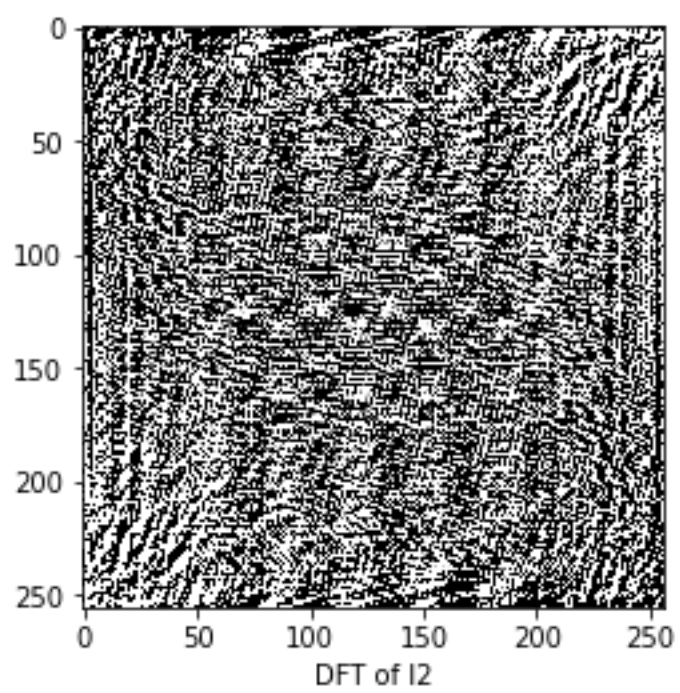
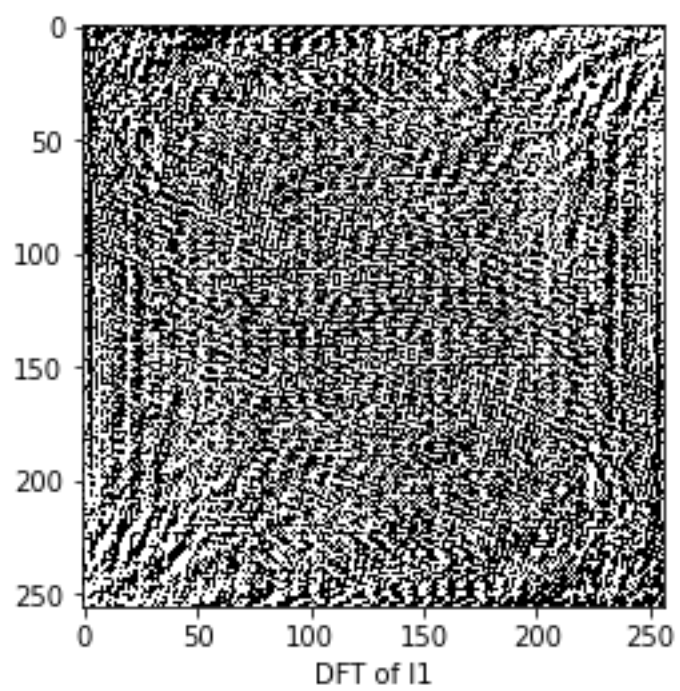
$$\frac{1}{\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot \left(I_t(k) + \frac{w_1 \cdot A_t^{(1)} \cdot \cos(2\pi f^{(1)} k + \varphi_t^{(1)})}{w_1 + w_2} + \frac{w_2 \cdot A_t^{(2)} \cdot \cos(2\pi f^{(2)} k + \varphi_t^{(2)})}{w_1 + w_2} \right) =$$

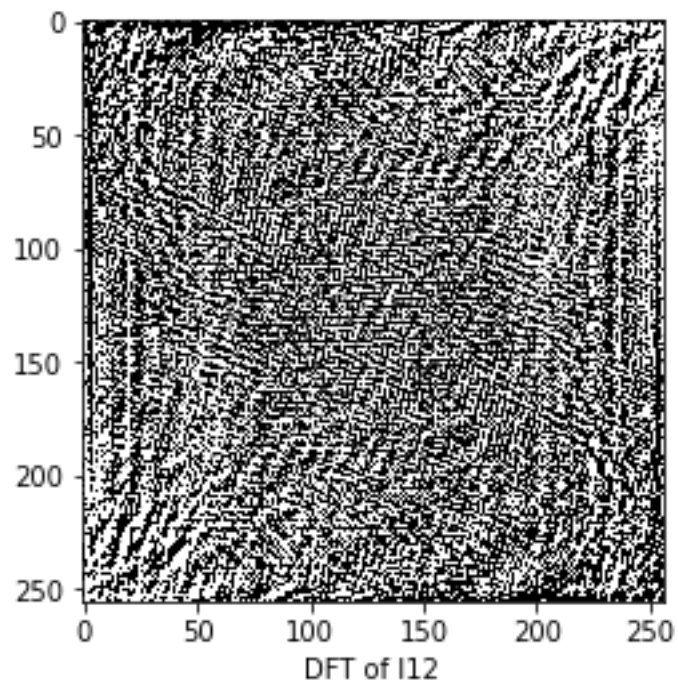
$$\frac{1}{\sqrt{n}} \sum_{k=0}^{n-1} W^{*j \cdot k} \cdot I_t(k) + \frac{w_1 A_t^{(1)} e^{i\varphi_t^{(1)} \sqrt{n}}}{2(w_1 + w_2)} \cdot \delta_{j,f^{(1)} \cdot n} + \frac{w_2 A_t^{(2)} e^{i\varphi_t^{(2)} \sqrt{n}}}{2(w_1 + w_2)} \cdot \delta_{j,f^{(2)} \cdot n} + \frac{w_1 A_t^{(1)} e^{-i\varphi_t^{(1)} \sqrt{n}}}{2(w_1 + w_2)} \cdot \delta_{j,-f^{(1)} \cdot n} + \frac{w_2 A_t^{(2)} e^{-i\varphi_t^{(2)} \sqrt{n}}}{2(w_1 + w_2)} \cdot \delta_{j,-f^{(2)} \cdot n}$$

d.









e.

MSE of I1 is: 2.7109850297764546
MSE of I2 is: 2.7002239091778053
MSE of I12 is: 359.91252877302543

