ASSIGNMENT - 3

Write the proof and valious properties related to 2D DFT. DFT: $F(u,v) = \sum_{\chi=0}^{M-1} \sum_{g=0}^{N-1} f(\chi_1 g) e^{-\frac{i}{2}2\pi \left(\frac{u\chi}{M} + \frac{u\chi}{N}\right)}$

Linearity: If
$$f_1(x_1) \hookrightarrow F_1(u_10)$$

 $b_2(x_1) \hookrightarrow F_2(u_10)$
then, $a f_1(x_1) + b f_2(x_1) \hookrightarrow a F_1(u_10) + b F_2(u_10)$

$$P200f: DFT \left(a f_{1}(x, b) + b f_{2}(x, b) \right) = \sum_{x} \sum_{y} (a f_{1}(x, b) + b f_{2}(x, b)) e^{-j2\pi \left(\frac{ux}{m} + \frac{uy}{n}\right)}$$

$$= a \sum_{x} \sum_{y} f_{1}(x, b) e^{-j2\pi \left(\frac{ux}{m} + \frac{uy}{n}\right)} + b \sum_{x} \sum_{y} f_{2}(x, b) e^{-j2\pi \left(\frac{ux}{m} + \frac{uy}{n}\right)}$$

Prof.

DFT
$$\left\{ b_1(x)b_2(y) \right\} = \sum_{x} \sum_{y} b_1(x)b_2(y)e^{-\frac{1}{2}\pi vy}$$

$$= \sum_{x} b_1(x)e^{-\frac{1}{2}\frac{2\pi vx}{M}} \sum_{y} b_2(y)e^{-\frac{1}{2}\frac{2\pi vy}{N}} = F_1(u)F_2(v)$$

$$= \sum_{x} b_1(x)e^{-\frac{1}{2}\frac{2\pi vx}{M}} \sum_{y} b_2(y)e^{-\frac{1}{2}\frac{2\pi vy}{N}} = F_1(u)F_2(v)$$

Periodicity: (M,N): F(u,v) = F(u+N,v) = F(u,v+N) = F(u+M,v+N)

Spatial domain:
$$\delta(x-x_0, y-y_0) \longleftrightarrow F(u,v) e^{-j2\pi \left(\frac{ux_0+vy_0}{M} + \frac{vy_0}{N}\right)}$$

Proof: $DFT \left\{ \delta(x-x_0, y-y_0) \right\} = \sum_{x} \sum_{y} \delta(x-x_0, y-y_0) e^{-j2\pi \left(\frac{ux_0}{M} + \frac{vy_0}{N}\right)}$
 $= \sum_{x} \sum_{y} \delta(x,y) e^{-j2\pi \left(\frac{ux_0}{M} + \frac{vy_0}{N}\right)} e^{-j2\pi \left(\frac{ux_0}{M} + \frac{vy_0}{N}\right)}$
 $= F(u,v) e^{-j2\pi \left(\frac{ux_0}{M} + \frac{vy_0}{N}\right)}$

Freq. domain:
$$f(x_1) = \frac{\partial \pi(u_0 x + v_0 y)}{m} = \sum_{x} \frac{\partial \pi(u_x + v_0 y)}{m} = \sum_{x} \frac{\partial$$

S) Symmetry: (i)
$$f(x,b)$$
 led: $F^{-1}(u,0) = F(-u,-v)$

Prob: $F^{+1}(u,v) = \sum_{x \in S} \int_{x}^{1} (x,b) e^{j2\pi (\frac{ux}{n} + \frac{vy}{n})}$
 $= \sum_{x \in S} \int_{x}^{1} (x,b) e^{-j2\pi (-\frac{ux}{n} - \frac{vy}{n})} = F(-u,-v)$

(ii) $f(x,b)$ inaginal: $F^{+1}(u,v) = -F(u,v)$
 $f(x,b) = \sum_{x \in S}^{1} \int_{x}^{1} (x,b) e^{j2\pi (\frac{ux}{n} + \frac{vy}{n})}$

$$\frac{1}{2} = -\sum_{x} \frac{1}{3} \left(\frac{1}{3} \right) e^{-\frac{1}{3} 2\pi \left(\frac{1}{3} + \frac{1}{3} \right)} = -F(u, 0)$$

Consolution (Civales):
$$f_{1}(x_{1}) \otimes f_{2}(x_{1}) \Leftrightarrow F_{1}(u_{1}) \otimes F_{2}(u_{1})$$

Prof: $f_{1}(x_{1}) \otimes f_{2}(x_{1}) = \sum_{x} \int_{y} (f_{1}(x_{1}) \otimes f_{2}(x_{1})) e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})}$

$$= \sum_{x} \sum_{x} \sum_{y} f_{1}(x_{1}) \int_{y} (x_{1}) \int_{y} (x_{1}) \int_{y} (x_{1}) \int_{y} e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})}$$

$$= \sum_{x} \sum_{y} \int_{y} f_{1}(x_{1}) \int_{z} \int_{z} f_{2}(x_{1}) \int_{y} (x_{1}) \int_{y} e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})} e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})}$$

$$= \sum_{x} \int_{y} f_{1}(x_{1}) \int_{z} \int_{z} f_{2}(x_{1}) \int_{y} e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})} e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})}$$

$$= \sum_{x} \int_{y} f_{1}(x_{1}) \int_{z} \int_{z} f_{2}(x_{1}) \int_{z} e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})} e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})} e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})}$$

$$= \sum_{x} \int_{z} f_{1}(x_{1}) \int_{z} f_{2}(x_{1}) \int_{z} e^{-j\lambda \pi (\frac{u_{1}}{H} + \frac{u_{2}}{H})} e^{-j\lambda \pi (\frac{u_{1}}{$$

Average value:
$$\frac{1}{N^2}F(0,0) = \frac{1}{6}(x,y)$$

$$F(0,0) = \sum_{x=0}^{N^2} \sum_{y=0}^{N^2} f(x,y) e^{-\frac{1}{2}x\pi} \left(\frac{0x}{n} + \frac{0y}{N}\right) = \sum_{x=0}^{N^2} \sum_{y=0}^{N^2} f(x,y)$$

$$= \frac{1}{N^2} \sum_{x=0}^{N^2} \frac{1}{N^2} f(x,y) = \frac{1}{N^2} F(0,0)$$

$$= \frac{1}{N^2} \sum_{x=0}^{N^2} \frac{1}{N^2} f(x,y) = \frac{1}{N^2} F(0,0)$$

(7)

March 2, 2021

- Q.2. (a) Write python from scratch for computing 2D DFT{X(k,l)} of the following 2D array
 - (i) x(m,n) = np.array([[1, 0],[2, 1]]) (ii) x(m,n) = np.array([[1,2, 3,4], [5, 6, 7, 8], [9,10,11,12], [13,14,15,16]])
 - (b) Verify the results of part (a) by analytical solution method (i.e. pen and paper based solution)

```
[27]: import numpy as np
      import cmath
      def DFT2D(x):
          M, N = x.shape
          dft2d = np.zeros((M,N),dtype=complex)
          for k in range(M):
              for 1 in range(N):
                  sum_matrix = 0.0
                  for m in range(M):
                      for n in range(N):
                          e = cmath.exp(-2j * np.pi * ((k * m) / M + (1 * n) / N))
                          sum_matrix += x[m,n] * e
                  dft2d[k,1] = sum_matrix
          return dft2d
      x1=np.array([[1, 0],[2, 1]])
      y1=np.round(DFT2D(x1),2)
      print(y1)
      x2=np.array([[1,2, 3,4], [5, 6, 7, 8], [9,10,11,12], [13,14,15,16]])
      y2=np.round(DFT2D(x2),2)
      print("\n",y2)
```

```
[[4.+0.j 2.-0.j]

[-2.-0.j 0.+0.j]]

[[136. +0.j -8. +8.j -8. -0.j -8. -8.j]

[-32.+32.j 0. +0.j 0. +0.j -0. +0.j]

[-32. -0.j 0. +0.j 0. +0.j -0. +0.j]

[-32.-32.j -0. +0.j -0. +0.j -0. +0.j]]
```

Q.2 b)
$$F(u,v) = \sum_{x=0}^{M-1} \sum_{y=0}^{M-1} \{(x,y) \in \overline{0}^{3m}(\frac{ux}{M} + \frac{ux}{n^2})\}$$

$$F(o,o) = \sum_{x=0}^{M-1} \sum_{y=0}^{M-1} \{(x,y) \in \overline{0}^{3m}(\frac{ux}{M} + \frac{ux}{n^2})\}$$

$$F(o,o) = \sum_{x=0}^{M-1} \sum_{y=0}^{M-1} \{(x,y) \in \overline{0}^{3m}(\frac{ux}{M} + \frac{ux}{n^2})\}$$

$$F(o,o) = \sum_{x=0}^{M-1} \sum_{y=0}^{M-1} \{(x,y) \in \overline{0}^{3m}(\frac{ux}{M} + \frac{ux}{n^2})\}$$

$$= 1 + (-o) + (2) + (-i) = 2$$

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$$= 1 + (-o) + (-$$

$$F(0,3) = \sum_{x=0}^{3} \sum_{y=0}^{2} \{(x,y) e^{-\frac{1}{2}x^{2}} (\frac{1}{2}x^{2}) = \{(0,0) + \{(0,1)(\frac{1}{2}) + \{(0,1)(\frac{1}{2})$$

$$F(3,0) = \sum_{x=0}^{3} \sum_{x=0}^{2} \{(x,0)\} e^{-\frac{1}{2}\pi i \left(\frac{2x}{4x}\right)} = 4(0,0) +$$

Lab Assignment 3

March 3, 2021

- Q.3. (a) Write python from scratch for 2D Circular convolution using Doubly Block Circulant matrices method between input=np.array([[1,2,3],[4,5,6],[7,8,9]])and filter=np.array([[1,2,1],[0,0,0],[-1,-2,-1]])
 - (b) verify the result of part (a) by analytical solution method (i.e. pen and paper-based solution)

```
In [2]: import numpy as np
        from scipy import signal
        #a
        i=np.array([[1,2,3],[4,5,6],[7,8,9]])
        f=np.array([[1,2,1],[0,0,0],[-1,-2,-1]])
        #i=np.array([[1,2,1],[1,3,-1],[0,1,0]])
        #f=np.array([[1,-1,0],[1,0,0],[0,0,0]])
        #f=np.array([[1,-1],[1,0]])
        print("Input=\n",i,"\nFilter=\n",f)
        def cir_convolv2D(i,f):
            result_shape=[max(i.shape[0],f.shape[0]),max(i.shape[1],f.shape[1])]
            print("\nResult shape: ",result_shape)
            # Padding filter with zeros based on input size
            f1=np.zeros(i.shape)
            f1[0:f.shape[0],0:f.shape[1]] = f
            #Convert input to vector
            inp_vector = i.flatten()
            h=np.zeros((f1.shape[0],f1.shape[1],i.shape[1]))
            for row in range(f1.shape[0]):
                for col in range(i.shape[1]):
                    h[row,:,col]=np.roll(f1[row,:],col).transpose()
            #print("Computing")
```

```
H1 = H = np.hstack(np.concatenate(([h[0]],h[::-1][:i.shape[0]-1]),axis=0))
           for ctr in range(1,h.shape[0]):
               H1 = np.hstack((h[ctr],H1[:,:-i.shape[1]]))
               H = np.vstack((H,H1))
           print("Doubly Block Circulant Matrix")
           print(H)
           result = np.matmul(H, inp_vector)
           result = result.reshape(result_shape)
           return result
       result=cir_convolv2D(i,f)
       print("\nResult:\n",result)
       # Scipy method
       result_scipy=signal.convolve2d(i,f,boundary='wrap')[0:max(i.shape[0],f.shape[0]),0:max
       print("\nScipy result:\n",result_scipy)
Input=
 [[1 2 3]
 [4 5 6]
 [7 8 9]]
Filter=
 [[1 2 1]
 [0 0 0]
 [-1 -2 -1]]
Result shape: [3, 3]
Doubly Block Circulant Matrix
[[1. 1. 2. -1. -1. -2. 0. 0.
                                 0.]
[ 2. 1. 1. -2. -1. -1. 0. 0. 0.]
 [ 1. 2. 1. -1. -2. -1. 0. 0. 0.]
 [ 0. 0. 0. 1. 1. 2. -1. -1. -2.]
 [0. 0. 0. 2. 1. 1. -2. -1. -1.]
 [ 0. 0. 0. 1. 2. 1. -1. -2. -1.]
 [-1. -1. -2. 0. 0. 0. 1. 1. 2.]
 [-2. -1. -1. 0. 0. 0. 2. 1. 1.]
 [-1. -2. -1. 0. 0. 0. 1. 2.
                                 1.]]
Result:
 [[-12. -12. -12.]
 [-12. -12. -12.]
 [ 24. 24. 24.]]
Scipy result:
 [[-12 -12 -12]
 [-12 -12 -12]
 [ 24 24 24]]
```

3) b) 2-D Circular Convolution using Doubly Block Circulant matrix

$$f(m,n) = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$$
 $h[m,n] = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$

=>
$$M_1 = 3$$
, $N_1 = 3$ $M_2 = 3$ ξ $N_2 = 3$

$$M_1 = M_2 \quad \xi \quad N_1 = N_2$$

Length of
$$y[m,n] = M \times N$$

 $M = Manc(M_1, M_2) = 3$
 $N = Manc(N_1, N_2) = 3$

Constructing Circulant Matrix for each now of h[m,n]

$$H_{j} = \begin{bmatrix} h[j,0] & h[j,N-1] & \cdots & h[j,1] \\ h[j,1] & h[j,0] & \cdots & h[j,2] \\ \vdots & \vdots & \ddots & \vdots \\ h[j,N-1] & h[j,N-2] & \cdots & h[j,0] \\ NXN \end{bmatrix}$$

where $j = 0, 1, \dots, M-1$

$$H_0 = \begin{bmatrix} 1 & 1 & 2 \\ 2 & 1 & 1 \\ 1 & 2 & 1 \end{bmatrix} \qquad H_1 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \qquad H_2 = \begin{bmatrix} -1 & -1 & -2 \\ -2 & -1 & -1 \\ -1 & -2 & -1 \end{bmatrix}$$

$$H_2 = \begin{vmatrix} -1 & -1 & -2 \\ -2 & -1 & -1 \\ -1 & -2 & -1 \end{vmatrix}$$

Number of Columns =N= number of columns. of f[m,n]

$$H = \begin{bmatrix} H_0 & H_2 & H_1 \\ H_1 & H_0 & H_2 \\ H_2 & H_1 & H_0 \end{bmatrix}$$

$$g = Hf = \begin{bmatrix} 1 & 1 & 2 & -1 & -1 & -2 & 0 & 0 & 0 \\ 2 & 1 & 1 & -2 & -1 & -1 & 0 & 0 & 0 \\ 1 & 2 & 1 & -1 & -2 & -1 & 0 & 0 & 0 \\ 1 & 2 & 1 & -1 & -2 & -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 2 & -1 & -1 & -2 \\ 0 & 0 & 0 & 1 & 2 & 1 & 1 & 2 & -1 \\ 0 & 0 & 0 & 1 & 2 & 1 & 1 & 2 & -1 \\ -1 & -1 & -2 & 10 & 0 & 0 & 12 & 1 & 1 \\ -2 & -1 & -1 & 0 & 0 & 0 & 12 & 1 & 1 \\ -1 & -2 & -1 & 0 & 0 & 0 & 1 & 2 & 1 \end{bmatrix}$$

$$g[m,n] = \begin{bmatrix} -12 & -12 & -12 \\ -12 & -12 & -12 \\ 24 & 24 & 24 \end{bmatrix}$$
3x3

Q4

March 5, 2021

- Q.4. (i) Write python from scratch for 2D DCT transform on the following matrix: (a) f(m.n) = [90, 100; 100, 175]
- (b) f(m.n) = [10, 20, 30; 40 50 60; 70 80, 90; 100,110,120; 130 140, 150] and also comment on energy compaction property of DCT. (c) f(m.n) = monalisa.tif
 - (ii) Write python from scratch for 2D IDCT transform and reconstruct f(m,n) from part(i)

```
[1]: import cv2 import matplotlib.pyplot as plt
```

```
[2]: import numpy as np
```

Helper for DCT/IDCT

```
[3]: def A(i,Z):
    if i:
        return np.sqrt(2/Z)
    else:
        return np.sqrt(1/Z)
```

DCT

IDCT

```
[5]: def IDCT(X):
         M=X.shape[0]
         N=X.shape[1]
         x=np.zeros((M,N))
         for m in range(M):
             for n in range(N):
                 num=0
                 for k in range(M):
                      for 1 in range(N):
                          num+=X[k][1]*A(k,M)*A(1,N)*np.cos((2*m+1)*k*np.pi/(2*M))*np.
      \hookrightarrow \cos((2*n+1)*l*np.pi/(2*N))
                 x[m][n]=num
         return x
    Matrix 1
[6]: x=np.array([[90,100],[100,175]])
[6]: array([[ 90, 100],
            [100, 175]])
[7]: X=DCT(x)
     X
[7]: array([[232.5, -42.5],
            [-42.5, 32.5]
[8]: IDCT(X)
[8]: array([[ 90., 100.],
            [100., 175.]])
    The original matrix is obtained from the IDCT.
    Matrix 2
[9]: x=np.array([[10, 20, 30],
                  [40, 50, 60],
                  [70, 80, 90],
                  [100,110,120],
                  [130,140,150]])
     Х
[9]: array([[ 10, 20,
                         30],
            [ 40, 50, 60],
            [70, 80, 90],
            [100, 110, 120],
```

```
[130, 140, 150]])
```

```
[10]: X=DCT(x)
     np.around(X,2) #Values rounded for clarity
[10]: array([[ 309.84,
                       -31.62,
                                  0.
                                    ],
            [-163.65,
                         0.,
                                 -0.],
            [ -0. ,
                         0.
                                  0.
                                    ],
            [-14.76,
                         0.,
                                 -0.],
            [ -0. ,
                         0.,
                                 -0. ]])
[11]: IDCT(X)
[11]: array([[ 10.,
                    20.,
            [ 40., 50.,
                          60.],
            [70., 80., 90.],
            [100., 110., 120.],
            [130., 140., 150.]])
```

The original matrix is obtained from the IDCT.

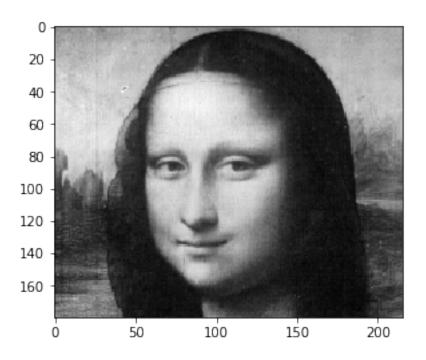
Energy Compaction

The DCT exhibits the property of energy compaction for correlated signals. The more correlated the input, the more concentrated is the energy of the output. This can be seen in the above two examples, as the DCT of the matrices tend to have larger values in the low-indexed coordinates (to the top-left). This is also true for the image given below.

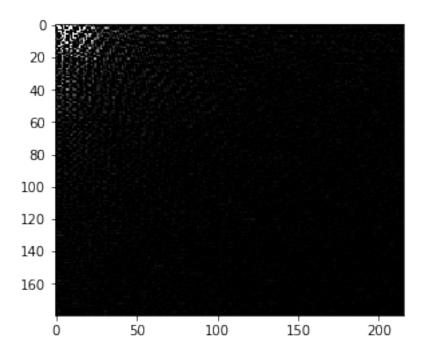
```
Image: monalisa.tif
```

```
[12]: im=cv2.imread('monalisa.tif')
plt.imshow(im)
```

[12]: <matplotlib.image.AxesImage at 0x26512c78080>



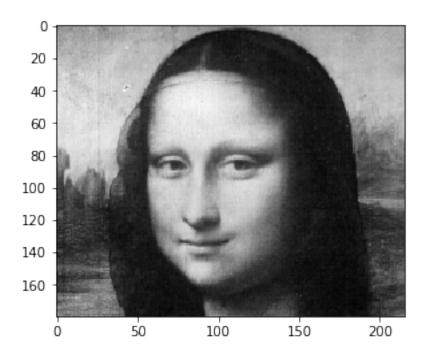
```
[13]: im.shape
[13]: (180, 216, 3)
[14]: img_flat = cv2.cvtColor(im, cv2.COLOR_BGR2GRAY) #To flatten image to 2D Matrix img_flat.shape
[14]: (180, 216)
[15]: img_DCT=DCT(img_flat)
[16]: plt.imshow(img_DCT, cmap='gray', vmin=0, vmax=255)
```



As previously noted under Energy Compaction, the higher values are concentrated at the top left of the image.

```
[17]: img_fin=IDCT(img_DCT)
[18]: plt.imshow(img_fin, cmap='gray')
```

[18]: <matplotlib.image.AxesImage at 0x26512d83278>



The original image is obtained from the IDCT.