

Assignment 3

COMP 478 Image Processing

Etienne Pham Do

40130483

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1.

$$\begin{aligned}
H(u, v) &= \sum_{x=-1}^1 \sum_{y=-1}^1 F(u, v) e^{-j2\pi(\frac{ux_0}{M} + \frac{vy_0}{N})} \\
&= \frac{1}{4} \sum_{x=-1}^1 e^{\frac{-j2\pi ux}{M}} \sum_{y=-1}^1 e^{\frac{-j2\pi vy}{N}} \\
&= \frac{1}{4} (e^{\frac{j2\pi u}{M}} + 0 + e^{\frac{-j2\pi u}{M}}) (e^{\frac{j2\pi v}{N}} + 0 + e^{\frac{-j2\pi v}{N}}) \\
&= \frac{1}{4} (2 \cos(\frac{2\pi u}{M})) (2 \cos(\frac{2\pi v}{N}))
\end{aligned}$$

2.

a)

Let $\text{Radon}(f(x, y)) = g(\rho, \theta)$, and f and g be $f(x, y)$ and $g(x, y)$

$$\begin{aligned}
\text{Radon}(af + bg) &= \iint_{-\infty}^{\infty} (af(x, y) + bg(x, y)) \delta(x \cos \theta + y \sin \theta - \rho) dx dy \\
&= a \text{Radon}(f(x, y)) + b \text{Radon}(g(x, y)) \\
\text{Radon}(af + bg) &= a \text{Radon}(f) + b \text{Radon}(g)
\end{aligned}$$

b)

$$\begin{aligned}
\text{Radon}(f(x - x_0, y - y_0)) &= \iint_{-\infty}^{\infty} f(x - x_0, y - y_0) \delta((x - x_0) \cos \theta + (y - y_0) \sin \theta - \rho) dx dy \\
&= \iint_{-\infty}^{\infty} f(x - x_0, y - y_0) \delta(x \cos \theta - x_0 \cos \theta + y \sin \theta - y_0 \sin \theta - \rho) dx dy \\
&= g(\rho - x_0 \cos \theta - y_0 \sin \theta, \theta)
\end{aligned}$$

Programming

1.

```
import numpy as np
```

```
import cv2
```

```
#loading the images
```

```
la = cv2.imread('house.tif', 0)
```

```
lb = cv2.imread('jet.tif', 0)
```

```
#applying the fourier transforms and center the transforms
```

```
Fa = cv2.dft(np.float32(la), flags = cv2.DFT_COMPLEX_OUTPUT)
```

```
dft_shift_a = np.fft.fftshift(Fa)
```

```
Fb = cv2.dft(np.float32(lb), flags = cv2.DFT_COMPLEX_OUTPUT)
```

```
dft_shift_b = np.fft.fftshift(Fb)
```

```
#getting the magnitudes and phases
```

```
magA, phaseA = cv2.cartToPolar(dft_shift_a[:, :, 0], dft_shift_a[:, :, 1])
```

```
magB, phaseB = cv2.cartToPolar(dft_shift_b[:, :, 0], dft_shift_b[:, :, 1])
```

```
#applying the switch in phases and merge the values in new frequency arrays
```

```
realA, imagA = cv2.polarToCart(magA, phaseB)
```

```
realB, imagB = cv2.polarToCart(magB, phaseA)
```

```
mergedCartA = cv2.merge([realA, imagA])
```

```
mergedCartB = cv2.merge([realB, imagB])
```

```
#undoing the shift
```

```
mergedCartA_ishift = np.fft.ifftshift(mergedCartA)
```

```
mergedCartB_ishift = np.fft.ifftshift(mergedCartB)
```

```

#inverse fourier transform

newA = cv2.idft(mergedCartA_ishift)
newB = cv2.idft(mergedCartB_ishift)

newA = cv2.magnitude(newA[:,0], newA[:,1])
newB = cv2.magnitude(newB[:,0], newB[:,1])

#converting back to 8 bit images from 32 bit

newA = cv2.normalize(newA, None, alpha=0, beta=255, norm_type=cv2.NORM_MINMAX,
dtype=cv2.CV_8U)

newB = cv2.normalize(newB, None, alpha=0, beta=255, norm_type=cv2.NORM_MINMAX,
dtype=cv2.CV_8U)

#showing the results

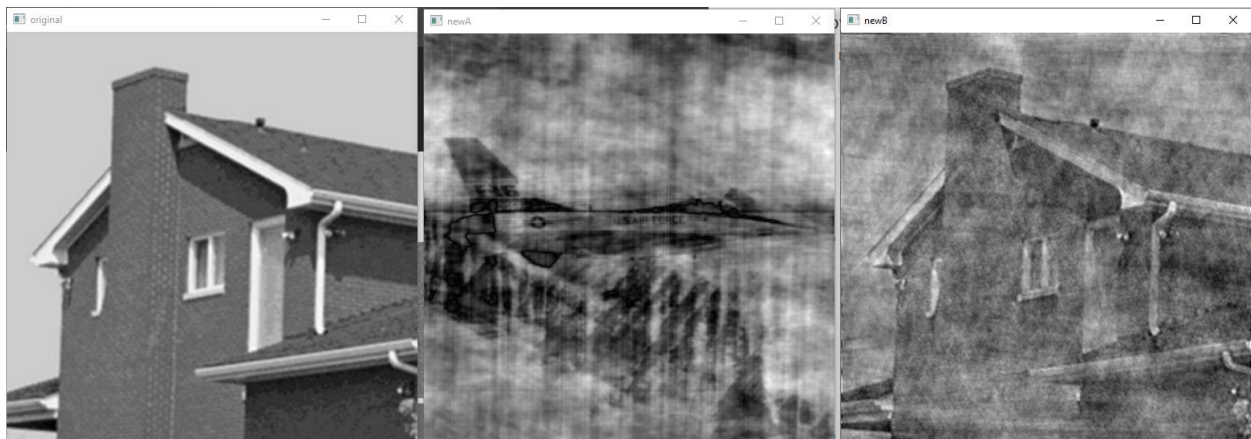
cv2.imshow('original', Ia)
cv2.imshow('newA', newA)
cv2.imshow('newB', newB)

cv2.waitKey(0)

cv2.destroyAllWindows()

```

I_2 should have a reconstruction closer to the original image I_A since it has the phase of F_A , which has information about the shape features of I_A .



2.

```
import numpy as np
```

```
import cv2
```

```
import time
```

```
houseImg = cv2.imread('house.tif', 0)
```

```
#Laplacian of Gaussian
```

```
start = time.time()
```

```
gaussianBlur = cv2.GaussianBlur(houseImg, (7,7), 0)
```

```
laplacianOfGuassian = cv2.Laplacian(gaussianBlur, cv2.CV_64F)
```

```
end = time.time()
```

```
print(str(end - start))
```

```
#Canny edge detection
```

```
start = time.time()
```

```
cannyEdge = cv2.Canny(houseImg, 100, 200)
```

```
end = time.time()
```

```
print(str(end - start))
```

```
cv2.imshow('LoG', laplacianOfGuassian)
```

```
cv2.imshow('Canny', cannyEdge)
```

```
cv2.waitKey(0)
```

```
cv2.destroyAllWindows()
```

a)

Laplacian of Gaussian:

- apply Gaussian blur on image

- apply Laplacian on Gaussian

Canny edge detection:

- Noise reduction
- Gradient calculation
- Non-maximum suppression
- Double threshold
- Edge linking

b)

Edge linking consists of transforming low intensity pixels into high intensity pixels

if there are neighboring high intensity pixels. The first method needs this step as it would be useful to reduce noise.

c)

Laplacian of Gaussian: kernel size affects performance, used size 7x7 since it makes the performance faster and has less noise.

Canny edge: lower and upper thresholds affect performance, used 100 and 200 since it makes the performance faster and has less noise.

d)

Laplacian of Gaussian shows the edges of the house, but has more noise than the Canny edge output, which shows the general outline of the house.

