Q1:

CF = 5

1. Ideal Lowpass Filter:







CF = 80

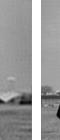






Original











As the cutoff frequency increased, the clearer the overall image became. This is especially evident when comparing the 5 and 210 cutoff frequencies. A cutoff frequency of 10 and 40 show significant ringing effects that are visible in the filtered image.

2. Butterworth Lowpass Filter:





CF = 40



CF = 80



CF = 120







The Butterworth filtering produced much clearer images than the ideal lowpass filter at lower cutoff frequencies and did not result in any noticeable ringing effects. The images at lower cutoff frequencies however do remain blurry. Higher cutoff frequencies resulted in clearer images.

3. Guassian Lowpass Filter:



CF = 10

CF = 40

CF = 80



CF = 210



Original

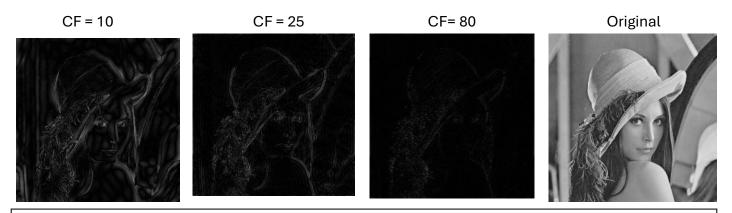






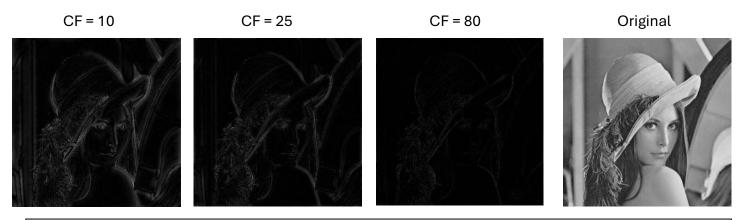
The Gaussian filtering produced no visible ringing effect but performed similarly to Butterworth at lower cutoff frequencies. In fact, Butterworth seemed to produce a marginally better image at 5 and 10 cutoff frequencies. Overall the image at cutoff frequency 210 was very similar to the Butterworth filter at the same frequency.

Q2:1.<u>Ideal Highpass Filter:</u>



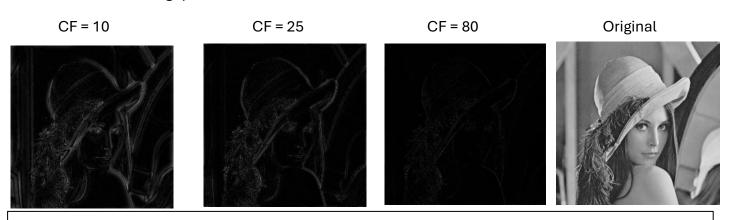
The Ideal Highpass Filter resulted in significant ringing effects at cutoff frequencies 10 and 25. These effects are not noticeable at a cuttoff frequency of 80.

3. Butterworth Highpass Filter:



The Butterworth Highpass Filter resulted in less ringing effects at cutoff frequencies 10 and 25 than the ideal highpass filter. These effects are not noticeable at a cuttoff frequency of 80.

4. Guassian Highpass Filter:



The Gaussian Highpass Filter produced results that were nearly identical as the results of the Butterworth highpass filter for this image. Gaussian Highpass Filter produced much better results than the ideal highpass filter at lower cutoff frequencies, similar to the Butterworth highpass filter.

Ideal Lowpass Filter Code

Gaussian Low Pass Filter Code

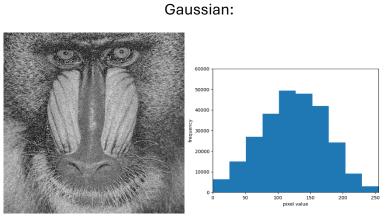
Butterworth High Pass Filter

Butterworth Lowpass Filter

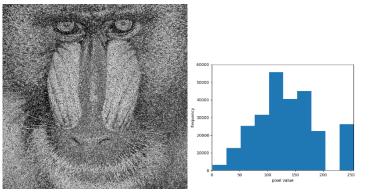
Ideal High Pass Filter Code

Gaussian High Pass Filter Code

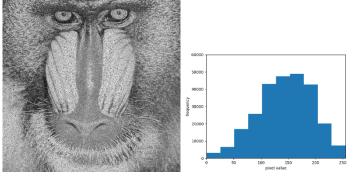
Q3:



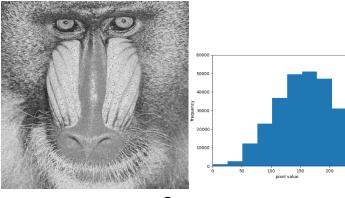
Salt and Pepper:



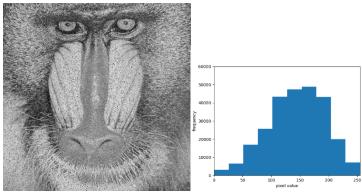
Exponential



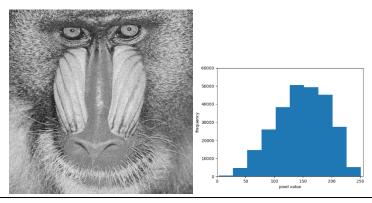
Rayleight:



Gamma:



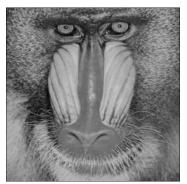
Uniform:



Each method produced different results when applied to the original image. Salt and Pepper noise appeared to have the largest impact on the clarity of the image. Uniform and Rayleigh noise seemed to have the smallest impact on visual clarity.

3. Gaussian Noisy image:

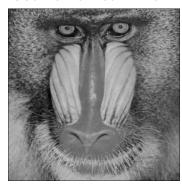
Arithmetic Mean Filter:



Both filters restored the image to almost its original condition.

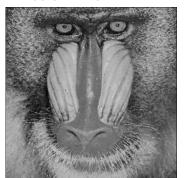
There was no noticeable difference between the results of the two filters.

Geometric Mean Filter:



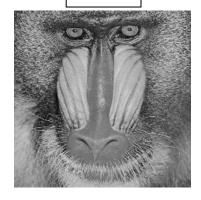
4. Salt and Pepper:

Median Filter:



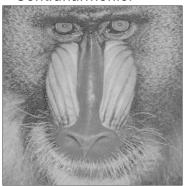
The median filter removed most of the salt and pepper noise with some still remaining. The finer details of the image were slightly blurred.

Original



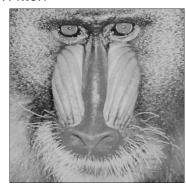
5.Pepper Noise:

Contraharmonic:



The max filter produced an image with less detail than the original with some artifacts from the noise still present. The contraharmonic filter produced an image that still maintained some artifacts from the noise, but maintained more fine details of the original image.

Max Filter:



Contraharmonic:



The Min Filter produced a cleaned image resulting in inflated dots. Overall, the resulting image appears to be low quality. The contraharmonic filter preserved smaller details present in the original image, however there is still a noticeable amount of salt noise.

Min Filter:



Code For Question 3:

Gaussian Noise

AddGaussianNoise():

Rayleigh Noise

```
def AddRayleighNoise():
    imagepath = 'C:/Users/caden/Downloads/standard_test_images/standard_test_images/mandril_color.tif
    f = cv2.imread(imagepath, 0)
    f = f/255
    # create rayleigh noise
    x, y = f.shape
    #mean = 0
    var = 0.01
    sigma = np.sqrt(var)
    n = np.random.rayleigh(scale=sigma, size=(x,y))
    # add a rayleigh noise
    g = f + n
    g = (g*255).round().astype[np.uint8)|
    plt.hist(g.flat)
    plt.xlabel('pixel value'); plt.ylabel('frequency')
    plt.show()
    os.chdir('C:/Users/caden/CompVision')
    filename = 'RayleighNoise.jpg'
    #saves file
    cv2.imwrite(filename,g)
```

С

Salt and Pepper

```
AddSaltandPepper(opt):
imagepath = 'C:/Users/caden/Downloads/standard_test_images/standard_test_images/mandril_color.tif
img = cv2.imread(imagepath, 0)
 img = img/255
x,y = img.shape
if(opt == 0):
if(opt == 1):
    pepper = 0
     pepper = 0.1
 for i in range(x):
          if rdn < pepper:
    g[i][j] = 0
          g[i][j] = 1
else:
g[i][j] = img[i][j]
#used for final image formating
#g = (g*255).round().astype(np.uint8)
cv2.imshow('image with noise', g)
cv2.waitKey(0)
cv2.destroyAllWindows()
plt.hist(g.flat)
plt.xlim([0,255]); plt.ylim([0,60000])
plt.xlabel('pixel value'); plt.ylabel('frequency')
```

Gamma Noise

```
def AddGammaNoise():
    imagepath = 'C:/Users/caden/Downloads/standard_test_images/standard_test_images/mandril_color.tif'
    f = cv2.imread(imagepath, 0)
    f = f/255
    # create gamma noise
    x, y = f.shape
    #mean = 0
    var = 0.01
    sigma = np.sqrt(var)
    n = np.random.gamma(1, scale=sigma, size=(x,y))
# add a gamma noise
    g = f + n
    g = (g*255).round().astype(np.uint8)
    plt.hist(g.flat)
    plt.xlim([0,255]); plt.ylim([0,60000])
    plt.xlabel('pixel value'); plt.ylabel('frequency')
    plt.show()
    os.chdir('C:/Users/caden/CompVision')
    filename = 'GammaNoise.jpg'
    #saves file
    cv2.imwrite(filename,g)
```

Exponential Noise Code

```
def AddExpNoise():
    imagepath = 'c:/Users/caden/Downloads/standard_test_images/standard_test_images/mandril_color.tif'
    f = cv2.imread(imagepath, 0)
    f = f/255
    # create gamma noise
    x, y = f.shape
    #man = 0
    var = 0.01
    sigma = np.sqrt(var)
    n = np.random.exponential(scale=sigma, size=(x,y))
# add a gamma noise
    g = f + n
    g = (g*255).round().astype(np.uint8)
    cv2.imshow('corrupted Image', g)
    cv2.waitKey(0)
    cv2.destroyAllWindows()

# hitogram: noise image
plt.hist(g.flat)
plt.xlabel('pixel value'); plt.ylabel('frequency')
plt.xlabel('pixel value'); plt.ylabel('frequency')
plt.xlabel('pixel value'); plt.ylabel('frequency')
plt.show()
    os.chdir('C:/Users/caden/CompVision')
    filename = 'ExpNoise.jpg'
#saves file
    cv2.imwrite(filename,g)
```

Geometric Mean Filter Code

Contraharmonic Filter Code

```
def Contra(img):
    m, n = img.shape
# Develop Averaging filter(3, 3) mask
mask = np.ones([3, 3], dtype = int)
# Convolve the 3X3 mask over the image
ing_new = np.zeros([m, n])
Q = 1.5
for i in range(1, m-1):
    top = (img[i-1, j-1]*mask[0, 0])** (Q+1)
        + (img[i-1, j]*mask[1, 1])** (Q+1) + (img[i-1, j + 1]*mask[0, 2])** (Q+1)
        + (img[i, j-1]*mask[1, 2])** (Q+1) + (img[i, j]*mask[1, 1])** (Q+1)
        + (img[i, j+1]*mask[2, 2])** (Q+1) + (img[i + 1, j-1]*mask[2, 0])** (Q+1)
        + (img[i, j+1]*mask[2, 1])** (Q+1) + img[i + 1, j-1]*mask[2, 2]

    bottom = (img[i-1, j-1]*mask[0, 0])** (Q) + (img[i-1, j]*mask[0, 1])** (Q)
        + (img[i, j]*mask[1, 1])** (Q) + (img[i, j+1]*mask[1, 2])** (Q)
        + (img[i, j]*mask[2, 0])** (Q) + (img[i, j+1]*mask[1, 2])** (Q)
        + (img[i, j]*mask[2, 0])** (Q) + (img[i, j+1]*mask[2, 1])** (Q)
        + img[i + 1, j+1]*mask[2, 2]

    if(bottom == 0):
        img_new[i, j] = img[i, j]
    else:
        img_new[i, j] = top/bottom

img_new= (img_new*255).round().astype(np.uint8)
    os.chdir('C:/Users/caden/CompVision')
filename = 'ContraFilterForSalt.jpg'
    #saves file
    cv2.imwrite(filename,img_new)
```

Uniform Noise Code

Median, Min, and Max Filter Code

Arithmetic Mean Filter

References:

https://github.com/Shahir-Abdullah/Digital-Image-Processing/tree/master

https://github.com/adenarayana/digital-image-processing/tree/main

https://www.geeksforgeeks.org/spatial-filters-averaging-filter-and-median-filter-in-image-processing/