

ECG 772 HW 1

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1 Histogram Equalization

1.1 Write a function to perform histogram equalization on gray-scale images

```
%performs histogram equalization on grayscale image

function [equalizedImage] = histEqualize(image, numberOfBins)
    [row,column] = size(image);
    pixelIntensities = 255;
    image = double(image(:));
    histogram = hist(image, numberOfBins)/length(image)';
    cumulativeSum = cumsum(histogram);
    image = floor(image*numberOfBins/pixelIntensities);
    equalizedImage = floor(cumulativeSum(image)*pixelIntensities);
    equalizedImage = uint8(reshape(equalizedImage,row,column));
end
```

1.2 Perform histogram equalization on 'jetplane.tif'.

The images equalized with a smaller number of bins have less detail.

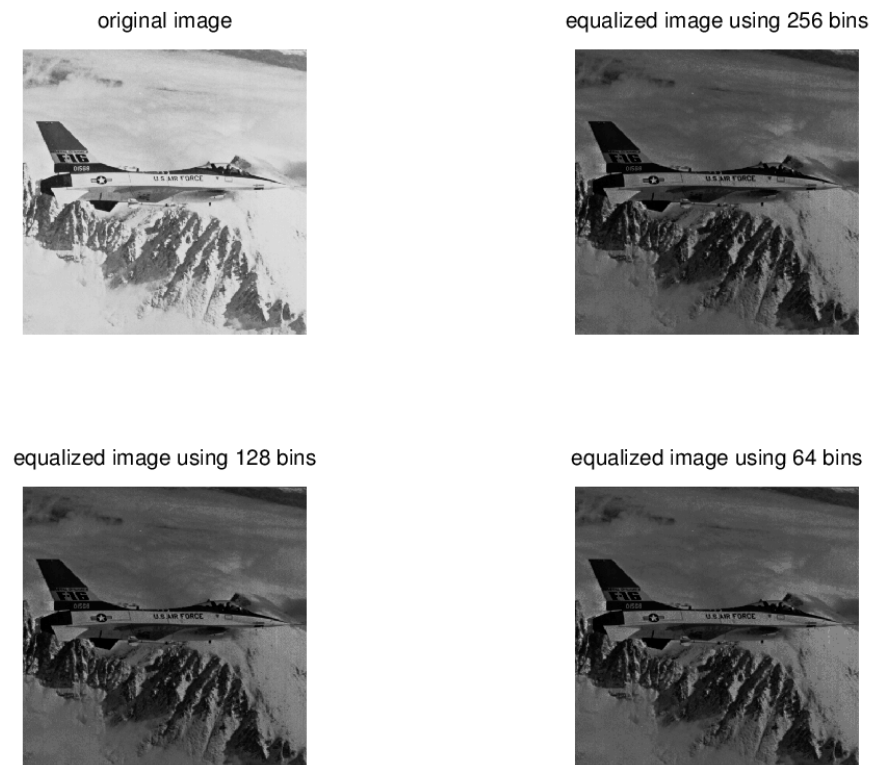


Figure 1: equalized images

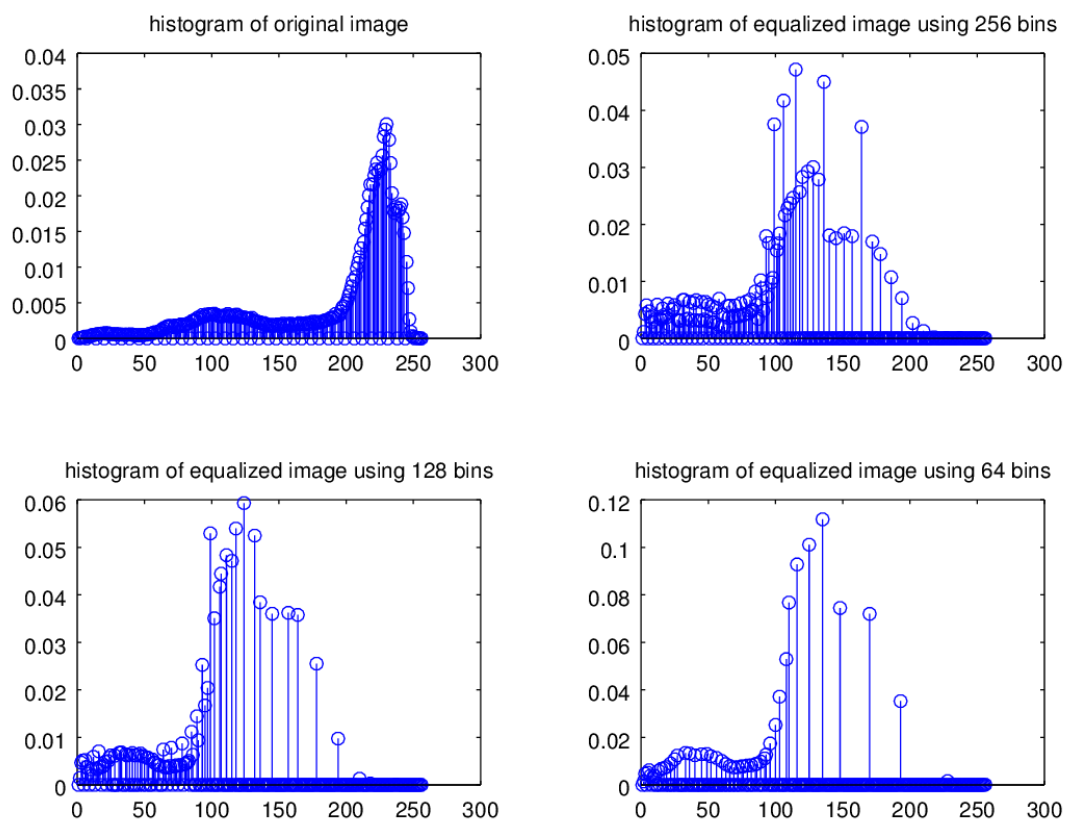


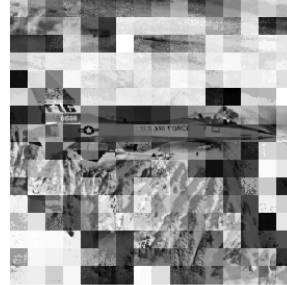
Figure 2: histograms of equalized images

1.3 Perform histogram equalization on 'jetplane.tif' using 32x32 blocks.

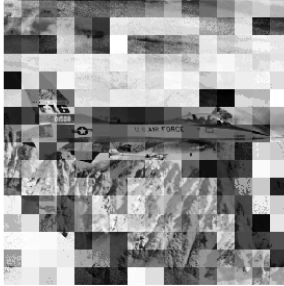
original image



equalized image using 256 bins in 32x32 blocks



equalized image using 128 bins in 32x32 blocks



equalized image using 64 bins in 32x32 blocks

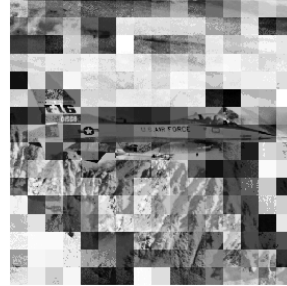


Figure 3: equalized images using 32x32 blocks

2 Basic Morphology

2.1 Detect faces of soccer players in team photos.

In order to detect the faces of soccer players I applied thresholding in HSV color space. To find the threshold ranges, I opened the image in GIMP and looked at the color of the pixels on the players foreheads. This gave me a threshold of S greater than 20, V greater 45, and H between 40 and 25. However, later I adjusted these values to S greater than 20, V greater than 20, and H between 35 and 5 to produce better results. After thresholding, I applied various morphological operations to get rid of traces of the crowd and to enhance objects with an aspect ratio of 3:2. The reason for this aspect ratio is that the faces of the players tend to be longer vertically and their knees tend to be longer horizontally. To achieve this I performed dilation using structural elements with a 3:2 aspect ratio and dilation with structural elements that were square.

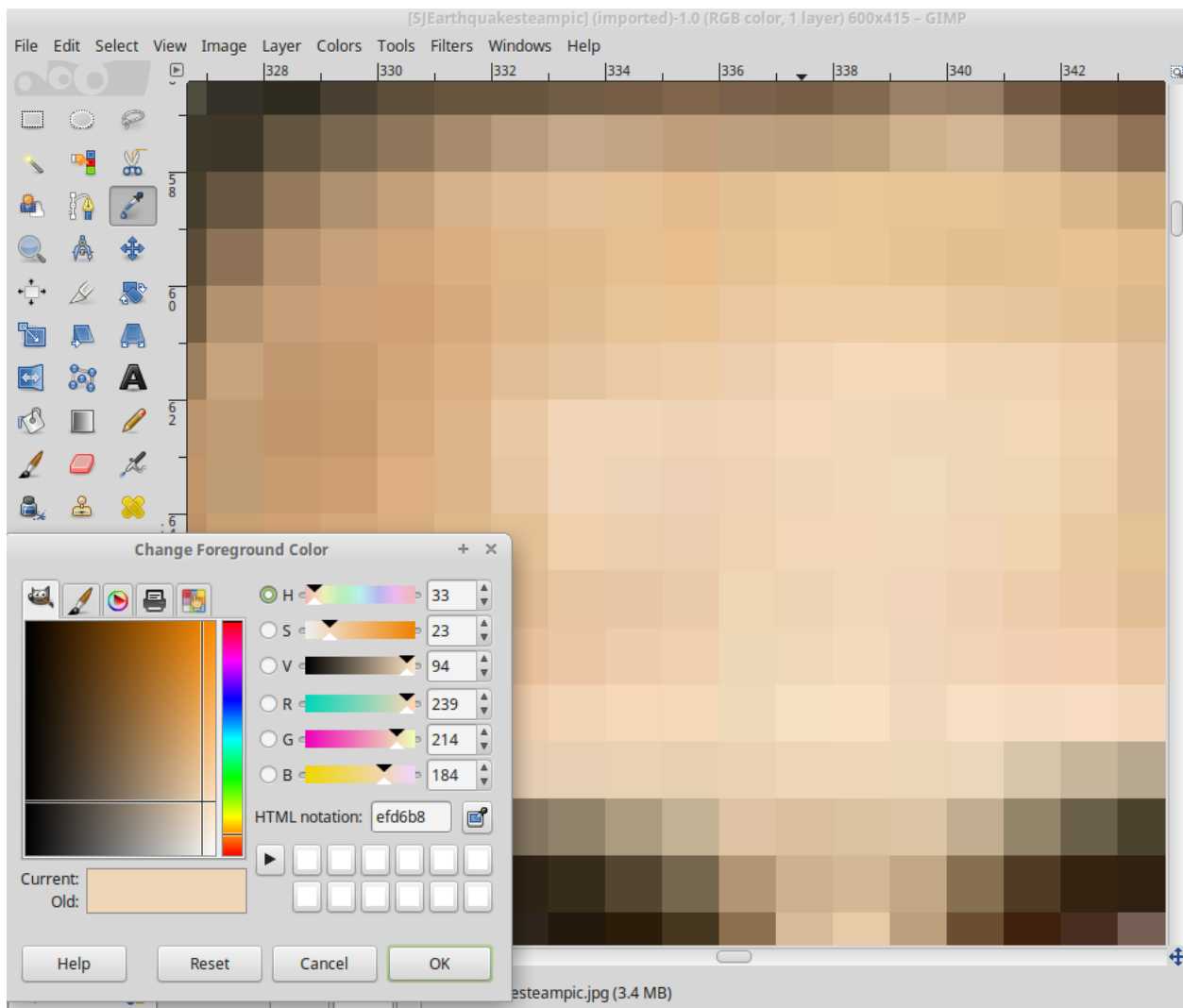


Figure 4: selection of threshold

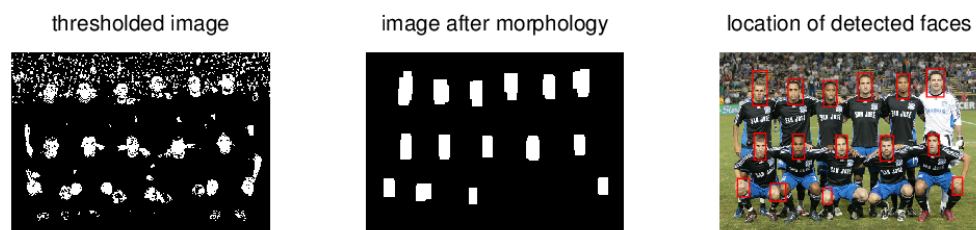


Figure 5: face detection on 'SJEarthquakesteampic.jpg'

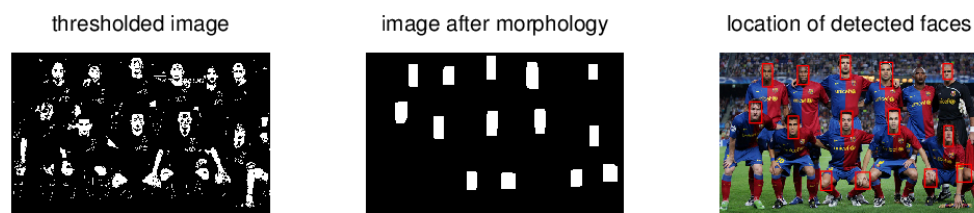


Figure 6: face detection on 'barcelona-team.jpg'

The results for the two images are similar. In the image of the Earthquakes almost all of the player's faces are detected with a few knees selected as faces. The player whose face is not detected has a darker skin tone and a round face. Both of these characteristics could be the reason for his face not being correctly detected. In both images knees that appear longer in the y direction were detected as faces.

3 Filtering

3.1 Use various filters to remove Gaussian noise with .005 variance from 'DSCN0479-001.JPG'.

Looking at the output images there seems to be a trade off between noise reduction and sharpness. The bigger filters make noise less noticeable, while at the same time making the image blurrier. The smaller filters remove less noise, but keep more of the sharpness of the image. The median filters appear to remove the same amount of noise as the mean filters, while preserving more sharpness.

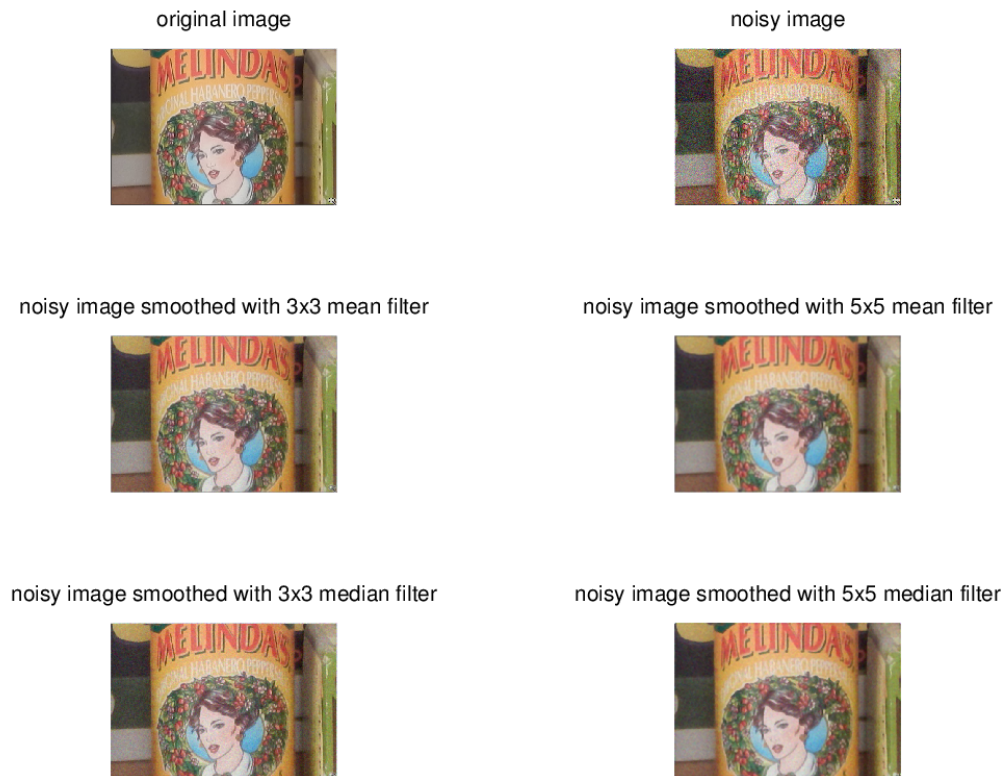


Figure 7: images after filtering

filter	MSE	PSNR
none	72.0	29.6
3x3 mean filter	32.9	33.0
5x5 mean filter	36.7	32.5
3x3 median filter	37.2	32.4
5x5 median filter	36.2	32.5

Table 1: SNR and PSNR after filtering the noisy image

The image with the smallest MSE and highest PSNR is the one produces by the 3x3 mean filter.

3.2 Use various filters to remove salt and pepper noise with .05 density from 'DSCN0479-001.JPG'.

Similar to the previous results, the bigger filters make the image blurrier and remove more noise. The median filters perform much better than the mean filters on salt and pepper noise. The best looking image is the one produced by the 3x3 median filter, since most of the noise is gone and the image is sharper than the one produced by the 5x5 median filter.



Figure 8: images after filtering

filter	MSE	PSNR
none	6.5	40.0
3x3 mean filter	42.4	31.9
5x5 mean filter	44.9	31.6
3x3 median filter	10.7	37.8
5x5 median filter	24.7	34.2

Table 2: SNR and PSNR after filtering the noisy image

The image with the smallest MSE and highest PSNR is the unfiltered noisy image. The cause of this is probably the density of errors in the filtered images. In other words the filtered images have small errors in every pixel, while the unfiltered image has large errors in a small number of images. This brings into question the reliability of MSE and PSNR as useful measures to compare the results of filtered images.

3.3 Use various filters to clean 'DSCN0482-001.JPG' a noisy version of 'DSCN0479-001.JPG'.

Just like before, the bigger filters produced blurrier images with less noise. However, the difference much smaller. The mean and median filters seem to produce very similar results. This makes the trade-off in performance undesirable. The 3x3 median filter takes 8 times longer than the 3x3 mean filter to produce an image, while the 5x5 median filter takes 16 times longer than the 5x5 mean filter to produce an image.

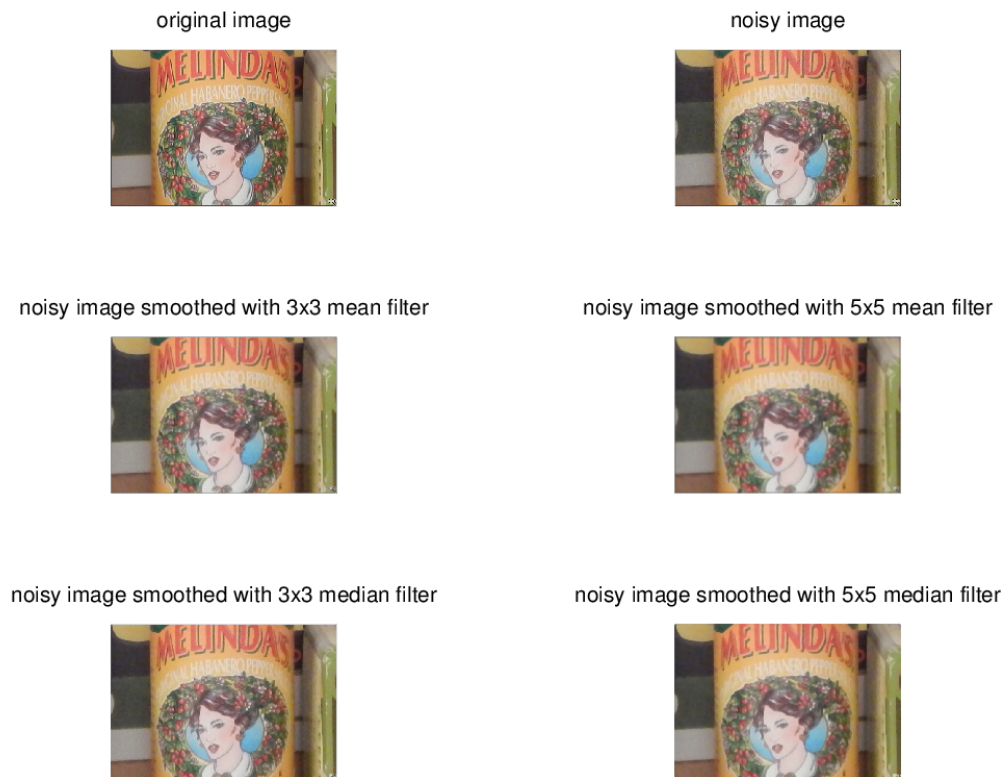


Figure 9: images after filtering

4 Code

4.1 Q1 code

histEqualize.m

```
%performs histogram equalization on grayscale image

function [equalizedImage] = histEqualize(image, numberOfBins)
    [row,column] = size(image);
    pixelIntensities = 255;
    image = double(image(:));
    histogram = hist(image, numberOfBins)/length(image)';
    cumulativeSum = cumsum(histogram);
    image = floor(image*numberOfBins/pixelIntensities);
    equalizedImage = floor(cumulativeSum(image)*pixelIntensities);
    equalizedImage = uint8(reshape(equalizedImage,row,column));
end
```

Q1PartB.m

```
bins = [256 128 64]';
figure(1);
image = imread('jetplane.tif');
numberOfPixels = length(image(:));
subplot(2,2,1);
imshow(image);
title('original_image');
figure(2);
subplot(2,2,1);
normalizedHistogram = hist(image(:),256)/numberOfPixels;
stem(normalizedHistogram);
title('histogram_of_original_image');
for i=1:length(bins)
    figure(1);
    eqImage = histEqualize(image,bins(i,1));
    subplot(2,2,i+1);
    imshow(eqImage);
    str = sprintf('equalized_image_using_%d_bins', bins(i,1));
    title(str);
    figure(2);
    subplot(2,2,i+1);
    normalizedHistogram = hist(eqImage(:),256)/numberOfPixels;
    stem(normalizedHistogram);
    str = sprintf('histogram_of_equalized_image_using_%d_bins', bins(i,1));
    title(str);
end
print(figure(1),'outputImages.png');
print(figure(2),'outputHistograms.png');
```

Q1PartC.m

```
bins = [256 128 64]';
figure(1);
image = imread('jetplane.tif');
numberOfPixels = length(image(:));
```

```

subplot(2,2,1);
imshow(image);
title('original_image');
for i=1:length(bins)
    figure(1)
    eqImage = blockproc(image,[32,32], @histEqualize,bins(i,1));
    subplot(2,2,i+1);
    imshow(eqImage);
    str = sprintf('equalized_image_using_%d_bins_in_32x32_blocks', bins(i,1));
    title(str);
end
print(figure(1),'outputImages32.png');

```

4.2 Q2 code

thresholdFaces.m

```
function [thresholdedImage] = thresholdFaces(image)
    HUpper = 35/360;
    HLower = 5/360;
    SLower = 20/100;
    VLower = 20/100;
    image = double(image)/255;
    imageHSV = rgb2hsv(image);
    mask = (HUpper >= imageHSV(:, :, 1)) .* (imageHSV(:, :, 1) >= HLower) .* (
        imageHSV(:, :, 2) >= SLower) .* (imageHSV(:, :, 3) > VLower);
    thresholdedImage = mat2gray(mask);
end
```

faceDetect.m

```
function [] = faceDetect(filename);

    image = imread(filename);
    thresholdedImage = thresholdFaces(image);
    figure(1);
    subplot(1,3,1);
    imshow(thresholdedImage);
    title('thresholded_image');

    structuralElement = strel('rectangle', [9 6]);
    morphImage = imdilate(thresholdedImage, structuralElement);

    structuralElement = strel('square', 27);
    morphImage = imerode(morphImage, structuralElement);
    structuralElement = strel('rectangle', [15 10]);
    morphImage = imdilate(morphImage, structuralElement);
    structuralElement = strel('square', 9);
    morphImage = imerode(morphImage, structuralElement);
    structuralElement = strel('rectangle', [30 20]);
    morphImage = imdilate(morphImage, structuralElement);
    figure(1)
    subplot(1,3,2);
    imshow(morphImage);
    title('image_after_morphology');

    subplot(1,3,3);
    imshow(image);
    title('location_of_detected_faces');
    labeledImage = bwlabel(morphImage);
    boundingBoxes = regionprops(labeledImage, 'BoundingBox');
    for i=1:rows(boundingBoxes)
        rectangle('Position', boundingBoxes(i,1).BoundingBox, 'EdgeColor', 'red');
    end

    print(figure(1), ['faceDetect' filename(1:end-3) 'png']);
end
```

4.3 Q3 code

Q3.m

```
function [] = Q3(originalImage , noisyImage);

    figure(1);
    subplot(3,2,1)
    imshow(originalImage);
    title('original_image');

    subplot(3,2,2)
    imshow(noisyImage);
    str = 'noisy_image'
    title(str);
    MSE = imageMSE(originalImage ,noisyImage)
    PSNR = imagePSNR(originalImage ,noisyImage)

    subplot(3,2,3)
    tic()
    smoothedImage = imsmooth(noisyImage , 'Average' ,3);
    time = toc();
    imshow(smoothedImage);
    str = 'noisy_image_smoothed_with_3x3_mean_filter'
    time
    title(str);
    MSE = imageMSE(originalImage ,smoothedImage)
    PSNR = imagePSNR(originalImage ,smoothedImage)

    subplot(3,2,4)
    tic()
    smoothedImage = imsmooth(noisyImage , 'Average' ,5);
    time = toc();
    imshow(smoothedImage);
    str = 'noisy_image_smoothed_with_5x5_mean_filter'
    time
    title(str);
    MSE = imageMSE(originalImage ,smoothedImage)
    PSNR = imagePSNR(originalImage ,smoothedImage)

    subplot(3,2,5)
    tic()
    smoothedImage = imsmooth(noisyImage , 'Median' ,3);
    time = toc();
    imshow(smoothedImage);
    str = 'noisy_image_smoothed_with_3x3_median_filter'
    time
    title(str);
    MSE = imageMSE(originalImage ,smoothedImage)
    PSNR = imagePSNR(originalImage ,smoothedImage)

    subplot(3,2,6)
    tic()
    smoothedImage = imsmooth(noisyImage , 'Median' ,5);
    time = toc();
```

```

imshow(smoothedImage);
str = 'noisy_image_smoothed_with_5x5_median_filter '
time
title(str);
MSE = imageMSE(originalImage,smoothedImage)
PSNR = imagePSNR(originalImage,smoothedImage)

```

PartA.m

```

originalImage = imread('DSCN0479-001.JPG');
noisyImage = imnoise(originalImage, 'gaussian', 0, .005);
Q3(originalImage, noisyImage);
print(figure(1), 'partA.png');

```

PartB.m

```

originalImage = imread('DSCN0479-001.JPG');
noisyImage = imnoise(originalImage, 'salt & pepper', .05);
Q3(originalImage, noisyImage);
print(figure(1), 'partB.png');

```

PartC.m

```

figure (1, 'visible', 'off');
originalImage = imread('DSCN0479-001.JPG');
noisyImage = imread('DSCN0482-001.JPG');
Q3(originalImage, noisyImage);
print(figure(1), 'partC.png');

```

4.4 latex

```
\documentclass{article}
\usepackage{listings}
\lstset{
    frame=single ,
    breaklines=true
}
\usepackage{geometry}
\geometry{margin=1in}
\usepackage{graphicx}
\usepackage{float}
\lstset{language=Octave}
\title{ECG 772 HW 1}
\date{2015-09-19}
\author{Carlo Lopez-Tello}

\begin{document}
\maketitle
    \newpage
    \section{Histogram Equalization}

    \subsection{Write a function to perform histogram equalization on gray
        -scale images}
    \lstinputlisting[language=Octave]{Q1/histEqualize.m}

    \subsection{Perform histogram equalization on 'jetplane.tif'.}
    The images equalized with a smaller number of bins have less detail.
    \begin{figure}[H]
        \includegraphics[width=\linewidth]{Q1/outputImages.png}
        \caption{equalized images}
    \end{figure}
    \begin{figure}[H]
        \includegraphics[width=\linewidth]{Q1/outputHistograms.png}
        \caption{histograms of equalized images}
    \end{figure}

    \subsection{Perform histogram equalization on 'jetplane.tif' using 32
        x32 blocks.}
    \begin{figure}[H]
        \includegraphics[width=\linewidth]{Q1/outputImages32.png}
        \caption{equalized images using 32x32 blocks}
    \end{figure}

    \newpage
    \section{Basic Morphology}

    \subsection{Detect faces of soccer players in team photos.}
```

In order to detect the faces of soccer players I applied thresholding in HSV color space. To **find** the threshold ranges, I opened the **image** in GIMP and looked at the color of the pixels on the players foreheads. This gave me a threshold of S greater than 20, V

greater 45, and H between 40 and 25. However, later I adjusted these values to S greater than 20, V greater than 20, and H between 35 and 5 to produce better results. After thresholding, I applied various morphological operations to **get** rid of traces of the crowd and to enhance objects with an aspect ratio of 3:2. The reason **for** this aspect ratio is that the faces of the players tend to be longer vertically and their knees tend to be longer horizontally. To achieve this I performed dilation using structural elements with a 3:2 aspect ratio and dilation with structural elements that were square.

```
\begin{figure}[H]
    \includegraphics[width=\linewidth]{Q2/ForeheadHSV.png}
    \caption{selection of threshold}
\end{figure}
\begin{figure}[H]
    \includegraphics[width=\linewidth]{Q2/
        faceDetectSJEarthquakesteampic.png}
    \caption{face detection on 'SJEarthquakesteampic.jpg'}
\end{figure}
\begin{figure}[H]
    \includegraphics[width=\linewidth]{Q2/faceDetectbarcelona-team
        .png}
    \caption{face detection on 'barcelona-team.jpg'}
\end{figure}
```

The results **for** the two images are similar. In the **image** of the Earthquakes almost **all** of the player's faces are detected with a few knees selected as faces. The player whose face is not detected has a darker skin tone and a **round** face. Both of these characteristics could be the reason **for** his face not being correctly detected. In both images knees that appear longer in the y direction were detected as faces.

```
\newpage
\section{Filtering}
\subsection{Use various filters to remove Gaussian noise with .005
    variance from 'DSCN0479-001.JPG'.}
```

Looking at the output images there seems to be a trade off between noise reduction and sharpness. The bigger filters make noise less noticeable, **while** at the same time making the **image** blurrier. The smaller filters remove less noise, but keep **more** of the sharpness of the **image**. The **median** filters appear to remove the same amount of noise as the **mean** filters, **while** preserving **more** sharpness.

```
\begin{figure}[H]
    \includegraphics[width=\linewidth]{Q3/partA.png}
    \caption{images after filtering}
\end{figure}
```

```

\begin{table}[H]
  \centering
  \begin{tabular}{c | c | c }
    filter & MSE & PSNR \\
    none & 72.0 & 29.6 \\
    3x3 mean filter & 32.9 & 33.0 \\
    5x5 mean filter & 36.7 & 32.5 \\
    3x3 median filter & 37.2 & 32.4 \\
    5x5 median filter & 36.2 & 32.5 \\
  \end{tabular}
  \caption{SNR and PSNR after filtering the noisy image}
\end{table}

```

The **image** with the smallest MSE and highest PSNR is the one produces by the 3x3 **mean filter**.

```

\subsection{Use various filters to remove salt and pepper noise with
.05 density from 'DSCN0479-001.JPG'.}

```

Similar to the previous results, the bigger filters make the **image** blurrier and remove **more** noise. The **median** filters perform much better than the **mean** filters on salt and pepper noise. The best looking **image** is the one produced by the 3x3 **median filter**, since most of the noise is gone and the **image** is sharper than the one produced by the 5x5 **median filter**.

```

\begin{figure}[H]
  \includegraphics[width=\linewidth]{Q3/partB.png}
  \caption{images after filtering}
\end{figure}

```

```

\begin{table}[H]
  \centering
  \begin{tabular}{c | c | c }
    filter & MSE & PSNR \\
    none & 6.5 & 40.0 \\
    3x3 mean filter & 42.4 & 31.9 \\
    5x5 mean filter & 44.9 & 31.6 \\
    3x3 median filter & 10.7 & 37.8 \\
    5x5 median filter & 24.7 & 34.2 \\
  \end{tabular}
  \caption{SNR and PSNR after filtering the noisy image}
\end{table}

```

The **image** with the smallest MSE and highest PSNR is the unfiltered noisy **image**. The cause of this is probably the density of errors in the filtered images. In other words the filtered images have small errors in every pixel, **while** the unfiltered **image** has large errors in a small number of images. This brings into question the reliability of MSE and PSNR as useful measures to compare the

results of filtered images.

```
\subsection{Use various filters to clean 'DSCN0482-001.JPG' a noisy  
  version of 'DSCN0479-001.JPG'.}
```

Just like before, the bigger filters produced blurrier images with less noise. However, the difference much smaller. The **mean** and **median** filters seem to produce very similar results. This makes the trade-off in performance undesirable. The 3x3 **median filter** takes 8 times longer than the 3x3 **mean filter** to produce an **image**, **while** the 5x5 **median filter** takes 16 times longer than the 5x5 **mean filter** to produce an **image**.

```
\begin{figure}[H]  
  \includegraphics[width=\linewidth]{Q3/partC.png}  
  \caption{images after filtering}  
\end{figure}
```

```
\newpage  
\section{Code}  
\subsection{Q1 code}  
histEqualize.m  
\lstinputlisting[language=Octave]{Q1/histEqualize.m}
```

```
Q1PartB.m  
\lstinputlisting[language=Octave]{Q1/Q1PartB.m}
```

```
Q1PartC.m  
\lstinputlisting[language=Octave]{Q1/Q1PartC.m}
```

```
\newpage  
\subsection{Q2 code}  
thresholdFaces.m  
\lstinputlisting[language=Octave]{Q2/thresholdFaces.m}
```

```
faceDetect.m  
\lstinputlisting[language=Octave]{Q2/faceDetect.m}
```

```
\newpage  
\subsection{Q3 code}  
Q3.m  
\lstinputlisting[language=Octave]{Q3/Q3.m}
```

```
PartA.m  
\lstinputlisting[language=Octave]{Q3/PartA.m}
```

```
PartB.m  
\lstinputlisting[language=Octave]{Q3/PartB.m}
```

```
PartC.m  
\lstinputlisting[language=Octave]{Q3/PartC.m}
```

```
\newpage
\subsection{latex}
\lstinputlisting{hw1.tex}
\end{document}
```