

```
diary on
format compact
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%EEL3135 Fall 2018
%Lab 4 Part 2
```

```
%2.1
```

```
%Plot the vectors hold1 to verify that it is indeed a zero-order hold
%derived from row.
```

```
%Code Given
row = (-2).^(0:6);
L = length(row);
nn= ceil((0.999:1:4*L)/4); %<---Round up to the integer
hold1 = row(nn);
```

```
%Question: What values are in the indexing vector nn, and why are they what
%they are?
%The values in the indexing vector nn is the length of the row, from 1 to 7.
%The length values index is called four times per a row value, by doing
%so, a consistence reconstruction of the vector is created in each section.
%Every point in the decimated array is duplicated in the row and column
%directions without change thus creating a square pulse.
```

```
type Interpolate
```

```
%Interpolate Interpolate a Lighthouse
%Script for 2.1.
```

```
%2.1.2.1
```

```
%Load the lighthouse image data from lighthouse.mat
load('lighthouse.mat')
```

```
%2.1.2.2
```

```
%Down-sample the lighthouse image by a factor of 3, call the new array x3.
x3 = xx(1:3:end,1:3:end);
```

```
%2.1.2.3
```

```
%Perform a zero-order hold on x3 to fill in the missing points:
%For an interpolation factor of 3, process all rows of x3 to fill in
%missing points in that direction. Call the result xrows.
```

```
row=size(x3,1);
col=size(x3,2);
nn= ceil((0.999:1:3*row)/3);
xrow=x3(nn,1:col);
```

```
%Dimension of xrow.
size(xrow)
```

```
ans =
```

```
327 142
```

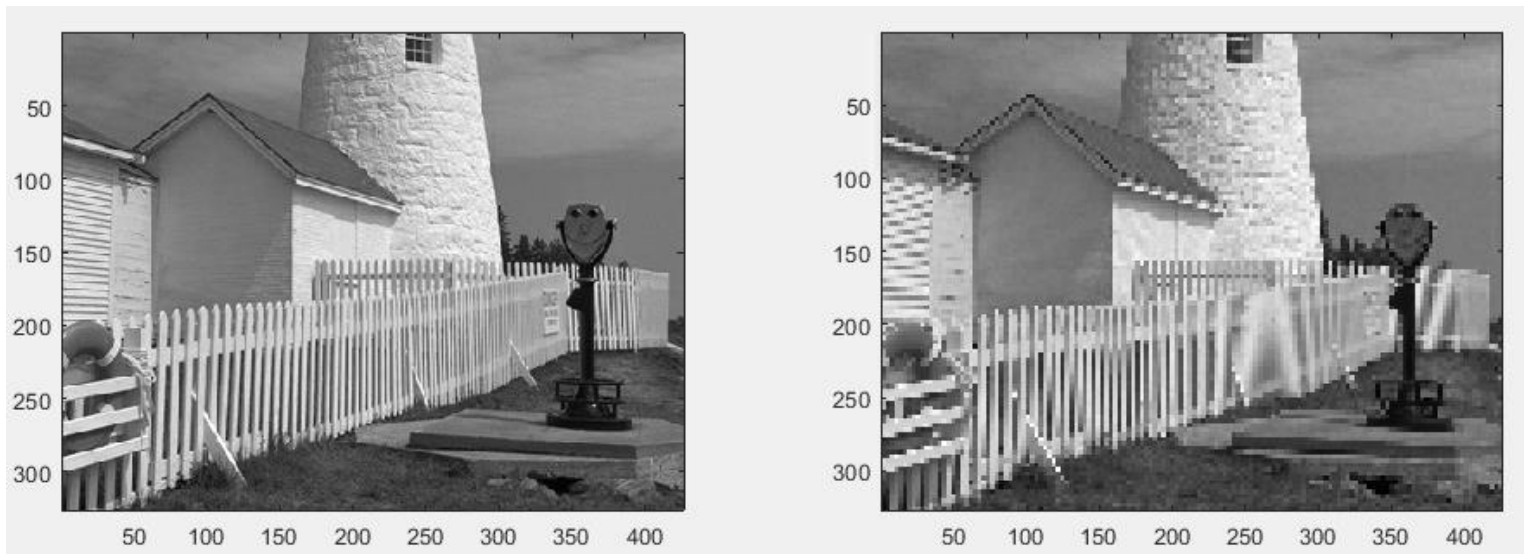
```
%Questions: What are the dimensions of the xrows two-dimensional array?  
%The dimensions of the xrows two-dimensional array is 327x142.
```

```
%2.1.2.4
```

```
%Process all the columns of xrows likewise (interpolation factor 3) to fill  
% in the missing points in each column. Call this result xhold.
```

```
row=size(x3,1);  
col=size(x3,2);  
mm= ceil((0.999:1:3*col)/3.);  
xhold=xrow(1:row*3,mm);
```

```
%Show the xhold (right) and original lighthouse (left) images on the same  
%plot.
```



```
%Compare them and explain any differences that you can see.  
%The original image (left) is very clear and detailed while the  
%reconstructed image, xhold (right), is blurry and pixelated due to the  
%repetitions.
```

```
%2.1.2.5
```

```
%Carry out linear interpolation operations on both the rows and the columns  
%of the down-sampled lighthouse image x3.
```

```
%Previous values/code.  
load('lighthouse.mat')  
x3 = xx(1:3:end,1:3:end);  
row=size(x3,1);  
col=size(x3,2);
```

```
lrow=1:row;
```

```

lcol=1:col;
%Ex:tt = 0:0.1:6; % <---locations between the n indices
ttr=1:1/3:length(lrow);
ttc=1:1/3:length(lcol);
%Apply linear interpolation operations.
rowlinear=interp1(lrow,x3,ttr);
%Transpose to fit dimension.
rowlineart=transpose(rowlinear);
%Apply linear interpolation operations.
holdlinear=interp1(lcol,rowlineart,ttc);
%Re-transpose to original dimension.
holdlineart=transpose(holdlinear);
show_img(holdlineart,0,1,'gray(255)');

```



%2.1.2.6

Show the original image, the down-sampled image, the zero-order-hold  
 %reconstructed image, and the linearly-interpolated reconstructed image in  
 %that order.



%Point out regions where the linear and zero-order reconstruction result  
 %images differ and try to justify this difference.  
 %In low frequency areas, such as the background and lighthouse, the linear  
 %interpolation reconstructed image looks better than the zero-order hold  
 %reconstructed image as it is more smoother and less pixelated. In high  
 %frequency areas, such as the fence due its sinusoidal repetition pattern,  
 %the linear interpolation reconstructed image is blurrier while the zero  
 %-order hold reconstructed image is more pixelated.

%Zero-order hold reconstruction, repeats certain parts in the image to fill  
 %in the missing portions resulting in a sharp transition between one set of  
 %values to another that forms defined borders, creating a pixelized image.  
 %In low frequency areas this pixelization is more noticeable as there is  
 %more variations in image while it is less noticeable in high frequency  
 %areas as there is less variations, long line of similar fences.

%Linear interpolation reconstruction, averages out the differences in the  
 %gap from the missing portions to fill it in, creating a smoother change.  
 %In low frequency areas there is a smooth transition between different  
 %pieces of the image and the color shift has improved from the zero-order  
 %hold reconstruction. In the high frequency areas, the image become blurrier  
 %because of the numerous gaps which has only slight deviations meaning the  
 %average is less definite.

```
%Question: Are edges low-frequency or high-frequency features?  
%The edges are low-frequency features.
```

```
%Question: Is the series of fence posts a low-frequency or high-frequency  
%feature?  
%The series of fence posts are a high-frequency feature.
```

```
%Question: Is the background a low-frequency or high-frequency feature?  
%The background is a low-frequency feature.
```

```
type Restoration
```

```
%Restoration Restoration Filter for 1-D Data  
%Script for 2.2
```

```
%2.2.1
```

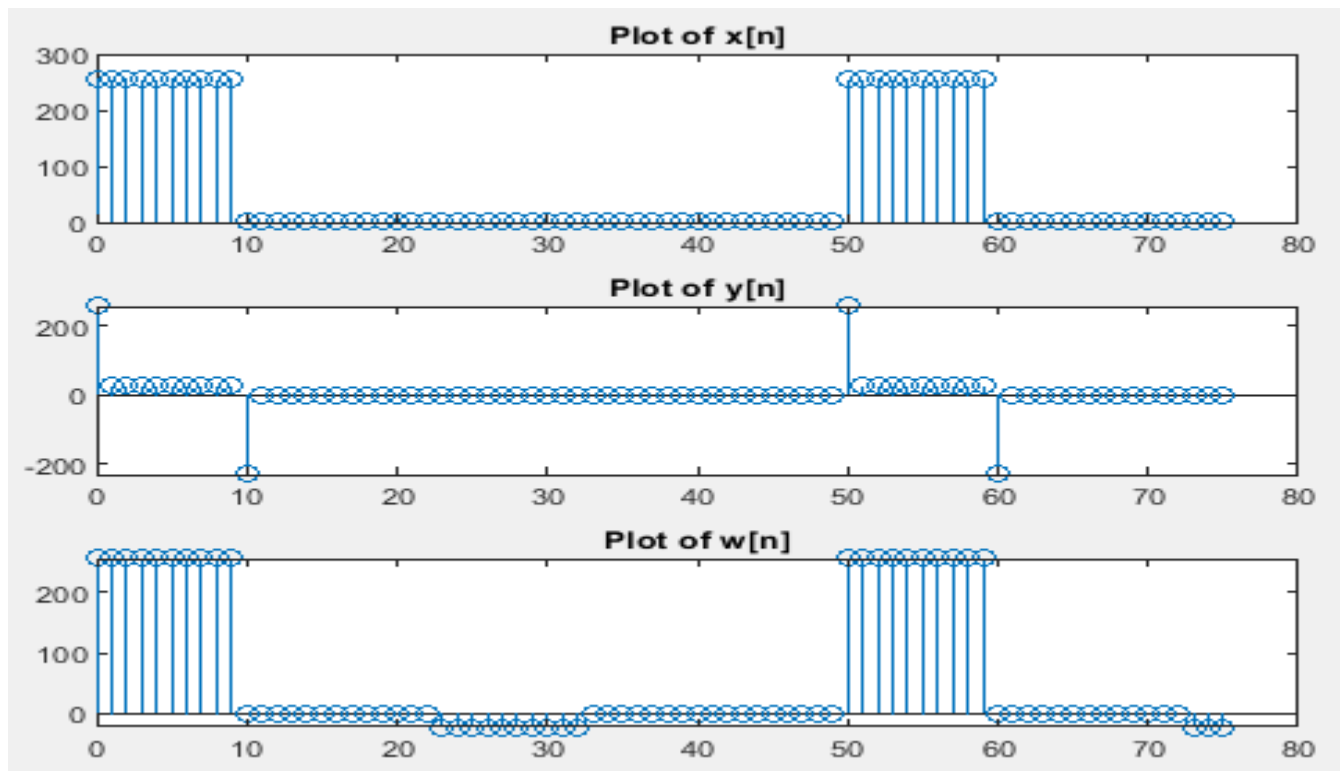
```
%Implement an FIR filter on the following input signal:  
%The function  $w[n] = x[n] - 0.9x[n-1]$  is an FIR filter.  
%The convolution range is 1 to -0.9.  
bb=[1,-0.9];  
xx = 256*(rem(0:100,50)<10);  
yy=firfilt(bb,xx);
```

```
%2.2.2
```

```
%Process the filtered signal  $w[n]$  from the previous section with this  
%restoration filter. Use  $r=0.9$  and  $M=22$ .  
r=0.9;  
M=22;  
%Value from r and M  
rm=0.9.^(0:M);  
%yy taken from part 2.2.1.  
ww=firfilt(rm,yy);
```

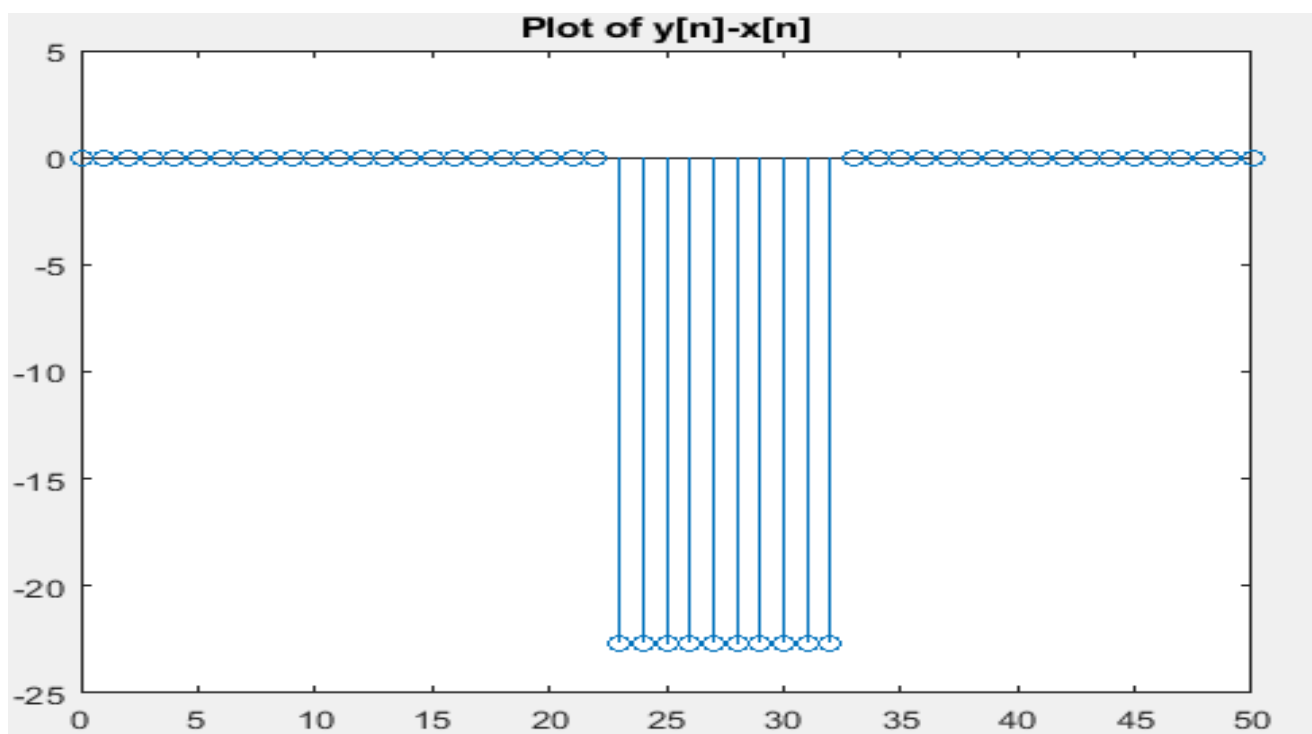
```
%2.2.3
```

```
%Plot  $x[n]$ ,  $w[n]$ ,  $y[n]$  and on the same figure.  
%Make the discrete-time signal plots with stem plots but restrict the  
%horizontal axis to the range  $0 \leq n \leq 75$ .  
%xx plot  
subplot(3,1,1);  
stem(0:75,xx(1:76));  
title('Plot of  $x[n]$ ');  
%yy plot  
subplot(3,1,2);  
stem(0:75,yy(1:76));  
title('Plot of  $y[n]$ ');  
%ww plot  
subplot(3,1,3);  
stem(0:75,ww(1:76));  
title('Plot of  $w[n]$ ');
```



%2.2.4

```
%Make a plot of the error (difference) between y[n] and x[n] over the range
%0<=n<50.
diff=yy(1:51)-xx(1:51);
stem(0:50,diff);
title('Plot of y[n]-x[n]');
```



%2.2.5

%The maximum difference between the above y[n] and x[n] over the range  
%0<=n<50.

%This is the worst-case error.

max(abs(diff))

ans =

22.6891

%Max value is rounded to 22.7.

%2.2.6

%What does the error plot and worst-case error tell you about the quality  
%of the restoration of x[n]?

%Since the error plots of x[n] and w[n] are very similar, the restoration  
%of x[n] is quite accurate but the worst-case error indicates that there  
%are slight deviations in the values from the original.

type Filtering

%Filtering Filtering Images with One-dimensional FIR filters.

%Script for 2.3.

%2.3.1.1

%Load in the echart image from the echart.m file.

load('echart.mat')

%Filter all the rows of the image with the conv2() function.

%To filter the image in the horizontal direction using a first-difference  
%filter.

%Code given.

b = [1, -1];

filtrow=conv2(echart,b);

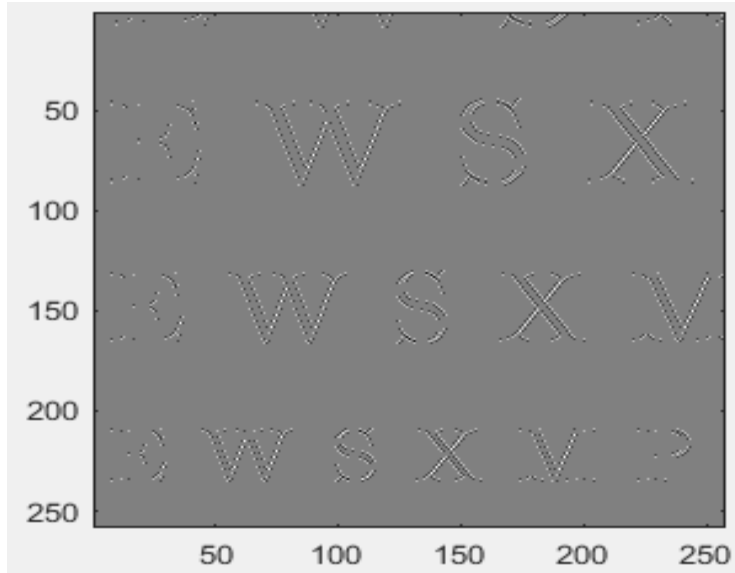
%2.3.1.2

%Filter the image in the vertical direction with this first-difference  
%filter to produce the image y. This can be done by transposing the image  
%then filtering the rows.

bt=transpose(b);

yy=conv2(filtrow,bt);

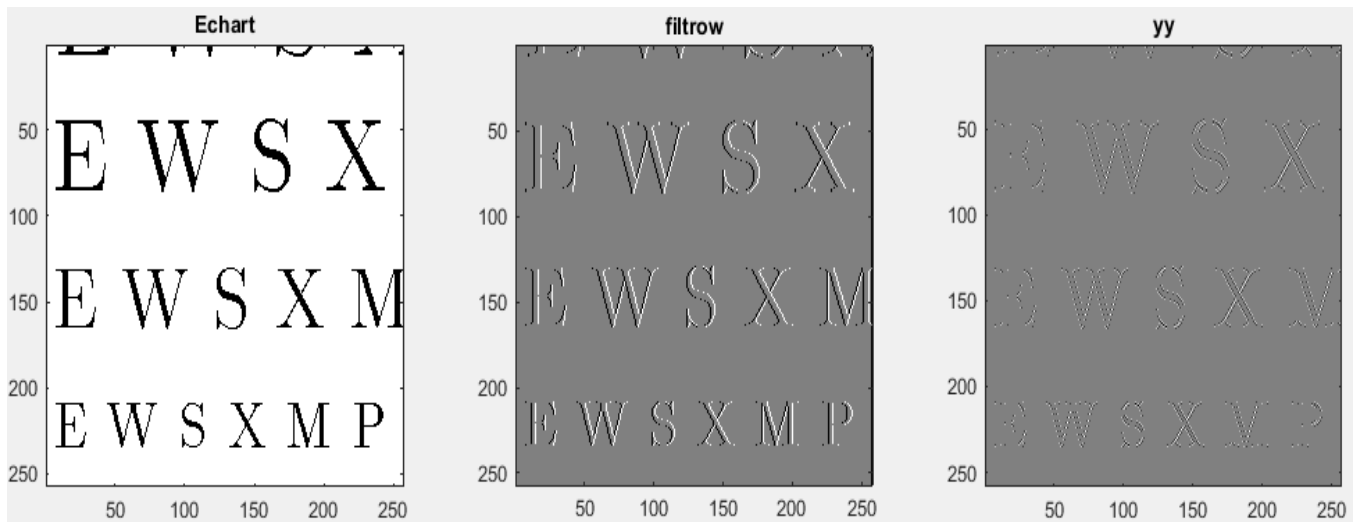
show\_img(yy,0,1,'gray(256)');



%2.3.1.3

%Display the input image echart, the intermediate image row-filtered, and  
%the output image y.

```
%Plot echart.
subplot(1,3,1);
show_img(echart,0,1,'gray(256)');
title('Echart');
%Plot Filtrow.
subplot(1,3,2);
show_img(filtrow,0,1,'gray(256)');
title('filtrow');
%Plot yy.
subplot(1,3,3);
show_img(yy,0,1,'gray(256)');
title('yy');
```





```
%Compare the three images and give a qualitative description of what you
%see.
%The original image, echart, is a very clear eye chart image with a strong
%contrast with the white background and black text. The image with the rows
%filter, filtrow, became grayed out image of the original which makes it
%difficult to read the text due to the lack of contrast with the text and
%background. The image with both the rows and columns filter, yy, is a
%very dark image from the original, which makes it super difficult to read
%the text as there is barely any contrast with the text and background to
%make out the text.
```

#### %2.3.2.1

```
%Load the echart image from the echart.m module. Pick q=0.9 for the first
%filter (FIR 1). Call the result echo90.
```

```
load('echart.mat')
```

```
q=0.9;
```

```
%Vector form.
```

```
qv=[1,-q];
```

```
%Using this filter, filter the echart image along the horizontal direction.
```

```
echo90a=conv2(echart,qv);
```

```
%Transpose the vector.
```

```
qvt=transpose(qv);
```

```
%Filter the resulting image vertically.
```

```
echo90=conv2(echo90a,qvt);
```

#### %2.3.2.2

```
%Convolve echo90 with FIR 2, choosing M=22 and r=0.9.
```

```
M=22;
```

```
r=0.9;
```

```
%Vector form.
```

```
mv=0:1:M;
```

```
%Produce  $r^{(1)}$ 
```

```
rm=r.^mv;
```

```
%Convolve back to original.
```

```
deconv1=conv2(echo90,rm);
```

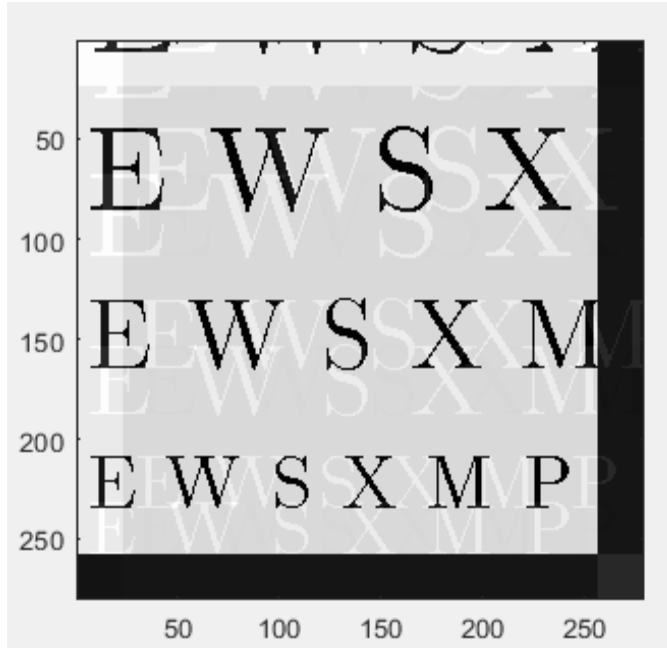
```
%Transpose the vector.
```

```
rmt=transpose(rm);
```

```
deconv2=conv2(deconv1,rmt);
```

```
%Display image.
```

```
show_img(deconv2,0,1,'gray(256)');
```



%Describe the visual appearance of the output qualitatively, showing the  
 %image, and explain its features by invoking mathematical understanding  
 %of the cascade filtering process and why you see “ghosts” in the output  
 %image.

%The visual output image has relatively clear and same size letter as the  
 %original image but also has visual 'echoes'. The visual echoes cause  
 %a repetition of the letters in different position known as ghosts as they  
 %are more faded and less visible. The interpolation of the image determines  
 %the position and color of the ghost where the first interpolation led to  
 %the shift in the horizontal axis, moving it to the right, while the second  
 %interpolation led to the shift in the vertical axis, moving it down. The  
 %cascading of interpolation results in overlaying shades with the respected  
 %shift of the image as since the ghost images heads downward right, the  
 %upper left area is left alone since there is no overlapping filter but the  
 %bottom right the two-overlapping filter therefore it gets increasingly  
 %darker.

%Determine the size and location of the “ghosts” relative to the original  
 %image.

%The size of the ghosts is the same as the original image, thus size of  
 %280x279 pixels.

%The location of the first ghosts, horizontal shift, is shifted 23 pixels  
 %to the right from the original. The location of the second ghosts,  
 %vertical shift, is shifted 23 pixels down from the original.

%2.3.2.3

%Evaluate the worst-case error.

```
err=echart(1:255)-deconv2(1:255);  
max(abs(err))
```

ans =

22.6005

%Worst-case error=22.6005.

%2.3.3.1

%Now try to deconvolve echo90 with several different FIR filters for FIR 2.  
%Set  $r=0.9$  and try several values for  $M$ , at the least  $M=11,22,33$ .

```
rr=0.9;
MM=11; %MM=22; %MM=33;
%Vector form.
mmv=0:1:MM;
%Produce  $r^{(1)}$ 
rmm=r.^mmv;
%Convolve back to original.
deconv1=conv2(echo90,rmm);
%Transpose the vector.
rmmt=transpose(rmm);
deconv2=conv2(deconv1,rmmt);
%Display image.
show_img(deconv2,0,1,'gray(256)');
```

%Pick the best result and explain why it is the best. Describe the visual appearance of the output qualitatively, and explain its features by invoking mathematical understanding of the cascade filtering process.  
%The best result came from  $M=33$  as its reconstruction resulted in the most clear and detailed image, the effect of the echoes was lessened and there was a more definite contrast with the background and letters.  
%This is because the reconstruction goes through more values as  $M$  is higher which enables a more accurate result. The cascading of images results in the ghosts in the image but since the ghosts are barely visible, the shift from the cascade is almost a full phase from the original which would result in producing an original image. However, increasing  $M$  also increases the size of the darken portion of the image.

%2.3.3.2

%Question: How large is the worst-case error (from the previous part) in terms of number of gray levels?

```
%Evaluate the worst-case error.
err=echart(1:255)-deconv2(1:255);
max(abs(err))
```

ans =

7.0923

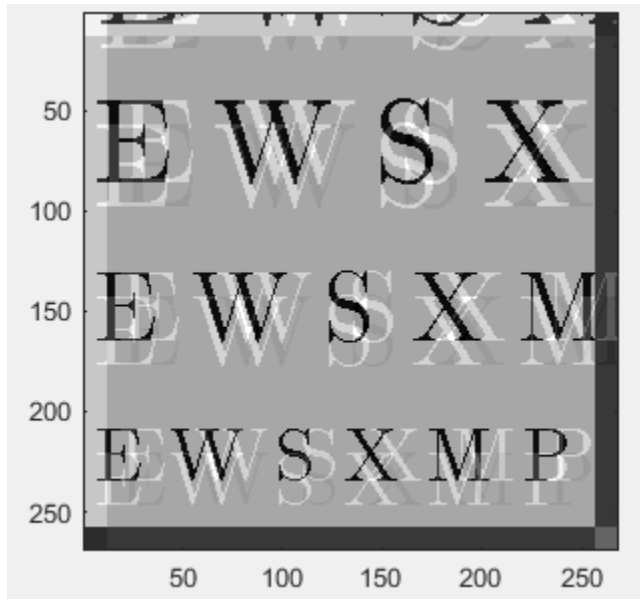
%The worst-case error is about 7.09.

%Question: Evaluate worst-case error for each of the three filters in the previous section.

```
%M=11;
err=echart(1:255)-deconv2(1:255);
max(abs(err))
ans =
```

72.0195

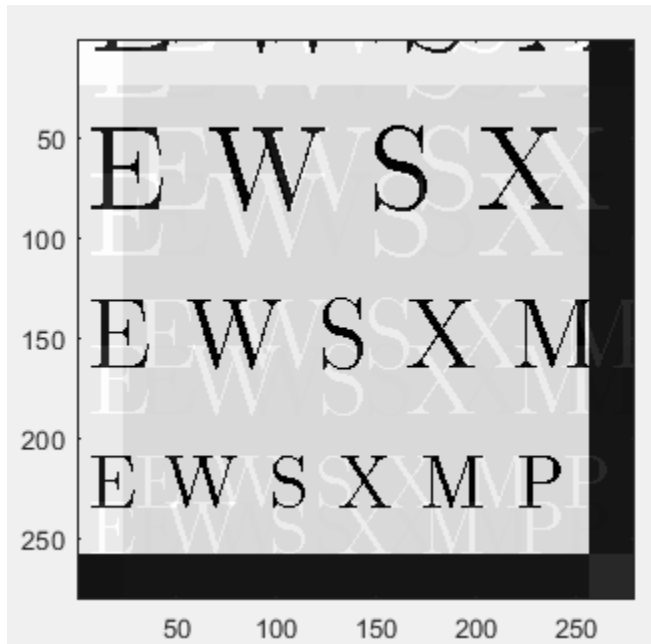
%The worst-case error is about 72.02.



```
%M=22;  
err=echart(1:255)-deconv2(1:255);  
max(abs(err))  
ans =
```

22.6005

%The worst-case error is about 22.60.

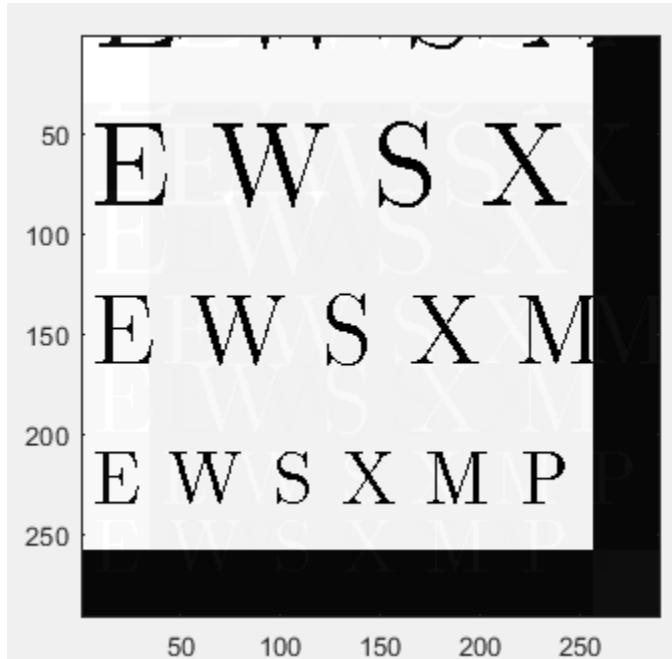


```
%M=33;  
err=echart(1:255)-deconv2(1:255);  
max(abs(err))
```

```
ans =
```

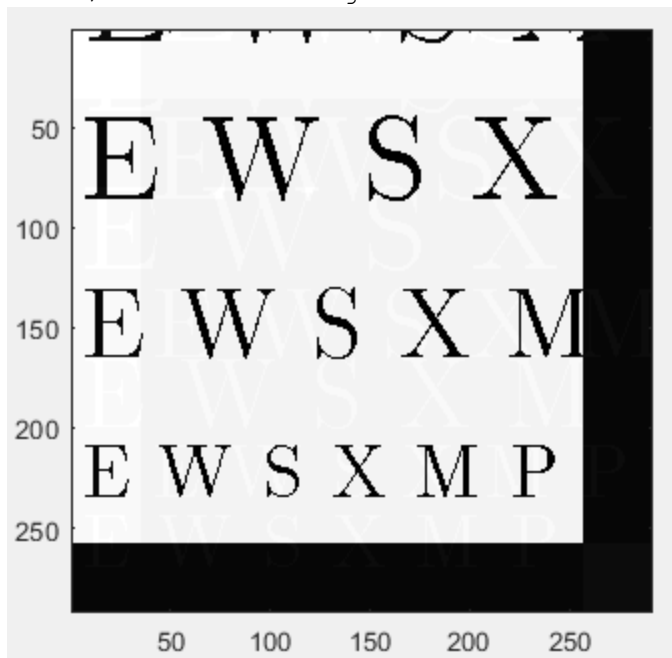
```
7.0923
```

```
%The worst-case error is about 7.09.
```



```
%Question: Can your eyes perceive a gray scale change of one level, one  
%part in 256?
```

```
%M=34; One level change.
```



```
%My eyes cannot perceive a gray scale change of one level as the plots  
%looks identical.
```

```
diary off
```