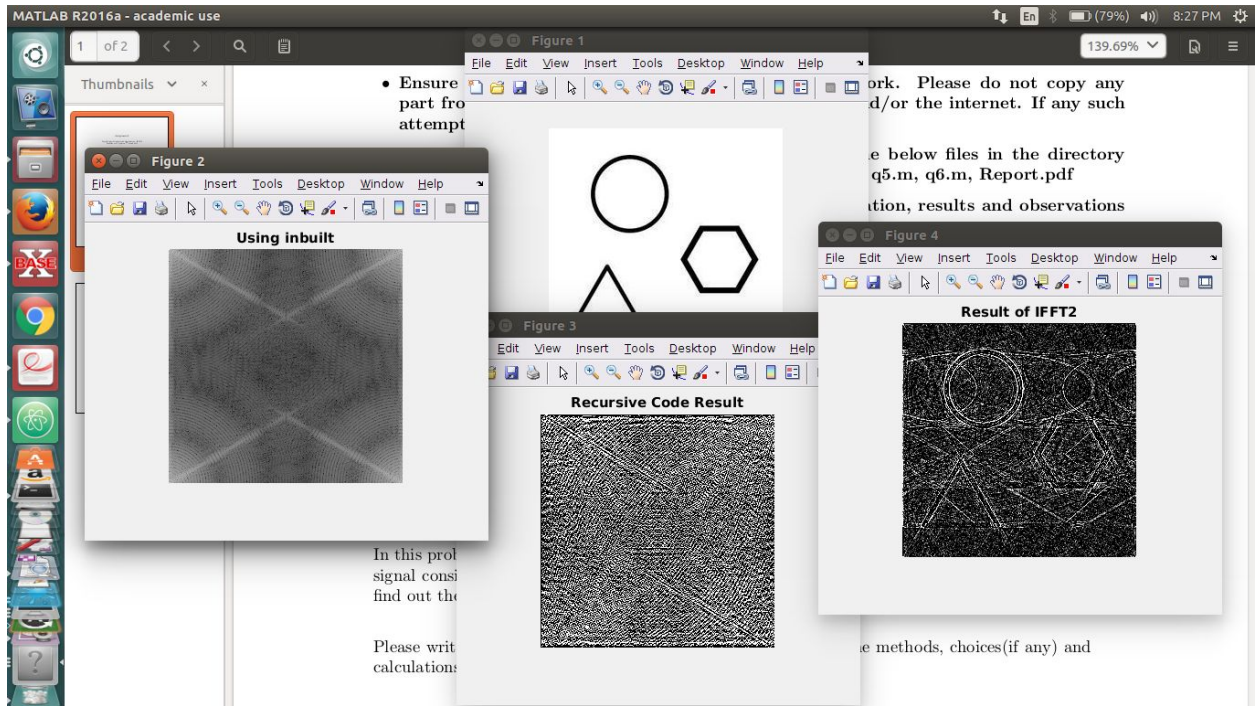


DSAA Assignment 2

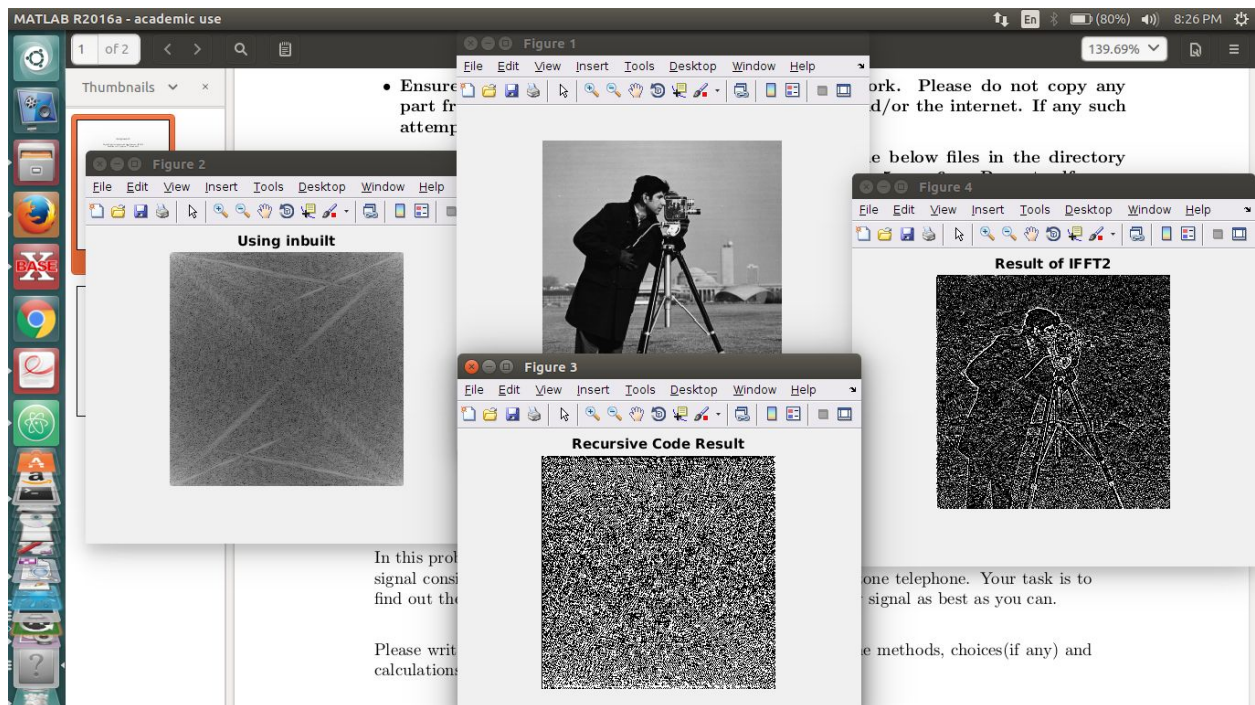
Abhishek Nalla

20161115

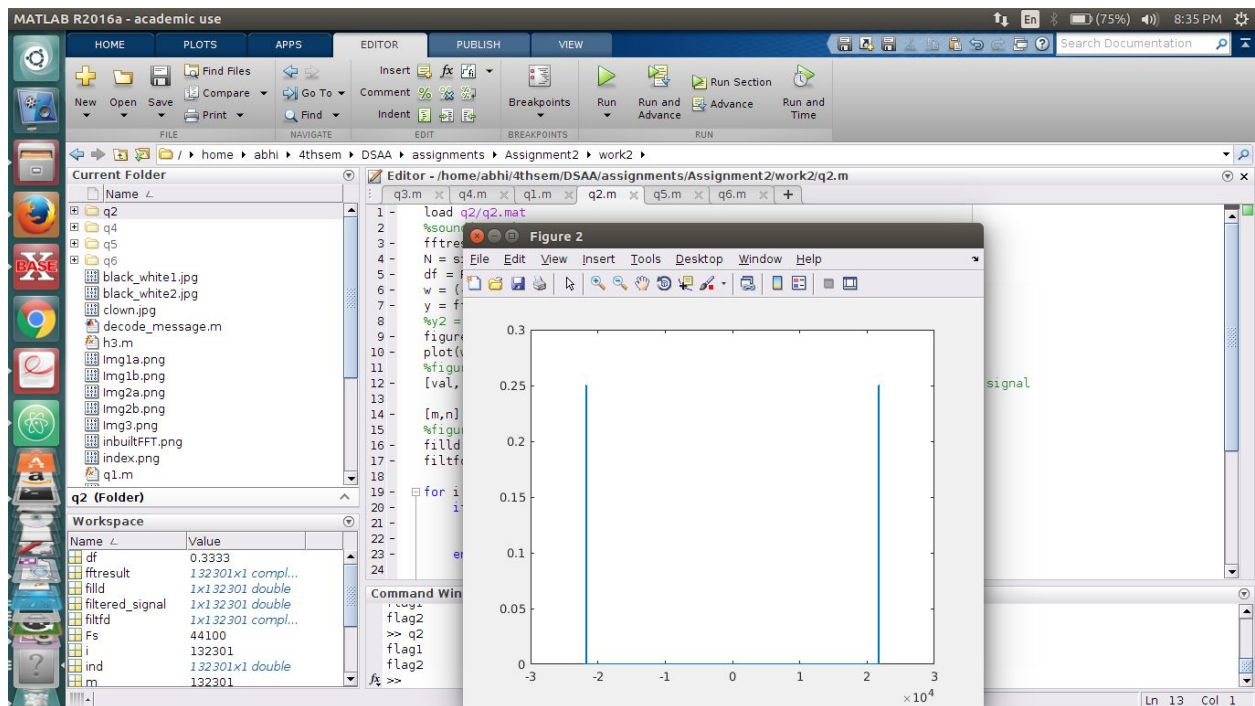
1. On a 2D black and white image:



On 'cameraman.tif':



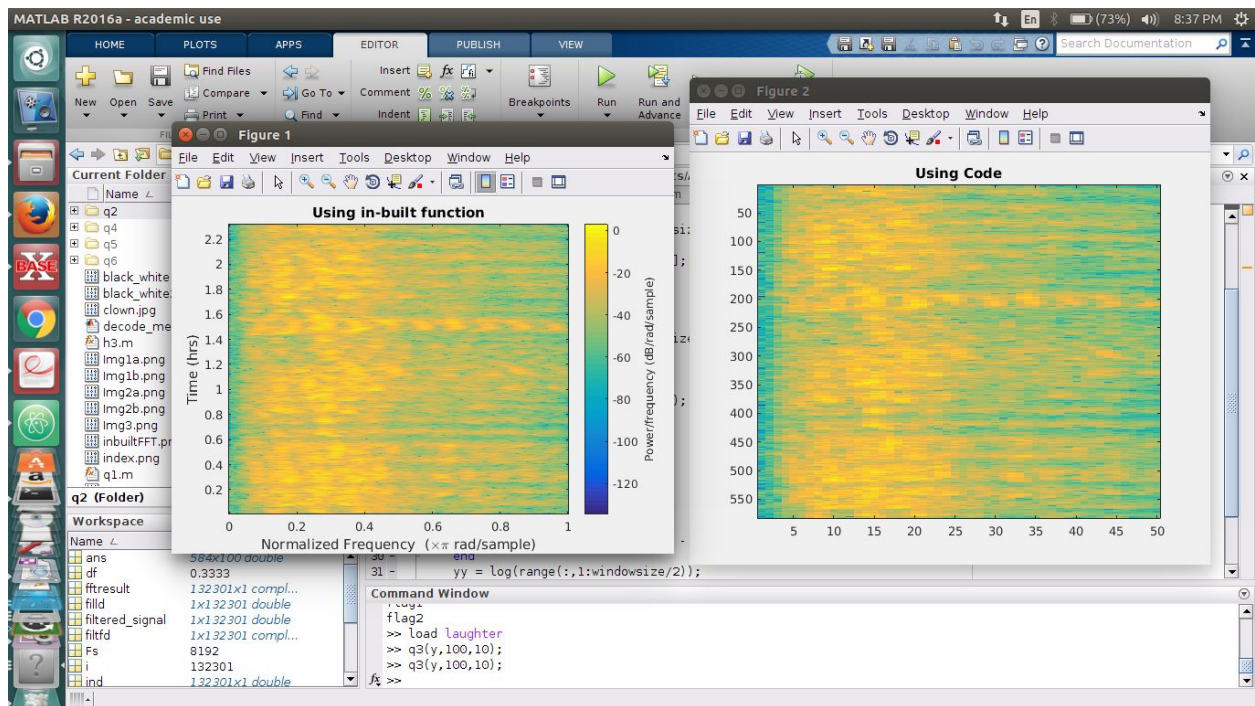
2. Frequencies detected:



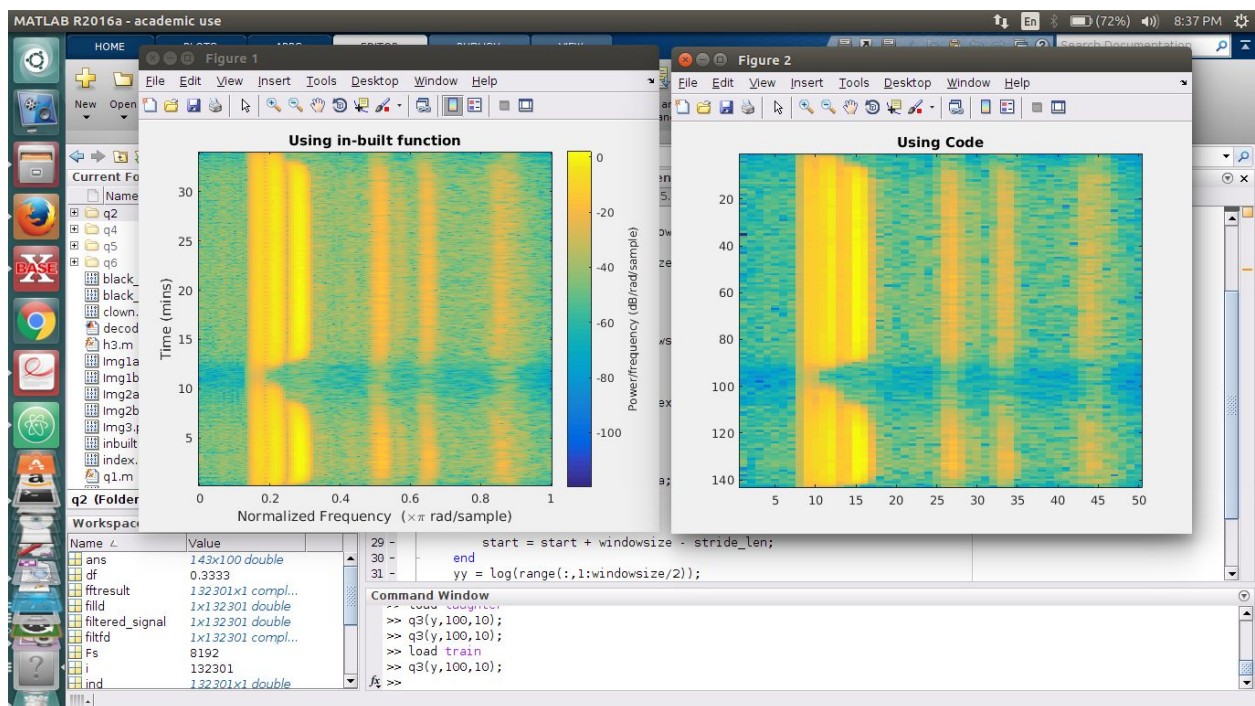
A telephone tone has 2 dominant frequencies and other frequencies which are considered noise.

What was done to filter out the noise and to get the frequencies of the sound was to do FFT on the input signal to transfer it to frequency domain and then sort the frequencies according to their amplitudes. Now the two highest frequencies would be located at the end of the array. Then I would take the values of the signals at these final 2 indices and perform IFFT on them to get the signal without noise back to time domain. Then I would play the sound with only these 2 frequencies.

3. On loading laughter:



On loading train:

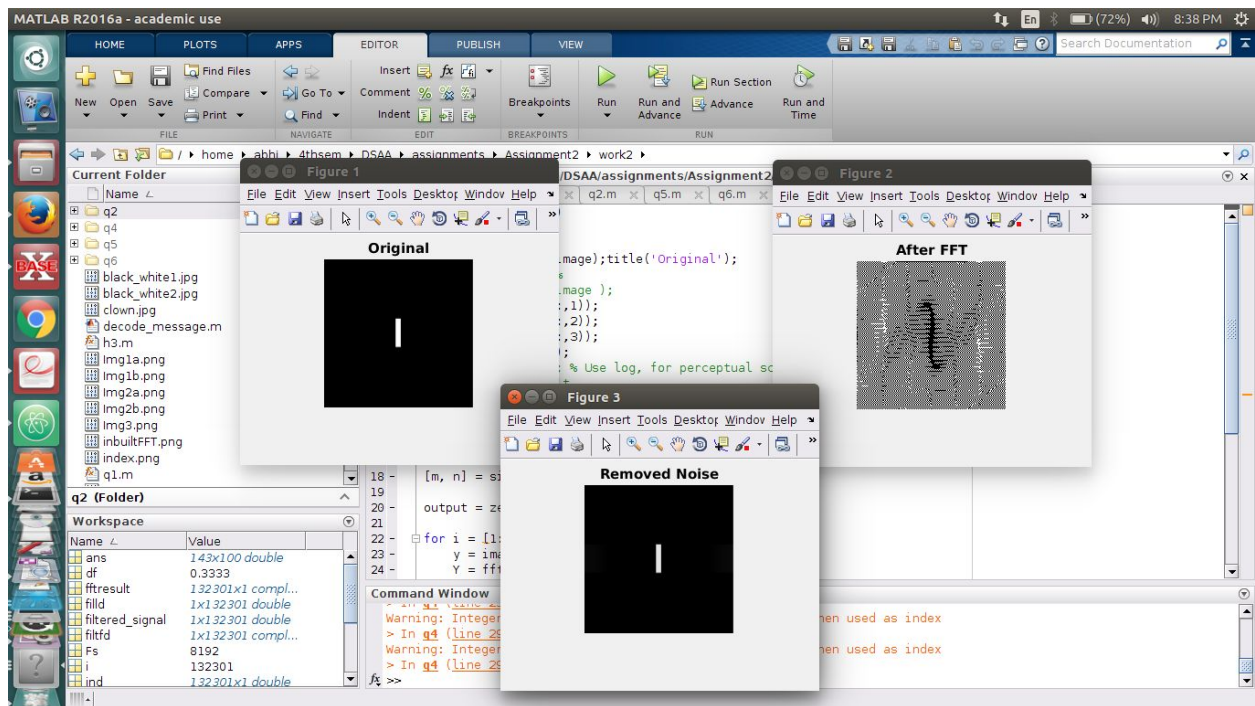


The time-frequency resolution of the spectrogram is dependent upon the chosen window size. A larger size will result in higher spectral, but lower temporal resolution, whereas the opposite will result in a lower spectral, but higher temporal resolution. If your application is such that you need time domain information to be more accurate, reduce the size of your windows. If the application demands frequency domain information to be more specific, then increase the size of the windows.

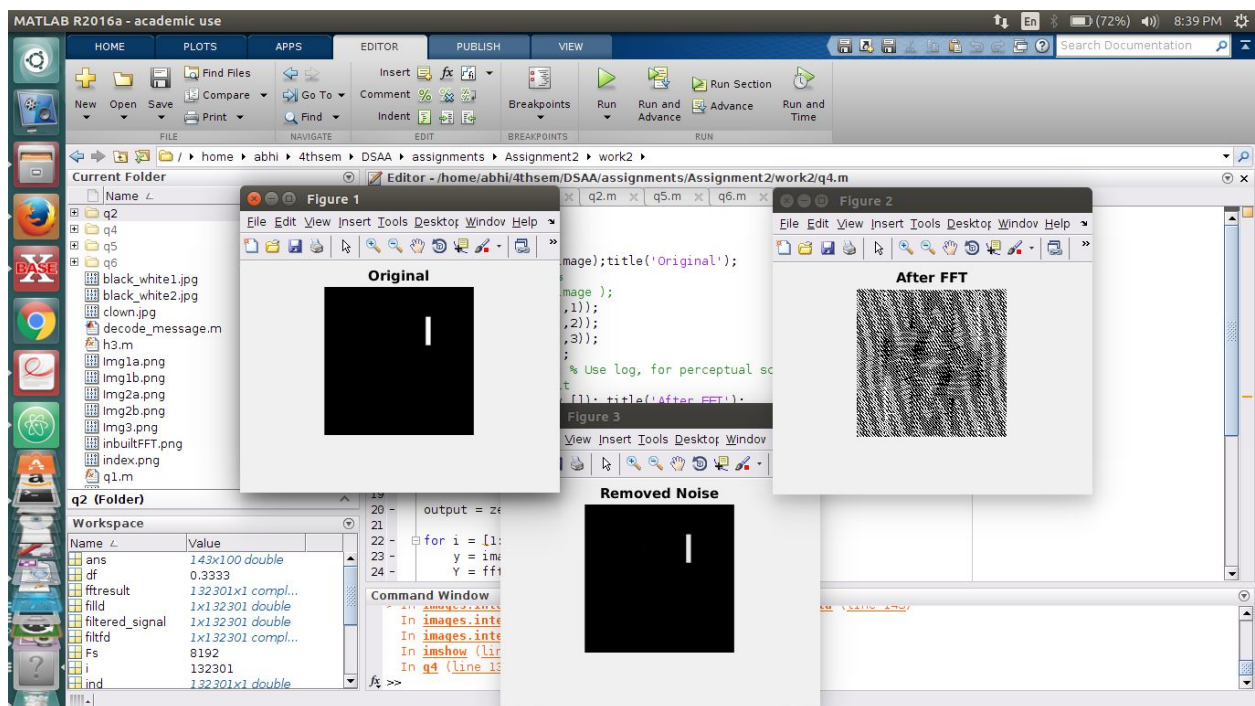
Step size (which is window length minus overlap) controls the horizontal scale of the spectrogram. Decrease it to stretch, or increase it to compress. Increasing step size will reduce time resolution, but decreasing it will not improve it much beyond the limits imposed by the window size (you do gain a little bit, depending on the shape of your window, as the peak of the window slides over peaks in the signal energy). The range 1-5 msec is good for speech.

On comparing the in-built spectrogram and the STFT result it is clear that there is some distortion in the coded one because we are considering each small window and thus compromising on frequency resolution.

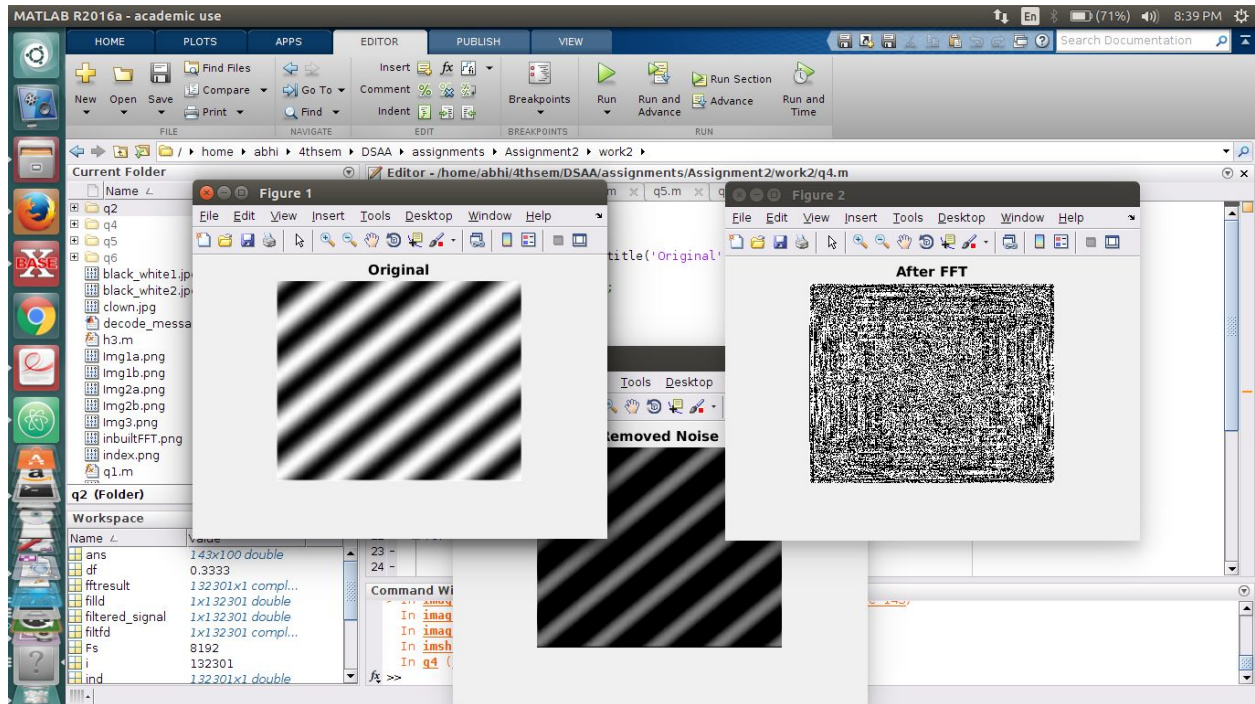
4. On running FFT on Img1a.png



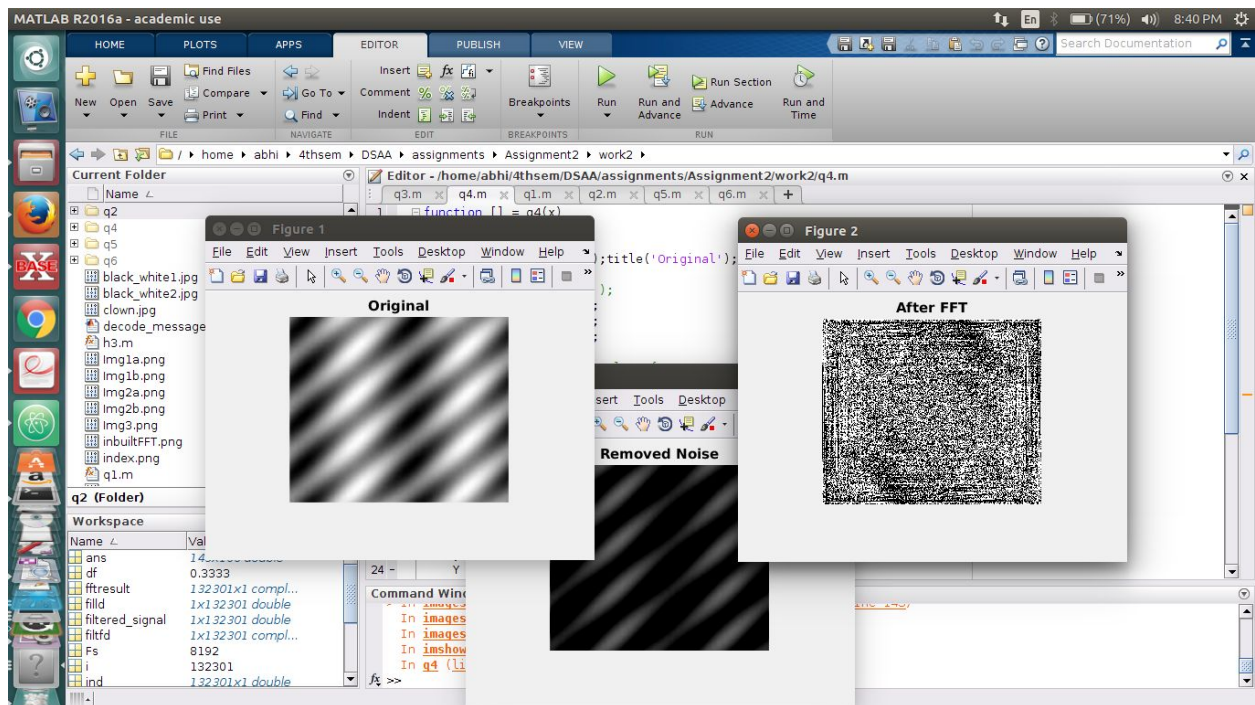
On running FFT on Img1b.png



On running FFT on Img2a



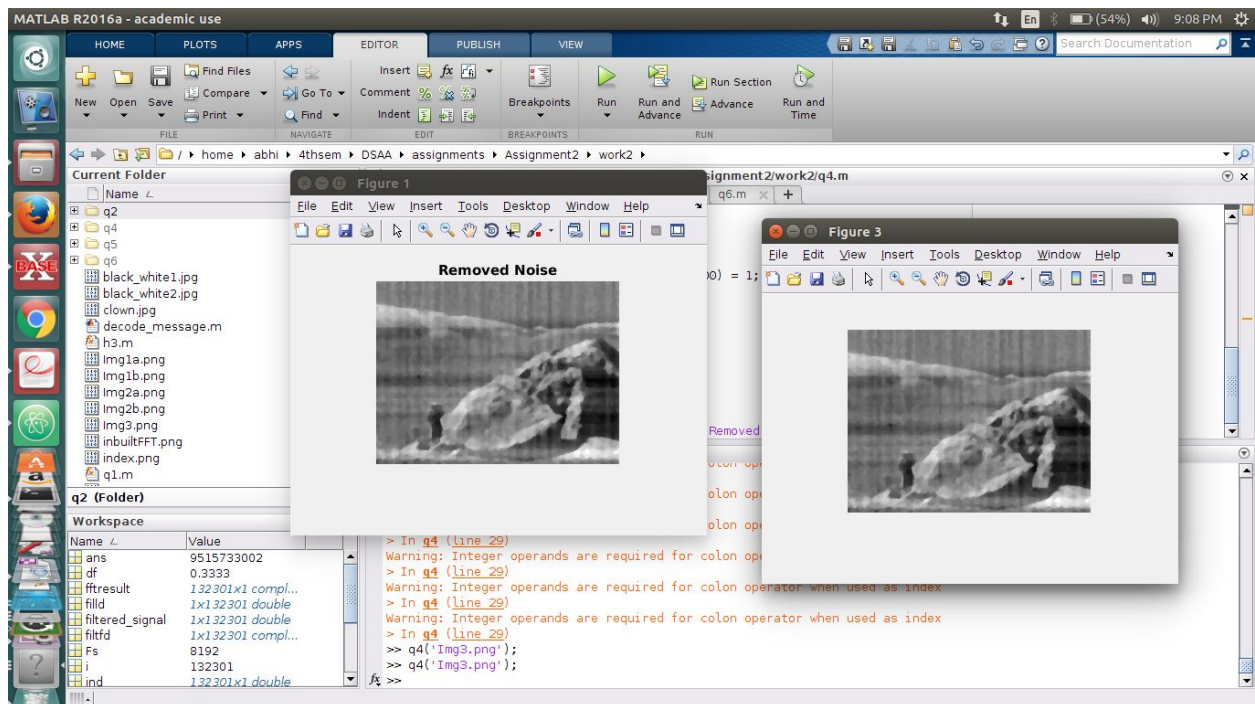
On running FFT on Img2b



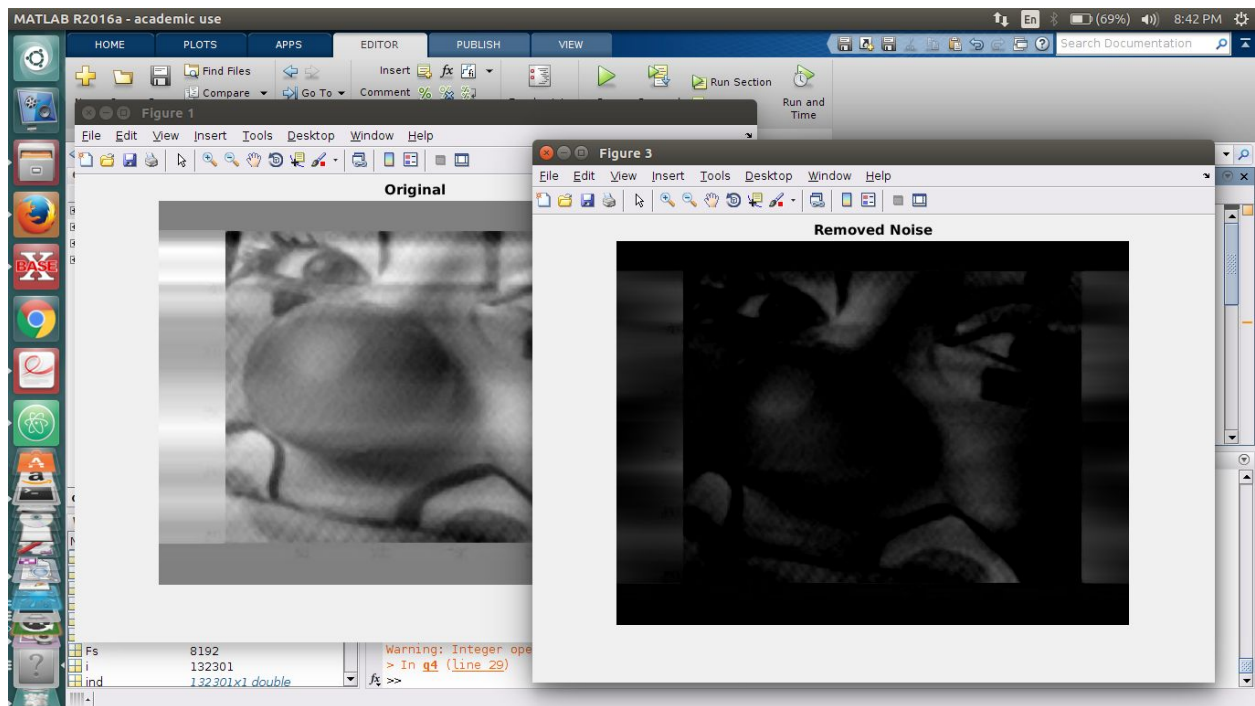
It is clear from doing FFT over the images that the Frequency outputs depend on the position of various frequencies on the image.

Noise can be removed from the images by converting it to Frequency domain, multiplying it with a rect function(point to point) and taking the IFFT of the resultant output. This guarantees smoothening of frequencies over the entire range. The sharp frequencies are preserved while the rest are smoothened by the rect filter.

On removing noise from Img3

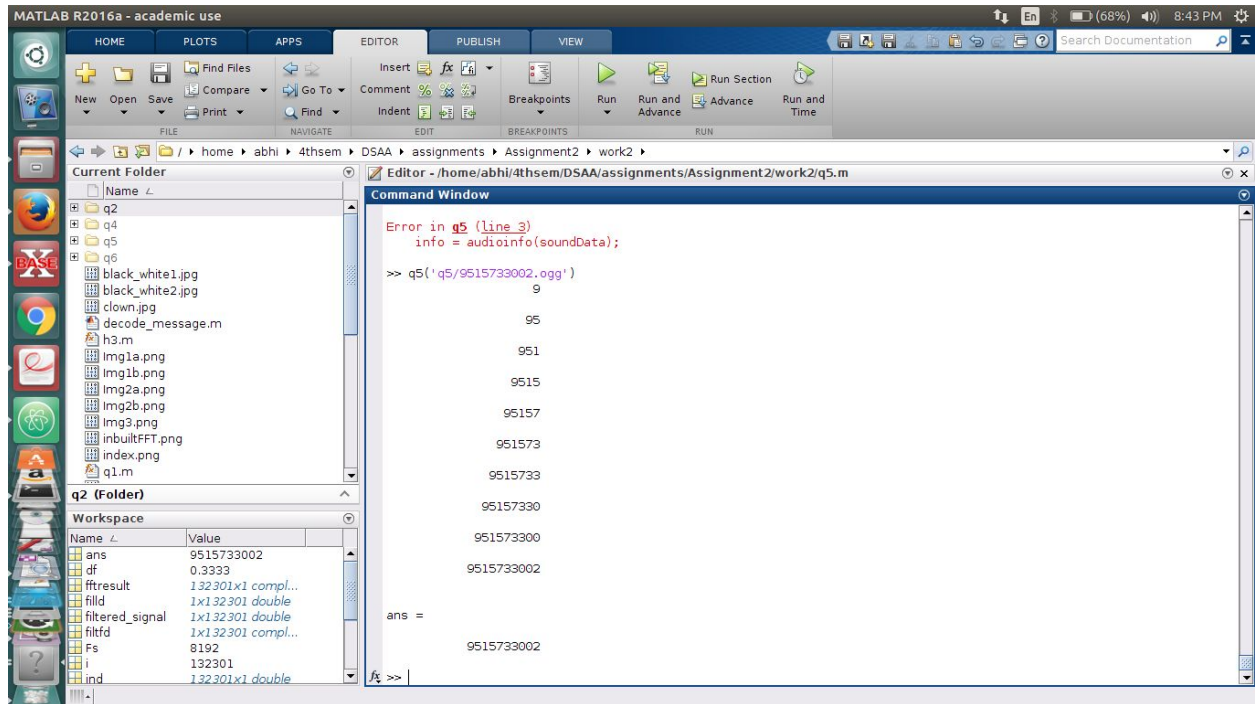


On removing noise from clown



5. Finding the number dialled:

I compare the frequencies dialled with the frequencies of the individual digits and then take a dot product to check which is the most matching over each range.



6. Decoding the encrypted sound signal:

