The background features several decorative elements: a green horizontal bar at the top left, a light blue horizontal bar below it, a light blue vertical bar on the right, a yellow vertical bar on the bottom left, a red vertical bar below the yellow one, and a yellow circle at the bottom right.


Fourier Transform Model of Image Formation

★ OVERVIEW OF THE ACTIVITY ★

The Fourier Transform is a mathematical method that permits the investigation and comprehension of a signal's frequency characteristics. It is often used in image processing to transform images from the spatial domain to the frequency domain.

[1] To calculate the Fourier transform (FT) of a two-dimensional signal $f(x,y)$, we use the formula, $F(f_x, f_y) = \iint f(x,y) \exp(-i2\pi(f_x x + f_y y)) dx dy$ where f_x and f_y are the spatial frequencies along the x and y axes, respectively. [2]



In this activity, however, we don't have to limit ourselves to this specific integral or function, and we are free to investigate the simulations themselves. Let's look into the widespread application of the Fourier Transform (FT) in image processing!

Two horizontal decorative bars are located at the bottom of the slide. The top bar is purple and spans the width of the text area. The bottom bar is red and is positioned further down, also spanning a significant portion of the slide width.



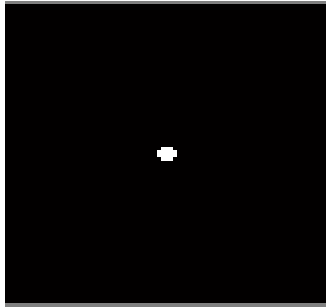
OBJECTIVES



- Understanding the Fourier transform and its application in image processing
 - Transforming images between the spatial and frequency domains
 - Applying FT, FFT shift, and logscale shift to various binary images
 - Comparing the resulting Fourier Transforms for different apertures and aperture diameters
 - Applying FT and inverse FT to grayscale square image
 - Exploring the concepts of Convolution as a simulation of an imaging system and Correlation in template matching
- 
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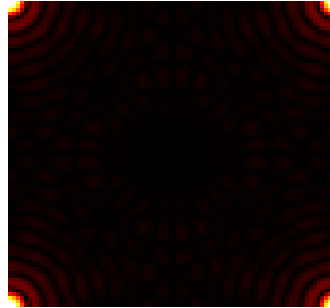
2.1 FAMILIARIZATION WITH DISCRETE FT

Circular aperture



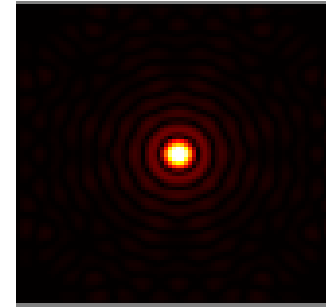
(a)

FT



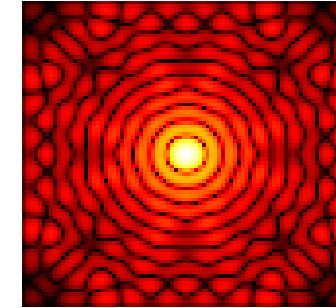
(b)

Shifted FT



(c)

Log Scaled Shifted FT



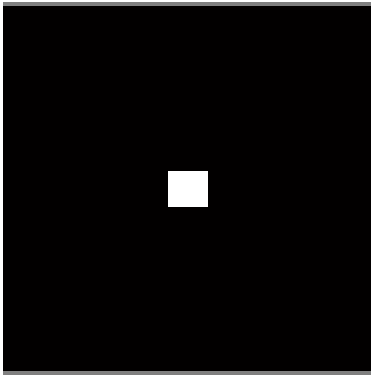
(d)

Figure 1

The first part of the exercise allowed us to become familiar with the Discrete Fourier Transform (DFT) by applying it to a binary image, which in this case is the circular aperture in Figure (a). **After creating this binary picture with MATLAB's built-in functions, we Fourier transformed it with the `fft2` function to convert its spatial domain representation to its frequency domain representation.** However, the transformed image is a complex-valued image, which means that, as shown in figures (b), the high frequency components were concentrated at the image's corners rather than in the image's center. This is because, by default, the output of `fft2` in MATLAB places the zero frequency component at the top left corner, so for easier visualization, **we shifted the zero-frequency component of the Fourier transform to the center of the matrix using the `fftshift` function.** [3] Lastly, we used the `log` and `rescale` functions to improve the visibility of the Fourier spectrum by taking the logarithm of the absolute value of the Fourier transform, and rescaling the values to be on a logarithmic scale. [4] Thus, Figure (d) shows the visual representation of the frequency content of the circular binary image. From here we can see **that the matrix's brighter regions represent higher frequency components, whereas the matrix's darker regions represent lower frequency components.**

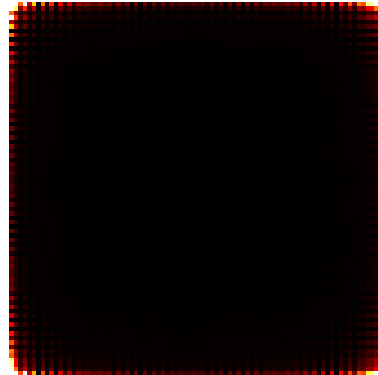
OTHER FT OUTPUTS USING DIFFERENT SHAPES OF APERTURE

Square aperture



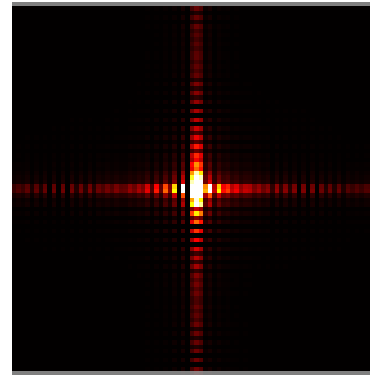
(a)

FT



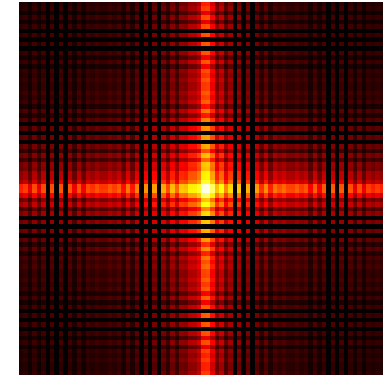
(b)

Shifted FT



(c)

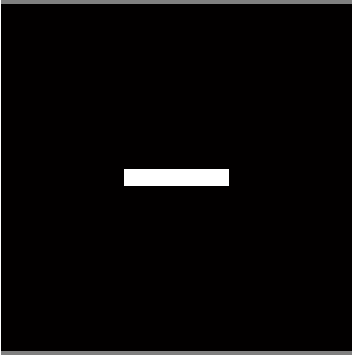
Log Scaled Shifted FT



(d)

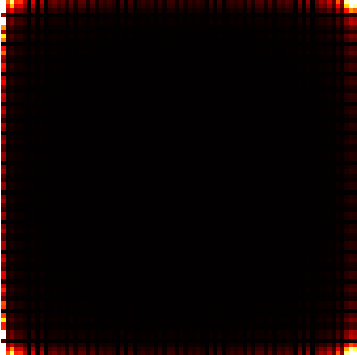
Figure 2

Rectangular aperture



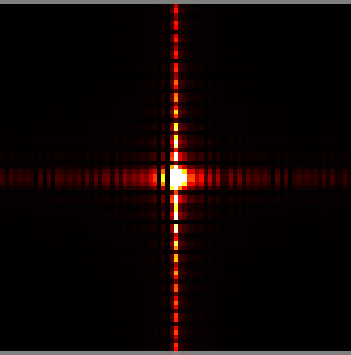
(a)

FT



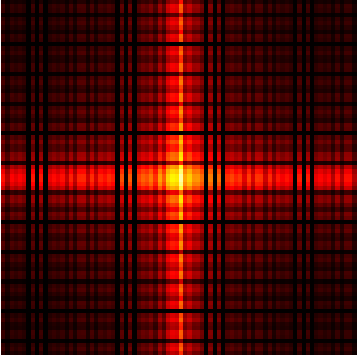
(b)

Shifted FT



(c)

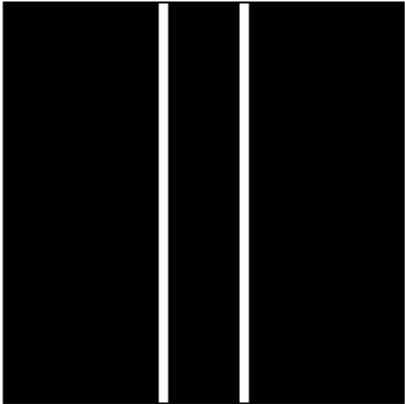
Log Scaled Shifted FT



(d)

Figure 3

double slit aperture



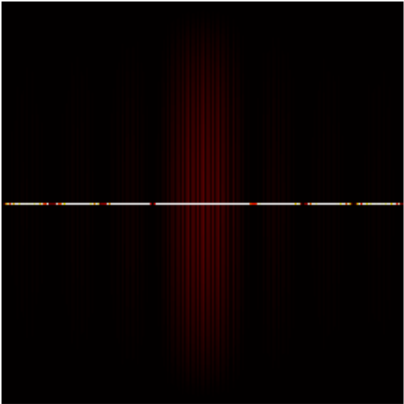
(a)

FT



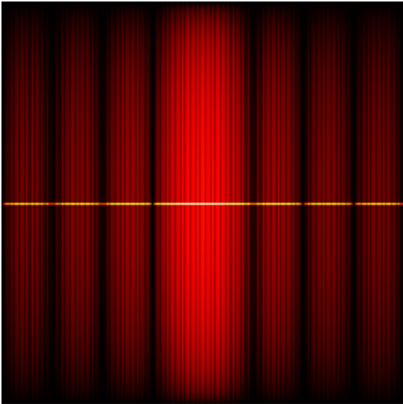
(b)

Shifted FT



(c)

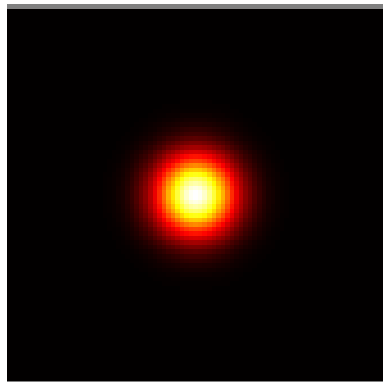
Log Scaled Shifted FT



(d)

Figure 4

Gaussian aperture



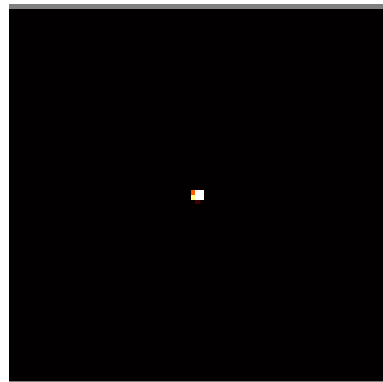
(a)

FT



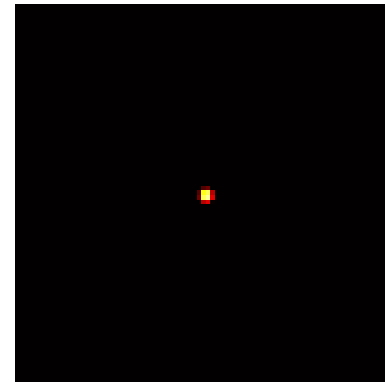
(b)

Shifted FT



(c)

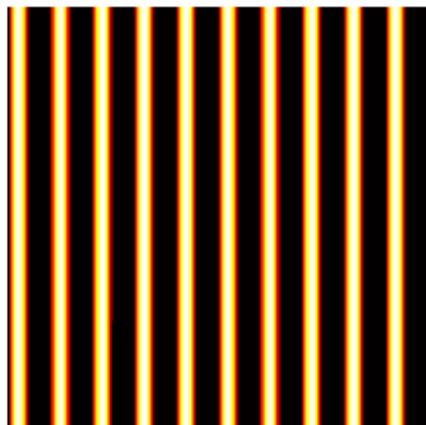
Log Scaled Shifted FT



(d)

Figure 5

Sinusoid along x aperture



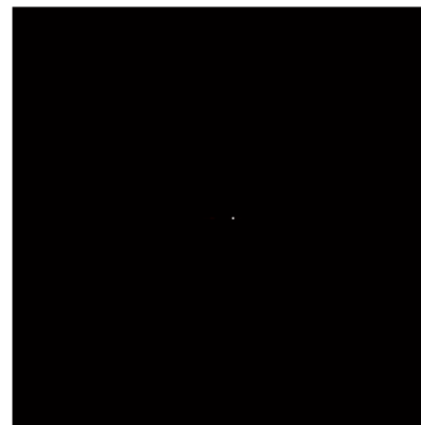
(a)

FT



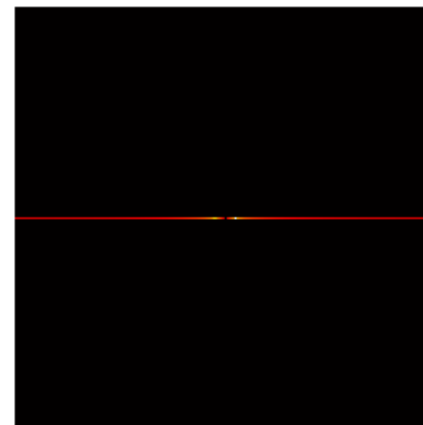
(b)

Shifted FT



(c)

Log Scaled Shifted FT

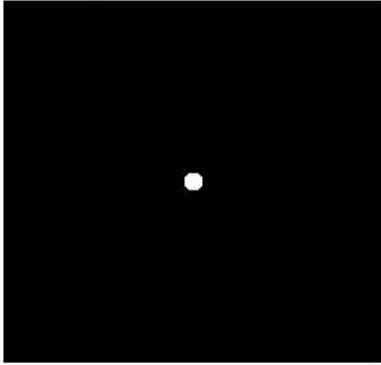


(d)

Figure 6

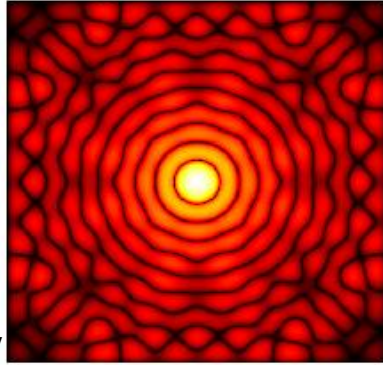
What if we vary the diameter of circular aperture?

circular aperture with 0.05 diameter



(a)

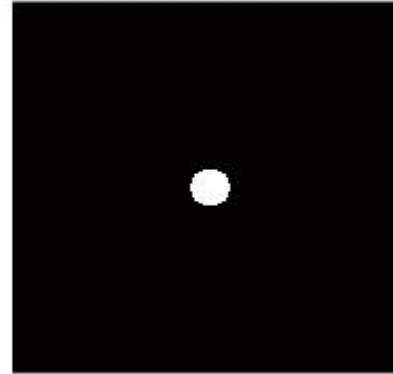
Log Scaled Shifted FT



(b)

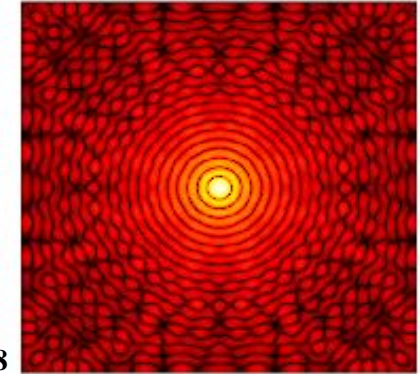
Figure 7

circular aperture with 0.1 diameter



(a)

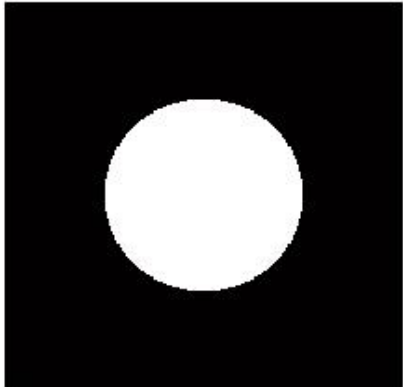
Log Scaled Shifted FT



(b)

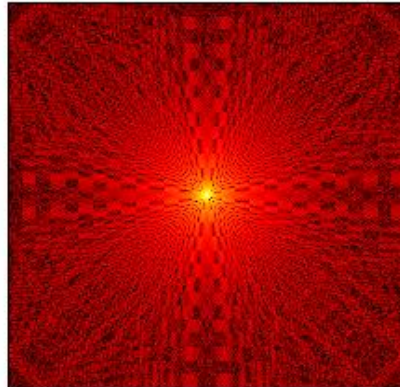
Figure 8

circular aperture with 0.5 diameter



(a)

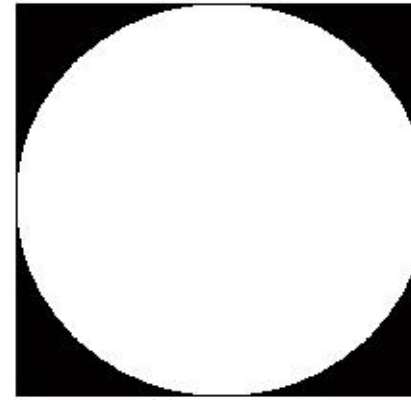
Log Scaled Shifted FT



(b)

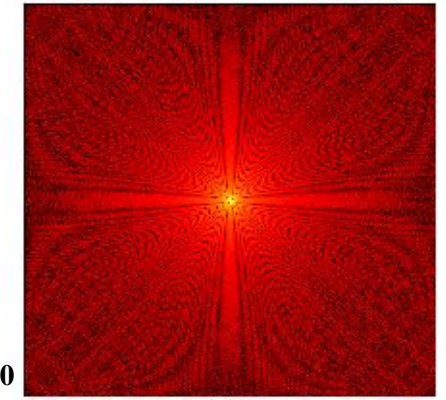
Figure 9

circular aperture with 1 diameter



(a)

Log Scaled Shifted FT



(b)

Figure 10

Here, we can observe that adjusting the aperture's diameter affects the log-shifted FFT of the diffused pattern. The **lowest diameter (0.05) has the least amount of high-frequency information in its pattern**, while **the biggest diameter (1) contains the most high-frequency information**. We can conclude that **increasing the diameter of the circular aperture permits more spatial frequencies to flow through it, resulting in a larger range of spatial frequencies in the diffraction pattern**. The opposite is true if the diameter of the circular aperture is decreased.

IMAGE RECONSTRUCTION USING FT

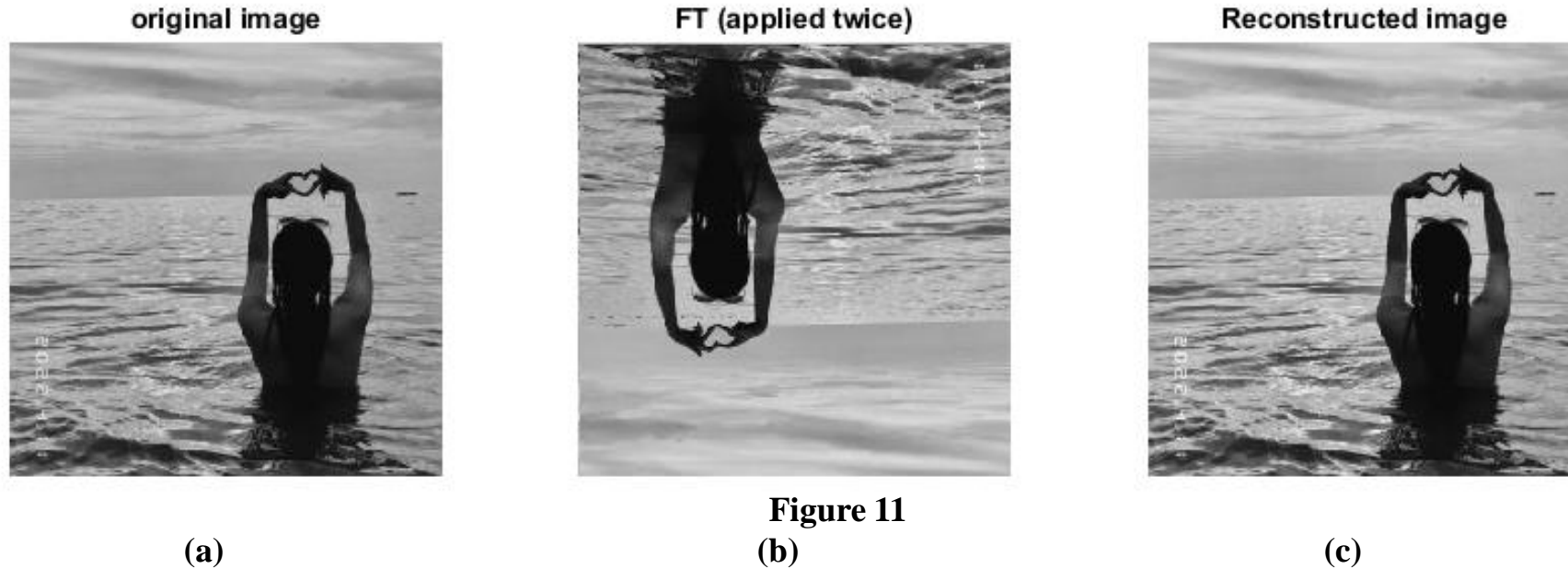
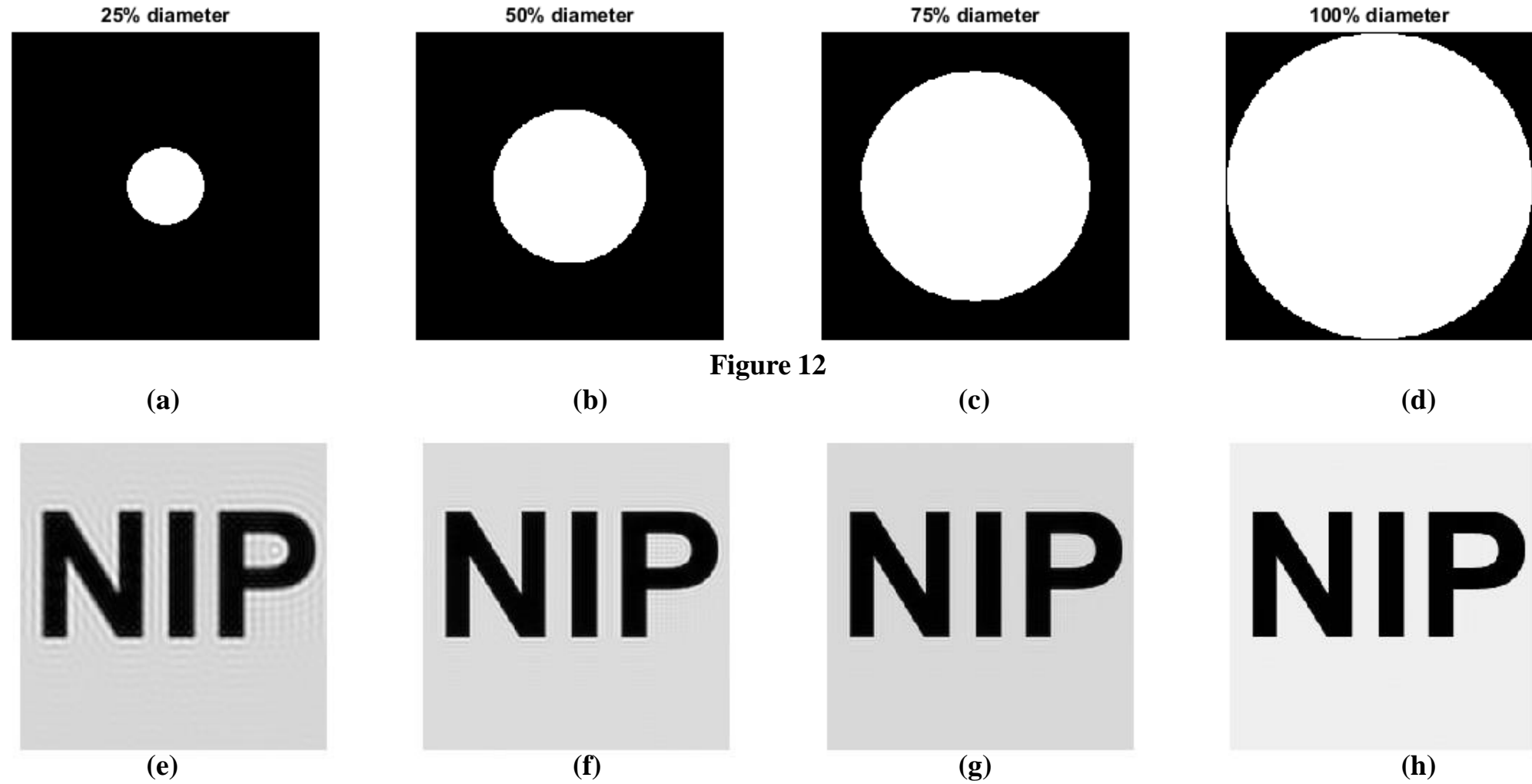


Figure 11

Here, a square RGB image was converted to grayscale using the built-in `rgb2gray` function, and the **Fourier Transform was applied twice to assess the image's frequency content at different scales and levels of detail.** Figure b shows that the resulting image was inverted or flipped upside down. This is anticipated since it is a DFT property that when DFT is applied twice to the same input data, the original signal gets inverted. (circularly). Refer to [5] and [6] for references. **Since Fourier Transform is a reversible process, we anticipated that the reconstructed picture, shown by Figure (c), would be identical to the original grayscale image.** [7]

2.2 SIMULATION OF AN IMAGING SYSTEM



ON SIMULATING IMAGING SYSTEM

In the second part of the activity, we simulated an imaging system using Fourier Transform convolution. Integrating between two 2-D functions f and g yields the following convolution formula: $h(x,y) = f \circledast g = \iint f(x',y')g(x-x',y-y')dx'dy'$. In lieu of dealing with this integral, we multiplied 2 FTs in that two functions, which in this instance are the aperture and the letters NIP, and inverted FT the resulting image, yielding the same result as convolution. We applied it with diameters of 25%, 50%, 75%, and 100% of the diameter of the array's width. The resulting images were shown on the slide preceding this one. Observing these images, we can infer that **a smaller aperture diameter**, such as 0.25, will result in more aggressive filtering of high spatial frequencies outside the circular aperture, **resulting in a blurrier image with a bigger circular zone of increased contrast**. While a larger aperture diameter, such as 1.0, will result in a less aggressive filtering of high spatial frequencies outside the circular aperture, **resulting in a less blurred image with a smaller circular zone of increased contrast**.

SIMULATION OF STAR IMAGE WITH JAMES WEBB SPACE TELESCOPE

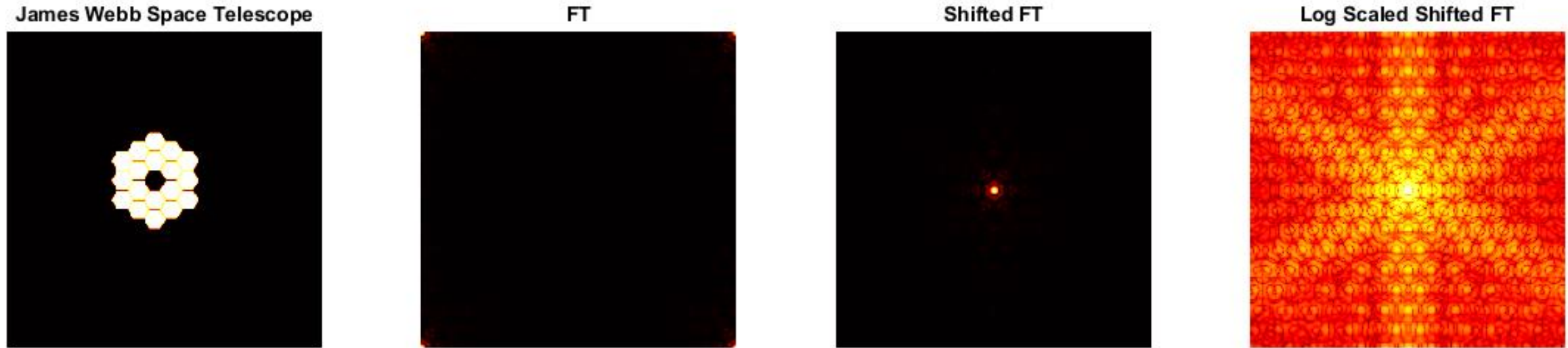
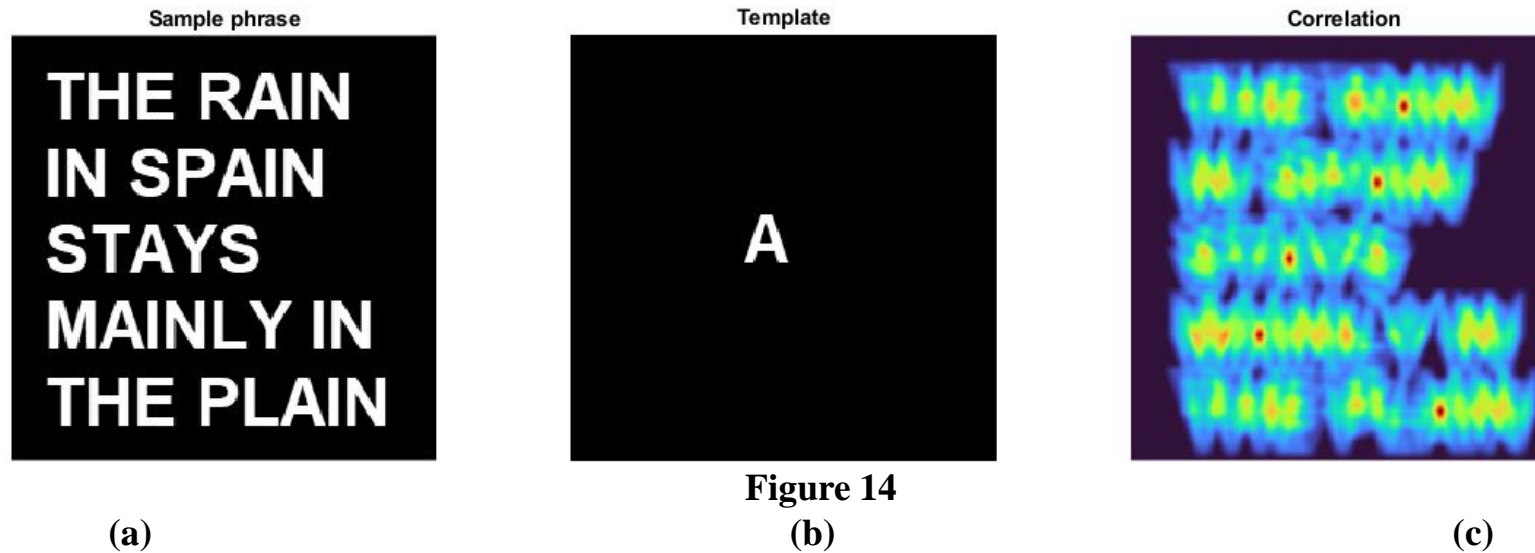


Figure 13

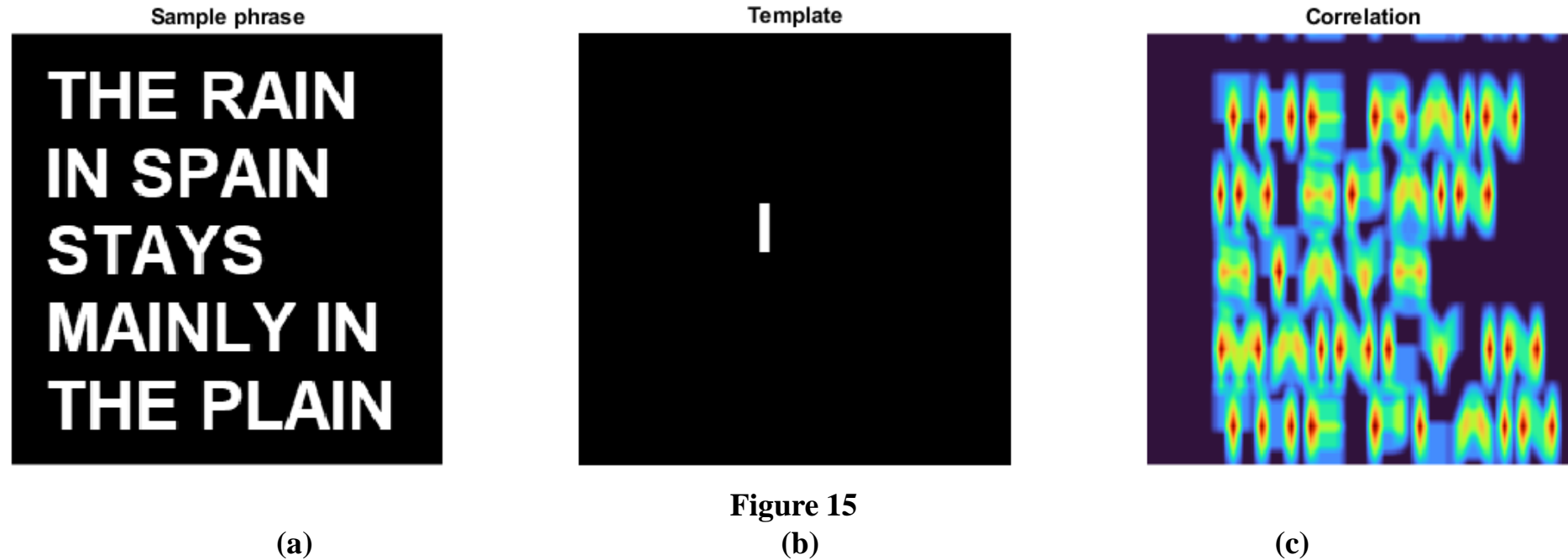
(a) (b) (c) (d)

Similar to what we did in the first part of this activity, we converted this JWST image to grayscale, computed the Fourier Transform of it using the built-in `fft2` function in MATLAB, shifted the FT using `fftshift` to better visualize the image, as seen in (c), and then enhanced the visibility of the smaller values through `rescale` and `log` functions, as shown in (d). **The resulting log scaled shifted FT image shows how a single star would appear in a JWST image.** However, it is essential to remember that stars are considered point objects, which lack a defined size, therefore the picture of a star in actuality would not be seen in the same manner as a larger object with more structure.

2.3 TEMPLATE MATCHING USING CORRELATION



Here, we searched the letter A, which served as our template, within the bigger image with the sample phrase: THE RAIN IN SPAIN STAYS MAINLY IN THE PLAIN. We put two grayscale pictures into MATLAB, then applied FT using the `fft2` function to each, resulting in a frequency domain representation of each image. We then obtained the complex conjugate of the FT of the template. And this complex conjugate was then multiplied to the FT of the sample phrase. The inverse FFT is then applied to the resultant product to generate the spatial domain representation of the correlation result; the resulting picture was then shifted to move the zero-frequency component to the image's center for improved visibility. The resulting correlation image is displayed in figure (c) above. It is evident that **the correlation highlighted the letter A inside the image's text. The placement of the letter A coincides with the red dots or peaks. These vibrant red regions suggest a strong connection between the template A and the sample picture.**



This is similar to the preceding slide, with the letter I serving as the template. The resulting image, shown in (c), displays a similar pattern of high values in locations where the template fits the sample phrase's pattern. However, **correlation values are lower than in the first case with A as the template.** This is due to the various vertical lines resembling those on the template. **This implies that the letter A is a better match than the letter I for the pattern in the example sentence.**



REFLECTION





I found this activity really enjoyable, particularly the coding part. Unlike in the last activity, I believe I had a better grasp of MATLAB in this one, but I still encountered errors at times. It was also much simpler to code than the first Activity. Seeing the predicted outcomes for each image inspired and motivated me to move on to the next one. I initially doubted some of my results, but was relieved to find out that my classmates had similar results. Lab hours really helped me finished this. Also special thanks to Mar Princer who helped me with the other codes and offered me a copy of his Canva-made JWST (credits and a big shout out to you). If trying various binary images except for the required image which is the circular aperture, then I think I deserve a bonus points or at least a perfect score. So, in short, I'll give myself a solid **10/10** :P





REFERENCES



- [1] <https://byjus.com/maths/fourier-transform/#:~:text=Fourier%20Transform%20is%20a%20mathematical,%2C%20RADAR%2C%20and%20so%20on.>
 - [2] Lab manual
 - [3] <https://www.mathworks.com/help/matlab/ref/fftshift.html>
 - [4] <https://www.mathworks.com/matlabcentral/answers/128367-show-fourier-transformed-image-in-different-scale>
 - [5] <https://dsp.stackexchange.com/questions/54864/performing-dft-twice-on-an-image-why-am-i-getting-an-inverted-image>
 - [6] <https://dsp.stackexchange.com/questions/29866/fourier-transform-of-a-fourier-transform/29870#29870>
 - [7] <https://www.cv.nrao.edu/~sransom/web/A1.html>
 - [8] - <https://www.mathworks.com/matlabcentral/answers/477280-different-colormaps-for-subplots>
- 
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The background is white with several abstract geometric elements. In the top left, there are two horizontal bars: a shorter green one above a longer light blue one. In the top right, there is a single vertical light blue bar. On the left side, there are two vertical bars: a yellow one on the left and a taller red one to its right. In the bottom right, there is a horizontal red bar and a yellow circle partially visible on the right edge.

End