Prob. 1: Show that if \( f(x,y) \) is separable into a product of a function of \( x \) and a function of \( y \), its Fourier Transform \( F(u,v) \) is also separable into a function of \( u \) and a function of \( v \). (2 points)

Prob. 2: [Exercise 3.3 - Trucco] Show that the derivative of a 1-D signal, \( I=I(x) \), amplifies the higher frequency components of the signal. Then, prove that even a slight amount of noise \( n=n(x) \), can be responsible for a large difference between the derivative \( I \) and \( I+n \) (Hint: Assume that \( n \) can be written as \( \epsilon \sin(\omega x) \) for some very small \( \epsilon \) and very large \( \omega \)). (2 points)

Prob. 3: Show that if you use the line equation \( x \cos \theta + y \sin \theta = \rho \), each image point \((x,y)\) results in a sinusoid in \((\rho,\theta)\) Hough space. Relate the amplitude and phase of the sinusoid to the point \((x,y)\). Does the period (or frequency) of the sinusoid vary with the image point \((x,y)\)? (2 points)

Prob. 4: (18 points)

Programming Assignment

Your task is to develop a vision system that recognizes lines in an image using the Hough Transform. We will call it the "line finder." Four gray-level images are provided to you: `hough_simple_1.pgm`, `hough_simple_2.pgm`, `hough_complex_1.pgm` and `empirestate.jpg`. It is enough that your line finder works on the "simple" images. If the results are good, you can try the complex and empirestate image. The task is divided into four parts, each corresponding to a program you need to write and submit.

Each program must accept arguments in the format specified as `program_name {1st argument} {2nd argument} ... {Nth argument}`.

(a) First you need to find the locations of edge points in the image. Write a program named `h1` that locates edges in a gray-level image and generates an "edge" image where the intensity at each point is proportional to edge magnitude:

`h1 {input gray-level image} {output gray-level edge image}`.

For this you may either use the squared gradient operator or the Laplacian. Since the Laplacian requires you to find zero-crossings in the image, you may choose to use the squared gradient operator. The convolution masks proposed by Sobel should work reasonably well. Else, try your favorite masks (OpenCV provides most of them) 4 points
(b) Threshold the edge image so that you are left with only strong edges. You should have a program named \texttt{h2} that thresholds a gray-level image at a certain threshold value:

\texttt{h2 \{input gray-level edge image\} \{input gray-level threshold\} \{output binary image\}.}

For that you could just rename \texttt{p1} from the previous programming assignment.

(c) Next, you need to implement the Hough Transform for line detection. Write a program named \texttt{h3} that generates an image of the Hough Transform space of a given binary edge image:

\texttt{h3 \{input binary edge image\} \{output gray-level Hough image\}.}

The brightness of each pixel (voting bin) in the output image should be proportional to the number of votes it receives. As discussed in class, the equation \( y = mx + c \) is not suitable as it requires the use of a huge accumulator array. So, use the line equation \( x \cos \theta + y \sin \theta = \rho \). Note that \( \theta \) must lie between 0 and \( \pi \), and very large values of \( \rho \) correspond to lines that lie outside the image. You can use these constraints to limit the size of the accumulator array. The resolution of the array must be selected carefully. Low resolution will not give you sufficient accuracy in estimated parameters, and very high resolution will increase computations and reduce the number of votes in each cell (or bin). You may want to vote for small patches rather than points in the accumulator array.

\textbf{6 points}

(d) Write a program named \texttt{h4} that finds lines in the image from its Hough Transform space, using a given threshold, and draw the detected lines on a copy of the original scene image:

\texttt{h4 \{input original gray-level image\} \{input gray-level Hough image\} \{input Hough threshold value\} \{output gray-level line image\}.}

[Note that it may be easier to write one main program \texttt{h34} that performs both tasks:

\texttt{h34 \{input original gray-level image\} \{input binary edge image\} \{output gray-level Hough image\} \{input Hough threshold value\} \{output gray-level line image\} ]

Make sure that the lines you draw are clearly visible irrespective of pixel brightness values (dark or bright). Straight lines in the image will produce blurred "areas of brightness" in the Hough space - not single points, as predicted by the theory. The theoretical model uses the approximation that lines are infinitely long and infinitely thin - which is not exactly true in practice. The suggested approach to deal with this problem is to first threshold the Hough space image by cutting out all pixels below the given threshold value, and then segment the result into a number of "areas of brightness", each area presumably corresponding to a particular straight line (for that you could adapt the code developed in the previous assignment). Then in order to calculate the parameters of the line, you would have to find the "center" of its bright area. Notice that the approach
we used when calculating positions of binary objects will probably not work, as different
points in the Hough space will have different brightness (number of votes), and should
have different contribution into the position of the "center". You may want to use a
weighted average based on points’ intensities to locate the "center" (similar to how you
would locate the center of mass of a non-homogeneous body). **8 points**

• (e) **Bonus Question:** Note that the above implementation does not detect end-points of
line segments in the image. Modify h4 to make it prune the detected lines so that they do
not extend beyond the objects they belong to. If your are successful, you get **4 bonus
points**!
As usual, you should submit a **README** file specifying what threshold values you used.