

Kunal Goyal
CS558 – Computer Vision
Homework2

Code:

```
import numpy as np
import random
import itertools
import math
import cv2
```

```
def apply_filter(filter, arr):
    try:
        """
        This method convolves an image with a given kernel.
        Parameters
        -----
        filter:array, Required

        arr: array, Required

        return tuple
        """

        canvas = updated_arr(arr)
        img = updated_arr(arr)
        for i in range(1, len(arr) - 1):
            for j in range(1, len(arr[i]) - 1):
                pos = 0
                for rows in range(len(filter)):
                    for columns in range(len(filter[rows])):
                        x = rows - (len(filter) // 2)
                        y = columns - (len(filter[rows]) // 2)
                        pos += filter[rows][columns] * arr[i + x][j + y]
                canvas[i][j] = pos

                img[i][j] = 0 if pos < 0 else 255 if pos > 255 else pos
        return (canvas, img)
    except:
        print("error in Apply_filter fn")
```

```
def overlay_image(item1, item2, bg = False):
    try:
        """
        This method used to overlays images based on gradient value.
```

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Parameters

item1:array , Required

item2:array, Required

bg:boolean, Optional

return tuple of item1 and image

"""

```
img = item1.copy()
```

```
for i in range(len(item1)):
```

```
    for j in range(len(item1[i])):
```

```
        if not bg:
```

```
            val = math.sqrt(item1[i][j] ** 2 + item2[i][j] ** 2)
```

```
            item1[i][j] = val
```

```
            img[i][j] = 0 if val < 0 else 255 if val > 255 else val
```

```
        else:
```

```
            if item2[i][j] > item1[i][j]:
```

```
                item1[i][j] = item2[i][j]
```

```
            img[i][j] = 0 if item1[i][j] < 0 else 255 if item1[i][j] > 255 else item1[i][j]
```

```
    return (item1, img)
```

```
except:
```

```
    print("Error in overlay_image function")
```

```
def non_max_supresn(arr, horizontal, vertical, mode):
```

```
    try:
```

```
        """
```

```
        This method supresses pixel intensities based on the neighborhood pixel,
```

Parameters

arr: array, Required

horizontal:array, Required

vertical:array, Required

mode : str , Required

return list

```
    """
```

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```
canvas = arr.copy()
img = arr.copy()
for i in range(len(arr)):
    for j in range(len(arr[i])):
        if mode == "edges":
            angle = math.atan2(vertical[i][j], horizontal[i][j])
            canvas[i][j] = arr[i][j]
            if i == 0 or j == 0 or i == len(arr) - 1 or j == len(arr[i]) - 1:
                canvas[i][j] = 0
            elif (angle >= -1*math.pi/8 and angle <= math.pi / 8) or (angle > 7*math.pi/8 and
angle <= -7*math.pi/8):
                if arr[i][j] <= arr[i][j+1] or arr[i][j] <= arr[i][j-1]:
                    canvas[i][j] = 0
                elif (angle < -1*math.pi/8 and angle >= -3*math.pi/8) or (angle > math.pi/8 and angle
<= 3*math.pi/8):
                    if arr[i][j] <= arr[i+1][j+1] or arr[i][j] <= arr[i-1][j-1]:
                        canvas[i][j] = 0
                    elif (angle < -3*math.pi/8 and angle >= -5*math.pi/8) or (angle > 3*math.pi/8 and
angle <= 5*math.pi/8):
                        if arr[i][j] <= arr[i+1][j] or arr[i][j] <= arr[i-1][j]:
                            canvas[i][j] = 0
                        elif (angle < -5*math.pi/8 and angle >= -7*math.pi/8) or (angle > 5*math.pi/8 and
angle <= 7*math.pi/8):
                            if arr[i][j] <= arr[i+1][j-1] or arr[i][j] <= arr[i-1][j+1]:
                                canvas[i][j] = 0
                            else:
                                canvas[i][j] = 0

            img[i][j] = 0 if canvas[i][j] < 0 else 255 if canvas[i][j] > 255 else canvas[i][j]

        elif mode == "corners":
            canvas[i][j] = arr[i][j]
            if (i == 0 or j == 0 or i == len(arr) - 1 or j == len(arr[i]) - 1):
                canvas[i][j] = 0
            elif not (arr[i][j] > arr[i+1][j+1] and arr[i][j] > arr[i-1][j-1] and arr[i][j] >
arr[i+1][j-1] and arr[i][j] > arr[i-1][j+1] and arr[i][j] > arr[i][j+1] and arr[i][j] > arr[i][j-1] and
arr[i][j] > arr[i+1][j] and arr[i][j] > arr[i-1][j]):
                canvas[i][j] = 0

            img[i][j] = 0 if canvas[i][j] < 0 else 255 if canvas[i][j] > 255 else canvas[i][j]

return [canvas, img]
```

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```
except:  
    print("Error in non_max_supresn fn")
```

```
def threshold(canvas,min_value,max_value):  
    try:
```

```
        """
```

This method used to set pixel value to zero when it less than min_value, and if it is greater than min_value and less than max_value then set to 125 else 255.

Parameters

canvas: , Required

min_value:int, Required

max_value:int, Required

return canvas

```
        """
```

```
        canvas_len=len(canvas)
```

```
        for i in range(canvas_len):
```

```
            canvas_item_len=len(canvas[i])
```

```
            for j in range(canvas_item_len):
```

```
                canvas[i][j] = 0 if canvas[i][j] < min_value else 125 if canvas[i][j] < max_value else 255
```

```
        return canvas
```

```
except:  
    print("Error in threshold function")
```

```
def hes_matrix(xxcord, yycord, xycord, yxcord, threshold):  
    try:
```

```
        """
```

This method used to hessian matrix to detect corners

Parameters

xxcord:array , Required

yycord:array, Required

xycord:array , Required

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yxcord:array, Required

threshold:float, Required

return list of list item

"""

```
arr = xxcord.copy()
img = xxcord.copy()
for i in range(len(xxcord)):
    for j in range(len(xxcord[i])):
        det = xxcord[i][j]*yycord[i][j]-xycord[i][j]*yxcord[i][j]
        trace = xxcord[i][j] + yycord[i][j]
        r = det - .06*(trace**2)
        arr[i][j] = r
        if r > threshold:
            img[i][j] = 255
        else:
            arr[i][j] = 0
            img[i][j] = 0
    return [arr, img]
except:
    print("Error in hes_matrix fn")
```

def updated_arr(img):

try:

"""

This method create a array of a image

Parameters

img:array, Required

return list

"""

```
arr_item = []
for i in range(len(img)):
    arr_item += [[]]
    for j in range(len(img[i])):
        arr_item[i] += [img[i][j]]
arr_item = np.array(arr_item, dtype="float32")
return arr_item
except:
    print("Error updated_arr in Function")
```

def corners_to_list(corners):

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```
try:
    """
    This methods change the list of corners into a list
```

Parameters

corners: array, Required

Return list of corners

"""

```
corners_list = []
for i in range(len(corners)):
    for j in range(len(corners[i])):
        if corners[i][j] > 0:
            corners_list += [(i, j)]
return corners_list
```

```
except:
    print("Error in Corner_to_list function")
```

```
def ransac_algo(corners, threshold, inliers, it):
```

```
    try:
        """
```

This methods uses the RANSAC algorithm on set of points of image.

Parameters

corners: array, Required

threshold: float, Required

inliers: Required

it: int, Required

Return list of list

"""

```
maxpts = []
passes = []
success = []
used = []
endpoints = []
for i in range(it):
    maxpts += [[0, 0]]
```

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```
passes += [(0, 0)]
success += [0]
used += [[]]
endpoints += [[]]
for j in range(17):
    items = random.sample(range(len(corners)), 2)
    endpoints[j] = [corners[items[0]], corners[items[1]]]
    passes[j] = [corners[items[0]], corners[items[1]]]
    line_dim = (corners[items[0]][0] -
corners[items[1]][0])**2 + (corners[items[0]][1] -
corners[items[1]][1])**2
    try:
        m = (corners[items[0]][0] - corners[items[1]][0]) / (corners[items[0]][1] -
corners[items[1]][1])
    except:
        continue
    if m == 0:
        continue
    b = -m*corners[items[0]][1] + corners[items[0]][0]
    for k in range(len(corners)):
        cornerx = (corners[k][1] / m + corners[k][0] - b)/(m + 1/m)
        cornery = cornerx * m + b
        d = (corners[k][1] - cornerx) ** 2 + (corners[k][0] - cornery) ** 2
        if d <= threshold ** 2:
            success[j] += 1
            used[j] += [corners[k]]
            first_dim = (corners[k][0] - endpoints[j][0][0])**2 + (corners[k][1] -
endpoints[j][0][1])**2
            second_dim = (corners[k][0] - endpoints[j][1][0])**2 + (corners[k][1] -
endpoints[j][1][1])**2
            if first_dim > line_dim or second_dim > line_dim:
                if first_dim <= second_dim:
                    if first_dim > maxpts[j][0]:
                        maxpts[j][0] = first_dim
                        passes[j][0] = corners[k]
                else:
                    if second_dim > maxpts[j][1]:
                        maxpts[j][1] = second_dim
                        passes[j][1] = corners[k]
            if success[j] >= inliers:
                return [passes[j], used[j], endpoints[j]]
winner = success.index(max(success))
return [passes[winner], used[winner], endpoints[winner]]
except:
    print("Error in ransac_algo function")
```

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```
def apply_ransac(img, corners, threshold =  
math.sqrt(3.84), inliers = 1000, features = 4, it = 17):
```

```
    try:
```

```
        """
```

This methods apply the RANSAC algorithm on an
image as per iteration value(it parameter) ,
and creates an image.

Parameters

img: array, Required

corners: array, Required

threshold: float, Optional

inliers: int, Optional

features: int, Optional

it: int, Optional

Return array

```
        """
```

```
        colors = list(itertools.product([0, 255], repeat = 3))
```

```
        color = random.sample(colors[1:-1], features + 1)
```

```
        for i in range(features):
```

```
            winner = ransac_algo(corners, threshold, inliers, it)
```

```
            img = cv2.line(img, (winner[0][0])[:-1], (winner[0][1])[:-1], color[i], 1)
```

```
            for j in range(len(winner[1])):
```

```
                pt = winner[1][j]
```

```
                for row in range(3):
```

```
                    for column in range(3):
```

```
                        x = row - 1
```

```
                        y = column - 1
```

```
                        img[pt[0] + x][pt[1] + y] = color[i] if pt != winner[2][0] and pt != winner[2][1]
```

```
        else color[-1]
```

```
            corners.remove(pt)
```

```
        return img
```

```
    except:
```

```
        print("Error in apply_ransac function")
```

```
def hough_transform(img, corners, angle, radius,
```


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features):

try:

"""

This method apply Hough Transform on a set of points.

Parameters

img:array, Required

corners:array, Required

angle: int, Optional

radius: Optional

features: int, Optional

return cordinates, list

"""

cordinates = []

for i in range(features):

 cordinates += [(0, 0), 0]

hough = []

for i in range(radius):

 hough += [[]]

 for j in range(angle):

 hough[i] += [0]

for i in range(len(corners)):

 for j in range(0, angle):

 r = corners[i][1]*math.cos(j * math.pi / angle) + corners[i][0]*math.sin(j * math.pi /

angle)

 rbin = math.floor(radius/2 / math.sqrt(len(img)**2 + len(img[1])**2) * r + radius/2)

 r = math.floor(r)

 hough[rbin][j] += 1

 for k in range(features):

 if hough[rbin][j] > cordinates[k][1]:

 if [(r, j * math.pi / angle), hough[rbin][j]-1] in cordinates:

 index = cordinates.index([(r, j * math.pi / angle), hough[rbin][j]-1])

 cordinates = cordinates[:k] + [(r, j * math.pi / angle), hough[rbin][j]] +

cordinates[k:index] + cordinates[index+1:]

 else:

 cordinates = cordinates[:k] + [(r, j * math.pi / angle), hough[rbin][j]] +

cordinates[k:-1]

 break

 return cordinates

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```
except:
    print("Error in hough_transform function")

def apply_hough(img, corners, angle = 180, radius =
None, features = 4):
    try:
        """
        This method apply Hough Transform a specified
        amount of times, and creates an image depicting the
        result.

        Parameters
        -----
        img:array, Required

        corners:array, Required

        angle: int, Optional

        radius: Optional

        features: int, Optional

        return array

        """
        if radius == None:
            radius = math.ceil(2*math.sqrt(len(img)**2 +
len(img[1])**2))
            hough = hough_transform(img, corners, angle,
radius, features)
            for i in range(features):
                pt1 = int(hough[i][0][0]/math.sin(hough[i][0][1]))
                pt2 = int((hough[i][0][0] -
len(img[1])*math.cos(hough[i][0][1]))/math.sin(hough[i][0][1]))
                img = cv2.line(img, (0, pt1), (len(img[1]), pt2), (0, 0 , 255), 1)
            return img
        except:
            print("Error in apply_hough function")

if __name__ == "__main__":

    try:
        """Filtering variable and corresponding values.
        g: gaussian
```

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h_sobel: Horizontal Sobel Filter values

v_sobel: Vertical Sobel Filter values

"""

```
g = [[0.077847, 0.123317, 0.077847],  
      [0.123317, 0.195346, 0.123317],  
      [0.077847, 0.123317, 0.077847]]
```

```
h_sobel = [[1, 2, 1],  
            [0, 0, 0],  
            [-1, -2, -1]]
```

```
v_sobel = [[1, 0, -1],  
            [2, 0, -2],  
            [1, 0, -1]]
```

```
img = cv2.imread("road.png", 0)  
arr = updated_arr(img)
```

```
gfilImg = apply_filter(g, arr)  
h = apply_filter(h_sobel, gfilImg[0])  
v = apply_filter(v_sobel, gfilImg[0])
```

"""Apply filter for Horizontal and Vertical sobel

cord1_xx: for XX coordinates

cord1_yy: for YY coordinates

cord1_xy: for XY coordinates

cord1_yx: for YX coordinates

"""

```
cord1_xx = apply_filter(h_sobel, h[0])  
cord1_yy = apply_filter(v_sobel, v[0])  
cord1_xy = apply_filter(v_sobel, h[0])  
cord1_yx = apply_filter(h_sobel, v[0])
```

```
edges = overlay_image(h[0], v[0])  
suppress_edges = non_max_supresn(edges[0], h[0], v[0], "edges")  
threshold_edges = threshold(suppress_edges[0], 175, 60)
```

```
hess_matrix = hes_matrix(cord1_xx[0], cord1_yy[0], cord1_xy[0], cord1_yx[0], 175000)  
hess_threshold = non_max_supresn(hess_matrix[0], h[0], v[0], "corners")  
updated_hess_matrix = overlay_image(hess_threshold[0], threshold_edges / 4, bg=True)
```

```
colored = (updated_hess_matrix[1]).copy()  
colored = cv2.cvtColor(updated_hess_matrix[1], cv2.COLOR_GRAY2RGB)
```

"""

Applying RANSAC

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```
"""
```

```
ransac_image = colored.copy()  
ransac_image = apply_ransac(ransac_image,  
corners_to_list(hess_threshold[1]), it = 25)
```

```
"""
```

```
Hough Transformation with the hess_threshold values
```

```
"""
```

```
Hough_transform = colored.copy()  
Hough_transform = apply_hough(Hough_transform,  
corners_to_list(hess_threshold[1]), angle=45)
```

```
"""
```

```
Saving all output images
```

```
"""
```

```
cv2.imwrite("gaussian_filter.png", gfilImg[1])  
cv2.imwrite("h_sobel_filter.png", h[1])  
cv2.imwrite("v_sobel_filter.png", v[1])  
cv2.imwrite("cord1.png", cord1_xx[1])  
cv2.imwrite("cord2.png", cord1_yy[1])  
cv2.imwrite("cord3.png", cord1_xy[1])  
cv2.imwrite("cord4.png", cord1_yx[1])  
cv2.imwrite("edges_with_no_supress.png", edges[1])  
cv2.imwrite("edges_with_supress.png", suppress_edges[1])  
cv2.imwrite("edges_threshold.png", threshold_edges)  
cv2.imwrite("corners.png", hess_matrix[1])  
cv2.imwrite("corners_threshold.png", hess_threshold[1])  
cv2.imwrite("updated_corners.png", updated_hess_matrix[1])  
cv2.imwrite("ransac.png", ransac_image)  
cv2.imwrite("hough_image.png", Hough_transform)  
except:  
    print("Error n main Fn")
```

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Explanation and Images:

Pre-Processing Step

The code reads in the image from the provided file, road.png, converts it to grayscale and then begins the pre-processing part of the code.

The pre-processing starts by first creating a Gaussian filter using the Gaussian value equation provided in the slides. The result of applying the filter to the image can be seen below in Figure 1.



Figure 1: Gaussian Filtered Image

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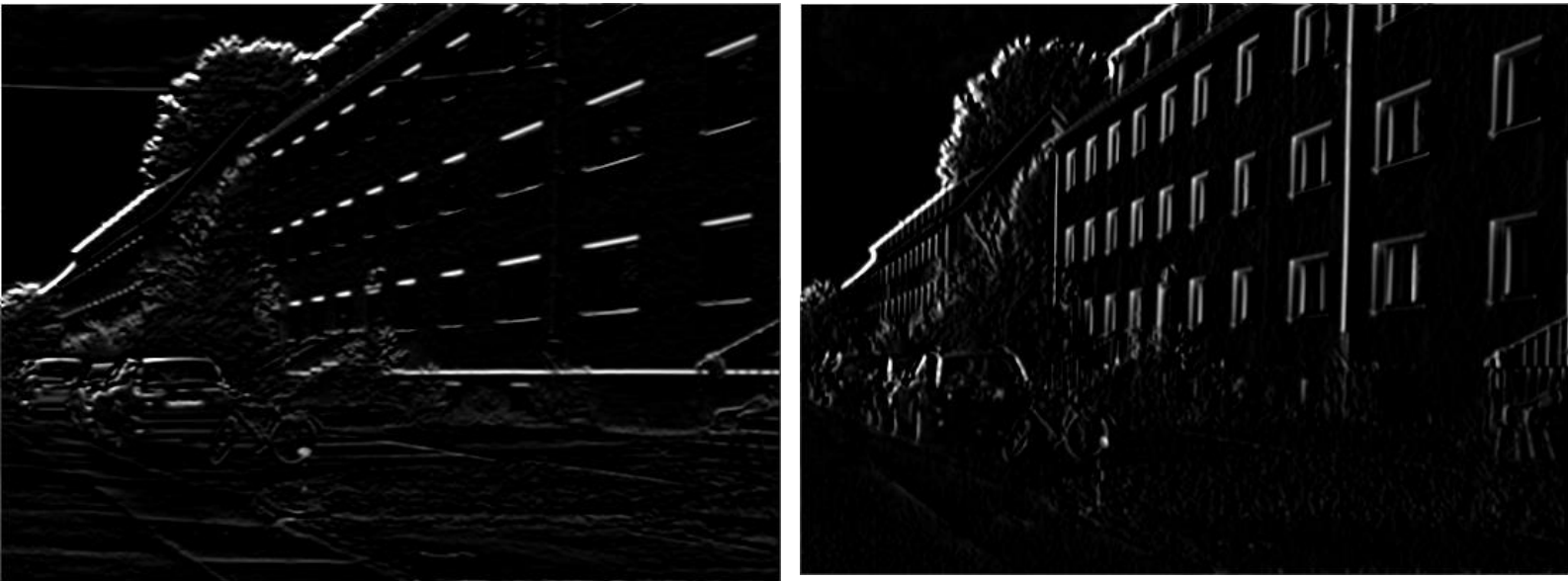


Figure 2: Horizontal Sobel Filtered Image (Left) and Vertical Sobel Filtered Image (Right)

The image is then subjected to Sobel filters as derivatives operators in order to determine the image's second x, second y, and x of y derivatives. The determinant of the Hessian matrix is then calculated using these derivative matrices. The Hessian matrix's determinant was then threshold, with all pixels with values less than 150 set to 0 and all pixels with values more than 150 set to 255. All of the places on the image where the Hessian detector responds significantly were shown as a result. The black tree and the extremely light sky have a significant intensity difference.



Thresholded Hessian Response

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Non Maximum Suppressed Response



Edges with No suppress

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Edges with suppress

The result was then subjected to non-maximum suppression in order to extract examples of the best Hessian detector responses. Figure 2 shows the thresholded Hessian detector response as well as the non-maximum suppressed version. The Hessian detector responds strongly to many of the angles and straight lines that make up the building's characteristics, which is good, but it also responds strongly to the tree in the image's upper left corner, which is not ideal.

RANSAC:

The RANSAC portion of the code takes within the yield of pre-processing step which ought to be an array of all the zones on the picture where the Hessian detector reacted exceptionally unequivocally after non-maximum suppression was connected. The algorithm first gathers all the points in the image where the locator reacted strongly and stores them as tuples in a list that may be easily accessed later. Two points are then picked at random from the list, and the parameters for the condition of a line, slant m , and y caught c , are discovered using some basic variable-based algebra. The rest of the focuses are at that point circled through in order to decide the focuses that are near sufficient to the line to be inliers for the line. This is often done by calculating the opposite separate between the point and the line and checking in the event that it is underneath a certain threshold.

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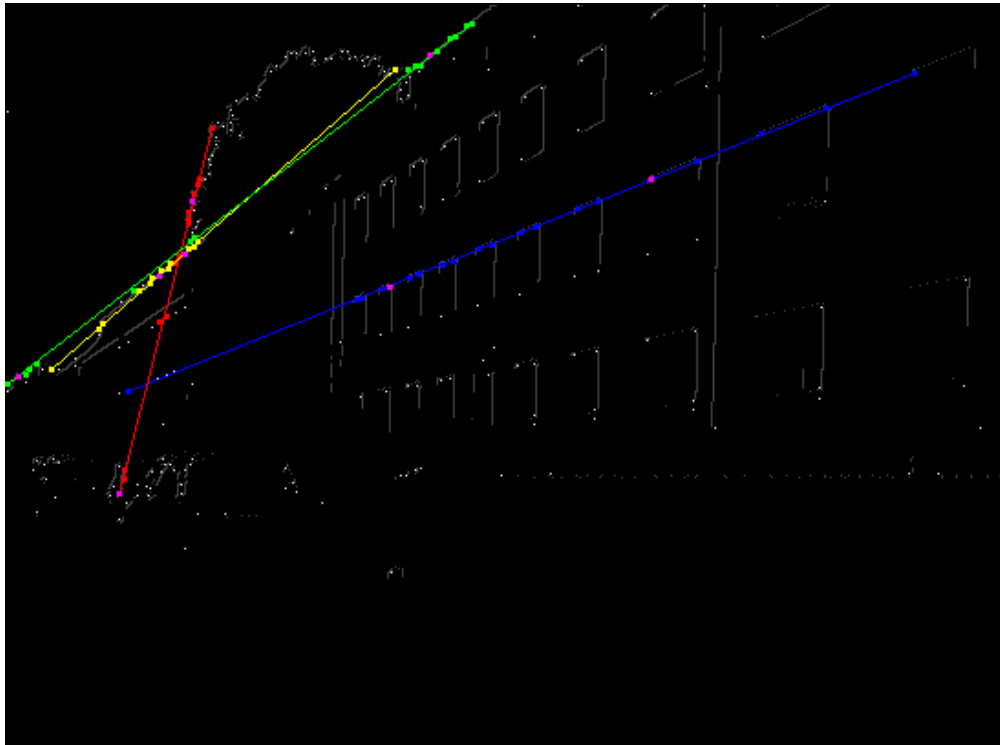
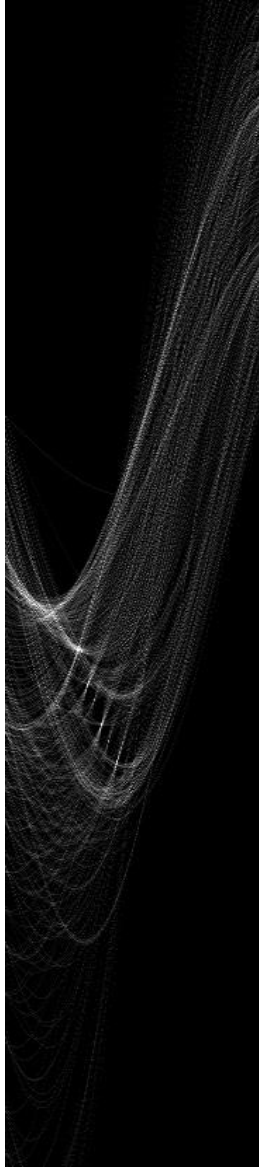


Figure 5: Four Best Lines Produced by RANSAC Plotted

Hough:

The Hough transform section of the code, like RANSAC, takes in the pre-processing step's output, which should be an array of all the locations on the image where the Hessian detector responded extremely strongly when non-maximum suppression was applied. The algorithm begins by calculating the size that the accumulator will require. To do this, the highest and lowest values for the function $\rho = x \cos \theta$. The algorithm begins by calculating the size that the accumulator will require. To do this, the highest and lowest values for the function $\rho = x \cos \theta$ Because we can't directly express the negative values that rho will take in the accumulator, the range of values is used for the first dimension. To compensate for this, we apply an offset, which is simply the negative of the low value multiplied by the rho given by the equation to produce a correct index to utilise.

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Visualized Hough Space

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Suppressed Hough Space

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Areas Removed From Hough Space

The algorithm extracts all of the points in the picture where the detector responded strongly and saves them as tuples in a list that may be retrieved later. The line in the image space with the most support based on the number of image space points that voted for it should correspond to the point in the Hough space with the highest value. Non maximum suppression is performed to the Hough space before identifying the points with the highest support in order to prevent near identical lines from being proposed. The model is transformed back into two linear form and two points are created in order to draw a line between them on the picture after the strongest point in suppressed Hough space is discovered.

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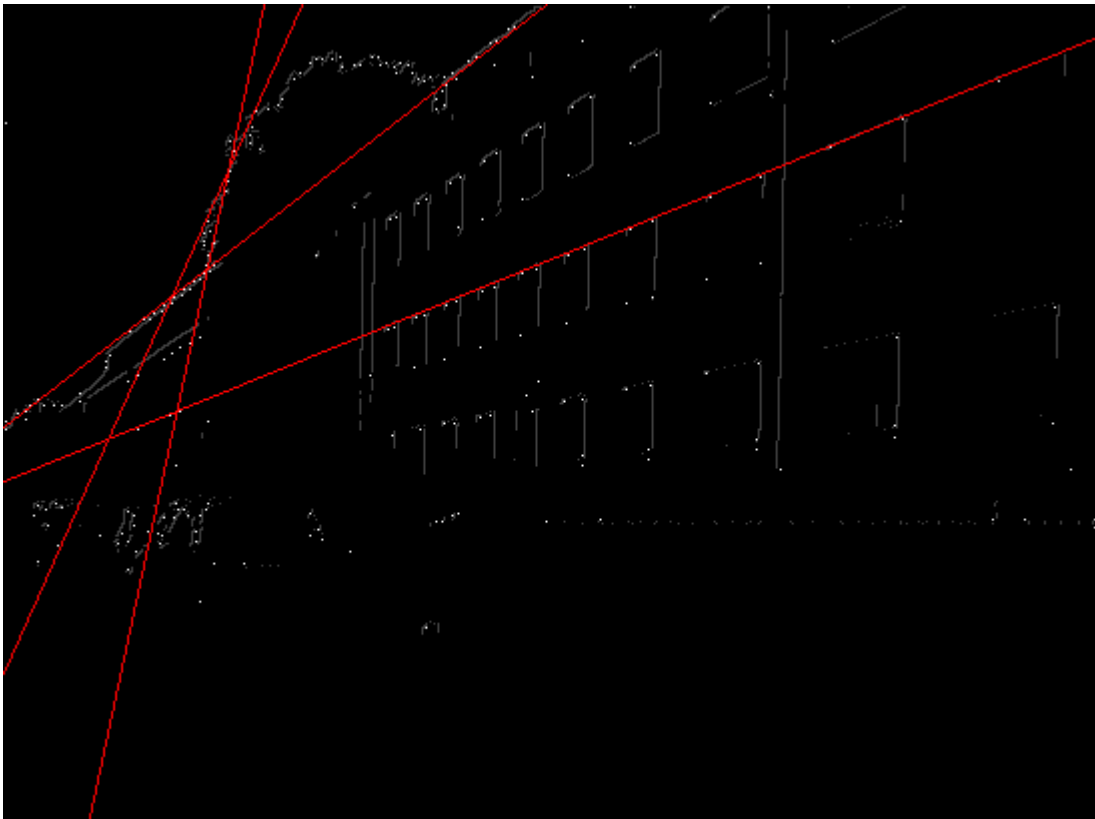


Figure 7: Hough Lines on Original Image