

Homework 2 - Generalized Hough Transform

Theory

Task 1: ii

Task 2: i

Programming

Find object in an image using a template:

title title

```
# Group G9
# Huaiyi Dai 408002
# Ruoxiao Wang 462318
# Di Xu 464995
#
#
#!/usr/bin/env python3
# -*- coding: utf-8 -*-
from tkinter import N
import cv2
import utils
import numpy as np
from matplotlib import pyplot as plt
from sklearn.metrics.pairwise import euclidean_distances

def nonMaxSuppression(img, d=5):
    """
    Given an image set all values to 0 that are not
    the maximum in its (2d+1,2d+1)-window

    Parameters
    -----
    img : ndarray
        an image
    d : int
        for each pixels consider the surrounding (2d+1,2d+1)-window

    Returns
    -----
    result : ndarray
    """
    rows,cols = img.shape
    result = np.zeros((rows,cols))
```

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# TODO
for i in range(rows):
    for j in range(cols): # iterate over pixels
        min_col = max(0, j-d)
        min_row = max(0, i-d)
        max_col = min(cols, j+d)
        max_row = min(rows, i+d)
        # iterate over (2d+1,2d+1) neighborhood window
        maxi = img[min_row:max_row,min_col:max_col].max()
        if img[i,j] == maxi:
            result[i,j] = maxi # store results in new array

return result

def calcBinaryMask(img, thresh = 0.3):
    """
    Compute the gradient of an image and compute a binary mask
    based on the threshold. Corresponds to 0^B in the slides.

    Parameters
    -----
    img : ndarray
        an image
    thresh : float
        A threshold value. The default is 0.3.

    Returns
    -----
    binary : ndarray
        A binary image.

    """

    # TODO:
    grad = utils.calcDirectionalGrad(img) # -compute gradients
    grad_abs = np.abs(grad)
    rows,cols = img.shape

    result = np.zeros([rows, cols])
    TH = thresh * np.max(grad_abs) # -threshold gradients
    for i in range(0, rows):
        for j in range(0, cols):
            if (grad_abs[i, j] < TH):
                result[i, j] = 0
            else:
                result[i, j] = 1

    return result # -return binary mask

```

```

def correlation(img, template):
    """
    Compute a correlation of gradients between an image and a
    template.

    Note:
    You should use the formula in the slides using the fourier
    transform.
    Then you are guaranteed to succeed.

    However, you can also compute the correlation directly.
    The resulting image must have high positive values at positions
    with high correlation.

    Parameters
    -----
    img : ndarray
        a grayscale image
    template : ndarray
        a grayscale image of the template

    Returns
    -----
    ndarray
        an image containing the correlation between image and template
    gradients.
    """

    # TODO:
    w, h = template.shape[::-1]
    #print(w,h)
    OB_temp = calcBinaryMask(template, thresh = 0.3)
    #result = cv2.matchTemplate(img, template, cv2.TM_CCORR_NORMED,
    None, OB_temp)

    #####
    # -compute gradient of the image
    II_img = utils.calcDirectionalGrad(img)
    # -compute gradient of the template
    OI_temp = utils.calcDirectionalGrad(template)
    T_temp = OI_temp * OB_temp
    # -copy template gradient into larger frame
    rows,cols = II_img.shape
    T_temp_larger = np.zeros((rows,cols), dtype=complex)
    T_temp_larger[0:h, 0:w] = T_temp
    # -apply a circular shift so the center of the original template
    is in the upper left corner
    temp_shift = utils.circularShift(T_temp_larger, int(w/2),
    int(h/2))

```

```

# -normalize template
temp_shift = temp_shift / np.max(np.abs(temp_shift))
# -compute correlation
F_temp = np.fft.fft2(temp_shift).conjugate()
F_img = np.fft.fft2(II_img)
result = np.real(np.fft.ifft2(F_img * F_temp))

return result

```

```

def GeneralizedHoughTransform(img, template, angles, scales):
    """
    Compute the generalized hough transform. Given an image and a
    template.

    Parameters
    -----
    img : ndarray
        A query image
    template : ndarray
        a template image
    angles : list[float]
        A list of angles provided in degrees
    scales : list[float]
        A list of scaling factors

    Returns
    -----
    hough_table : list[(correlation, angle, scaling)]
        The resulting hough table is a list of tuples.
        Each tuple contains the correlation and the corresponding
    combination
        of angle and scaling factors of the template.

        Note the order of these values.
    """
    # TODO:
    angles=np.array(angles)
    scales=np.array(scales)
    hough_table=[]
    for ele1 in angles:
        for ele2 in scales:
            temp = utils.rotateAndScale(template, ele1, ele2)

            corr = correlation(img, temp)
            hough_table.append([corr,ele1,ele2])

    return hough_table

```

Main Program

```
# Load query image and template
query = cv2.imread("data/query.jpg", cv2.IMREAD_GRAYSCALE)
template = cv2.imread("data/template.jpg", cv2.IMREAD_GRAYSCALE)

# Visualize images
utils.show(query)
utils.show(template)

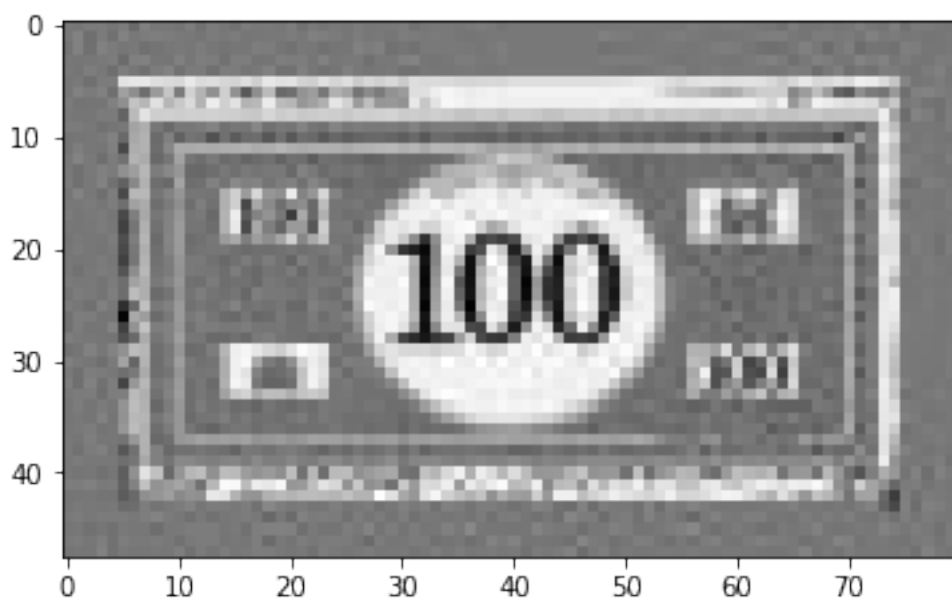
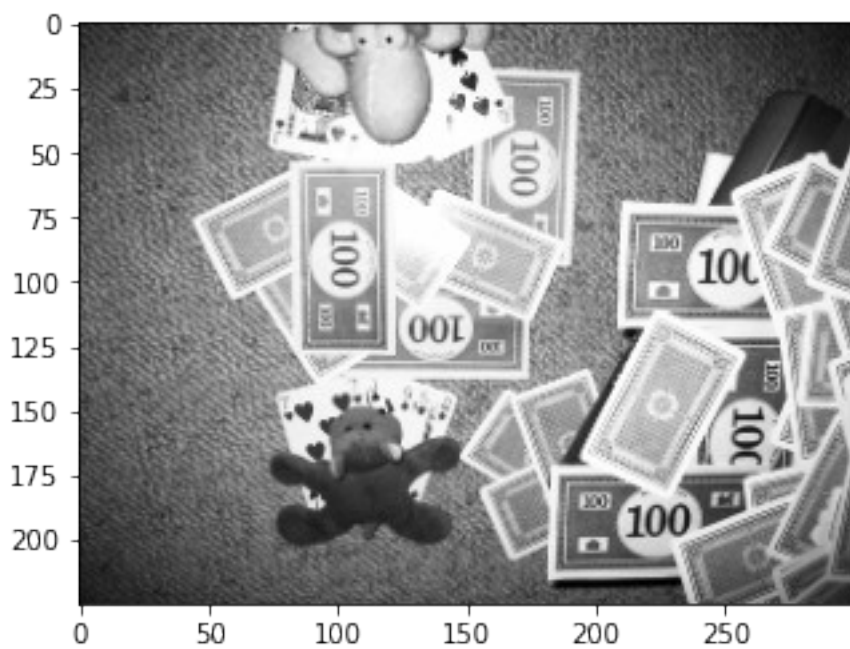
# Create search space and compute GHT
angles = np.linspace(0, 360, 36)
scales = np.linspace(0.9, 1.3, 10)
ght = GeneralizedHoughTransform(query, template, angles, scales)

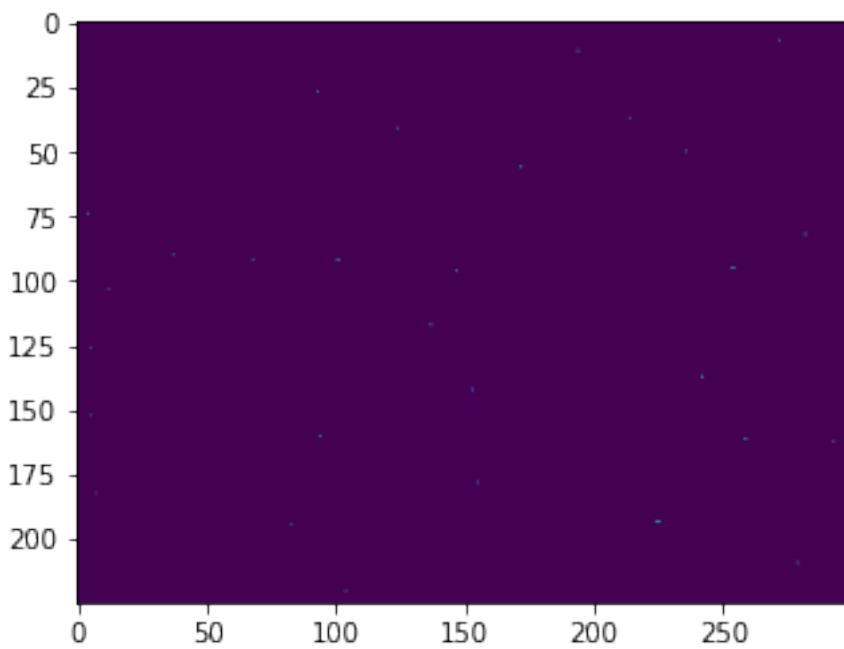
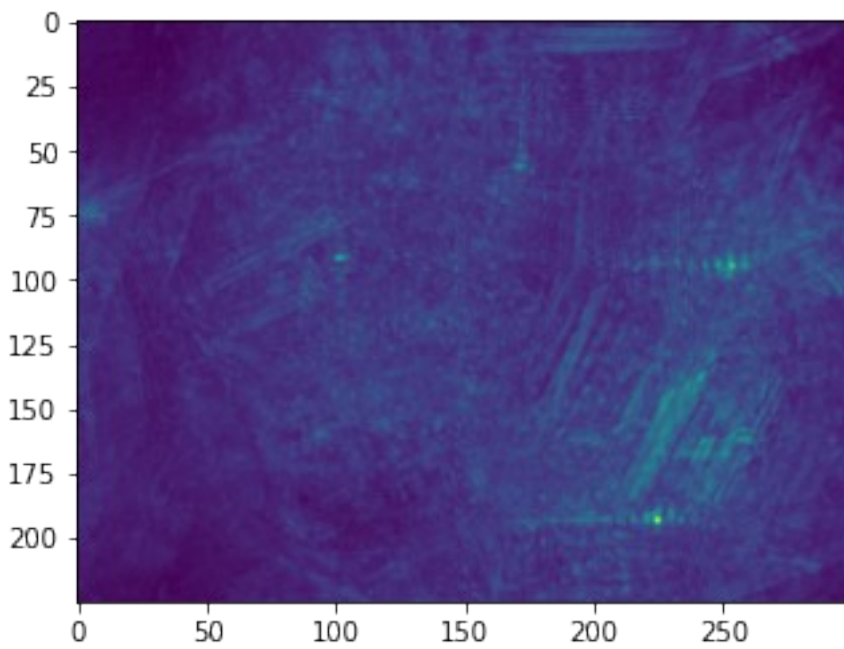
# extract votes (correlation) and parameters
votes, thetas, s = zip(*ght)

# Visualize votes
votes = np.stack(votes).max(0)
plt.imshow(votes)
plt.show()

# nonMaxSuppression
votes = nonMaxSuppression(votes, 20)
plt.imshow(votes)
plt.show()

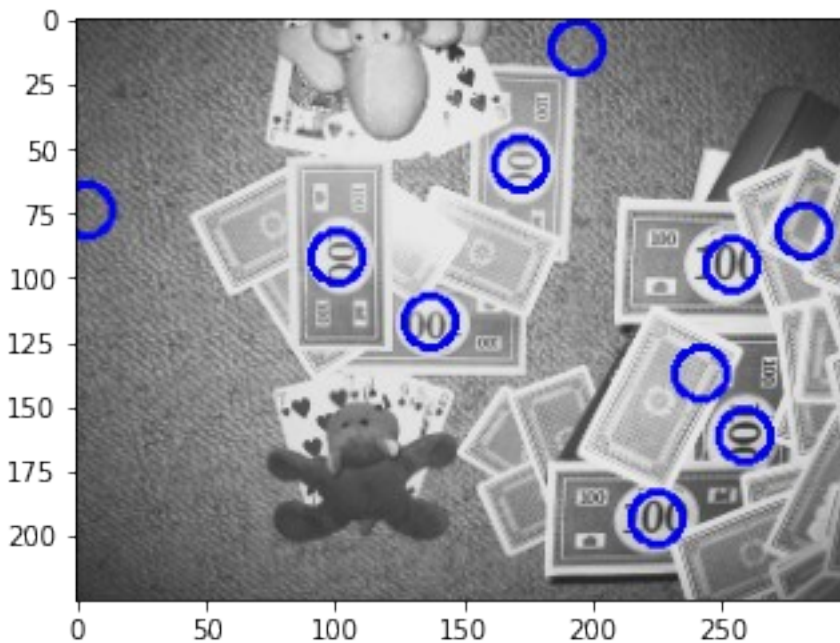
# Visualize n best matches
n = 10
coords = zip(*np.unravel_index(np.argsort(votes, -n, axis=None)[-n:], votes.shape))
vis = np.stack(3*[query], 2)
for y,x in coords:
    print(x,y)
    vis = cv2.circle(vis,(x,y), 10, (255,0,0), 2)
utils.show(vis)
```





194 11
172 56
282 82
137 117
101 92
254 95
242 137
225 193

4 74
259 161



Test your implementation

```
import utils
import cv2
import json
from matplotlib import pyplot as plt
import numpy as np
from sklearn.metrics.pairwise import euclidean_distances

from sklearn.metrics.pairwise import euclidean_distances

def testGHT():
    query = cv2.imread("data/query.jpg", cv2.IMREAD_GRAYSCALE)
    template = cv2.imread("data/template.jpg", cv2.IMREAD_GRAYSCALE)

    angles = np.linspace(0, 360, 36)
    scales = np.linspace(0.9, 1.3, 10)
    ght = GeneralizedHoughTransform(query, template, angles, scales)

    votes, thetas, s = zip(*ght)
    votes = np.stack(votes).max(0)
    plt.imshow(votes)
    plt.show()

    #votes = correlation(query, template)
```



```

votes = nonMaxSuppression(votes, 20)
plt.imshow(votes)
plt.show()

n = 10
coords = list(zip(*np.unravel_index(np.argpartition(votes, -n,
axis=None)[-n:], votes.shape)))

vis = np.stack(3*[query],2)
for y,x in coords:
    vis = cv2.circle(vis,(x,y), 10, (255,0,0), 2)
utils.show(vis)

f = open("centroids.txt", "r")
centroids = f.read()
f.close()

centroids = centroids.split("\n")[:-1]
centroids = [centroid.split() for centroid in centroids]
centroids = np.array([[int(centroid[0]),int(centroid[1])]] for
centroid in centroids])

vis = np.stack(3*[query],2)
for x,y in centroids:
    vis = cv2.circle(vis,(x,y), 10, (255,0,0), 2)
utils.show(vis)

coords = np.array(coords)[:,:,:-1]

d = euclidean_distances(centroids, coords).min(1)

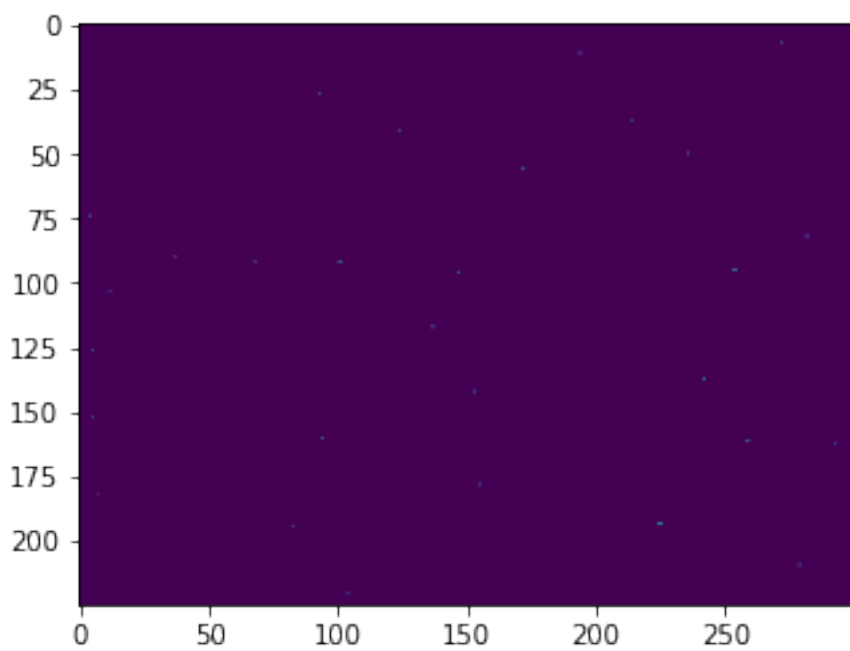
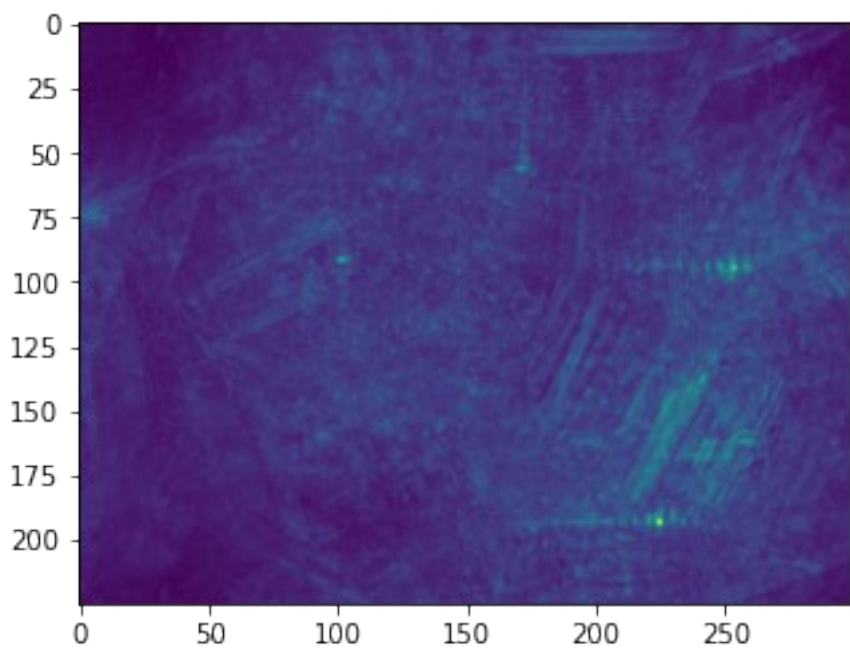
correct_detections = np.count_nonzero((d<10))

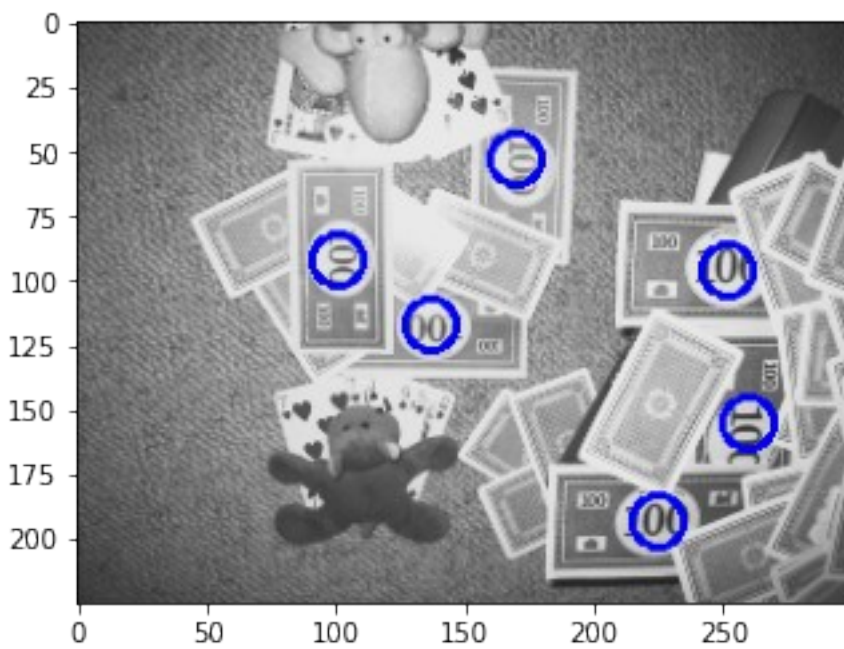
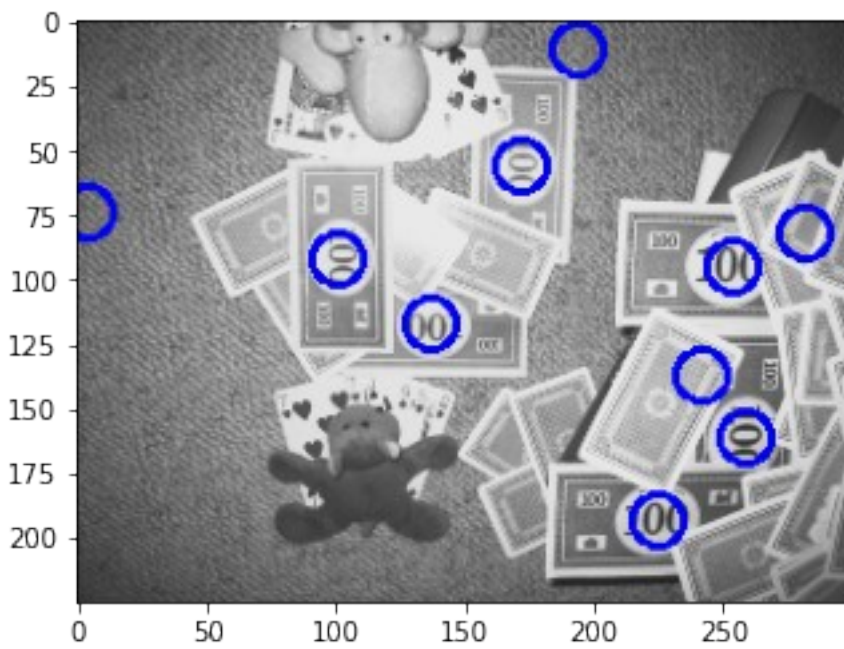
score = { "scores": {"Correct_Detections": correct_detections }}

print(json.dumps(score))

testGHT()

```





{"scores": {"Correct_Detections": 6}}