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watershed.py
import numpy as np
import cv2
from PIL import Image
import math
import matplotlib.mlab as mlab
import matplotlib.pyplot as plt
import random
from numba import jit
@jit#(nopython=True)
def find_neighbours(cent_cords, count):
        # calculating the distance from every good rect centre to every other
        # rect centre in x, y and vector magnitude
        p_to_p_dist = np.zeros((count, count, 3))
        # recording the closest to fourth closest neighbours delt_x and delt_y
        # and vector mag values
        closest_neighbours = np.zeros((count, 4, 3))
        closest_neighbours = closest_neighbours + 3000
        # looping over start points
        for s_p in xrange(0, count):
                # looping over end points
                for e_p in xrange(0, count):
                        if(s_p != e_p):
                                p_to_p_dist[s_p][e_p][0] = cent_cords[s_p][0] - cent_co
rds[e_p][0]
                                p_to_p_dist[s_p][e_p][1] = cent_cords[s_p][1] - cent_co
rds[e_p][1]
                                p_to_p_dist[s_p][e_p][2] = math.sqrt((p_to_p_dist[s_p][
e p][0] ** 2) \
                                         + (p_to_p_dist[s_p][e_p][1] ** 2))
                                if(closest_neighbours[s_p][3][2] > p_to_p_dist[s_p][e_p
][2]):
                                        if(closest_neighbours[s_p][0][2] > p_to_p_dist[
s_p][e_p][2]):
                                                closest_neighbours[s_p][0][0] = p_to_p_
dist[s_p][e_p][0]
                                                closest_neighbours[s_p][0][1] = p_to_p_
dist[s_p][e_p][1]
                                                closest_neighbours[s_p][0][2] = p_to_p_
dist[s_p][e_p][2]
                                        elif(closest_neighbours[s_p][1][2] > p_to_p_dis
t[s_p][e_p][2]):
                                                closest_neighbours[s_p][1][0] = p_to_p_
dist[s p][e p][0]
                                                closest_neighbours[s_p][1][1] = p_to_p_
dist[s_p][e_p][1]
                                                closest_neighbours[s_p][1][2] = p_to_p_
dist[s_p][e_p][2]
                                        elif(closest_neighbours[s_p][2][2] > p_to_p_dis
t[s_p][e_p][2]):
                                                closest_neighbours[s_p][2][0] = p_to_p_
dist[s_p][e_p][0]
                                                closest_neighbours[s_p][2][1] = p_to_p_
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dist[s\_p][e\_p][1]

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                                                closest_neighbours[s_p][2][2] = p_to_p_
dist[s_p][e_p][2]
                                        else:
                                                closest_neighbours[s_p][3][0] = p_to_p_
dist[s_p][e_p][0]
                                                closest_neighbours[s_p][3][1] = p_to_p_
dist[s_p][e_p][1]
                                                closest_neighbours[s_p][3][2] = p_to_p_
dist[s_p][e_p][2]
        return closest_neighbours
def output(pix_scale_omni, angle_omni, hor_neigh_count, \
         ver_neigh_count):
        pix_scale_std_err = np.std(pix_scale_omni) / math.sqrt(pix_scale_omni.size)
        pix_scale_mean = np.mean(pix_scale_omni)
        angle_std_err = np.std(angle_omni) / math.sqrt(angle_omni.size)
        angle_mean = np.mean(angle_omni)
        return angle_mean, angle_std_err, pix_scale_mean, pix_scale_std_err
def error_minimize(filename, mode=0):
        # iterates over values of C to find the best angle and pix_scale disagreement
        max C = 15
        length = 2 * max_C
        min_good_rects = 10
        # angle_mean, angle_std_err, pix_scale_mean, pix_scale_std_err, count, cent_cor
ds
        # good_rect_log
        data_out = np.zeros((5, length))
        for C in xrange(-max_C, max_C):
                data_out[0][C + max_C], data_out[1][C + max_C], data_out[2][C + max_C],
                        data_out[3][C + max_C], data_out[4][C + max_C], cent_cords, goo
d_rect_log = main(filename, C)
                if(np.count_nonzero(good_rect_log) < min_good_rects):</pre>
                        data_out[1][C + max_C] = float('nan')
                        data_out[3][C + max_C] = float('nan')
        angle_std_err = data_out[1,:]
        pix_scale_std_err = data_out[3,:]
        rects = data_out[4,:]
        ang_weight = 1 / np.nanmean(angle_std_err)
        pix_weight = 1 / np.nanmean(pix_scale_std_err)
        avrg_err = (angle_std_err * ang_weight + pix_scale_std_err * pix_weight) / \
                (ang_weight + pix_weight)
        min_angle_err_C = np.nanargmin(angle_std_err) - max_C
        min_pix_err_C = np.nanargmin(pix_scale_std_err) - max_C
        min_avrg_err_c = np.nanargmin(avrg_err) - max_C
        max_rects_C = np.nanargmax(rects) - max_C
        if(mode == 0):
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C = min\_avrg\_err\_c

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        elif(mode == 1):
                C = max_rects_C
        else:
                print('error, incorrect mode, must be either 0 or 1')
                return 0
        data = data_out[:,C + max_C]
        data_out[0][C + max_C], data_out[1][C + max_C], data_out[2][C + max_C], \
                data_out[3][C + max_C], data_out[4][C + max_C], cent_cords, good_rect_1
og = main(filename, C)
        print('best C = ')
        print(min_avrg_err_c)
        return data, cent_cords, good_rect_log
# when flag == '-p' image threshold and fitted rectangles will be displayed
def main(filename, C, flag='-n'):
        # thresholding parameters, iterate over C perhaps, strong effect on angle disag
reement
        #C = -10 # background subtraction when thresholding, 2 reccomended value
        kern_dims = 21 # size of gaussian blurring kernal, default 21
        adapt_box_size = 40 # size of adaptive threshold box, good value of 40
        # rectangle sifting parameters
        min_area = 2500.0 # 2500 - 2700 default
        max_area = 3700.0 # 3700 default
        max_side_ratio = 1.1 # max accepted value of longer side divided by shorter sid
        # spacing of city blocks in mm
        horizontal_spacing = 2.2
        vertical spacing = 2.5
        # creating random colours for circles representing rectangle points
        R = random.random()
        G = random.random()
        B = random.random()
        \#print('R = \%f, G = \%f, B = \%f') \% (R, G, B)
        input_filename = filename
        pil_image = Image.open(input_filename).convert('RGB')
        open_cv_image = np.array(pil_image)
        gray = cv2.cvtColor(open_cv_image,cv2.COLOR_BGR2GRAY)
        # gaussian blur input image
        gray = cv2.GaussianBlur(gray, (kern_dims, kern_dims), 0)
        if((adapt_box_size % 2) == 0):
                adapt box size += 1
        # performs adaptive thresholding of image, C = 3 by default
        thresh = cv2.adaptiveThreshold(gray.astype(np.uint8), 255, \
                cv2.ADAPTIVE_THRESH_MEAN_C, cv2.THRESH_BINARY, adapt_box_size, C)
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# finds OTSU binarisation of image
        #ret, thresh = cv2.threshold(gray,0,255,cv2.THRESH_BINARY_INV+cv2.THRESH_OTSU)
        #thresh = np.invert(thresh)
        # saving hard copy of thresh to avoid damage from findcontours
        thresh_perm = np.copy(thresh)
        # find array of contours
        (cnts, ) = cv2.findContours(thresh, cv2.RETR TREE, cv2.CHAIN APPROX SIMPLE)
        shape = np.shape(cnts)
        length = int(shape[0])
        areas = np.zeros(length)
        side_lengths = np.zeros((length, 4))
        cent\_cords = [0, 0]
        #cent_cords = np.zeros((length, 2))
       side_ratio = np.zeros(length)
        # reduced array containing only coordinates of
        count = 0
        # cycle over contours
        for i in xrange(0,length):
                rect = cv2.minAreaRect(cnts[i])
                box = cv2.cv.BoxPoints(rect)
                box = np.int0(box)
                for side in xrange(0,4):
                        if(side == 3):
                                side_plus = 0
                        else:
                                side_plus = side + 1
                        side_lengths[i][side] = math.sqrt((box[side][0] - box[side_plus
][0]) \
                                ** 2 + (box[side][1] - box[side_plus][1]) ** 2)
                areas[i] = side_lengths[i][0] * side_lengths[i][1]
                # ensure no zero area rectangles are further processed
                if((side_lengths[i][0] != 0) & (side_lengths[i][1] != 0)):
                        side_ratio[i] = side_lengths[i][0] / side_lengths[i][1]
                        if(side_ratio[i] < 1):</pre>
                                side_ratio[i] = 1.0 / side_ratio[i]
                        if(areas[i] > 5000):
                                areas[i] = 0
                        if((areas[i] > min area) & (areas[i] < max area) & (side ratio[</pre>
i] \
                                < max_side_ratio)):
                                count += 1
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                                delt_x_1 = box[0][0] - box[1][0]
                                delt_y_1 = box[0][1] - box[1][1]
                                delt_x_2 = box[1][0] - box[2][0]
                                delt_y_2 = box[1][1] - box[2][1]
                                delt_x = (delt_x_1 / 2) + (delt_x_2 / 2)
                                delt_y = (delt_y_1 / 2) + (delt_y_2 / 2)
                                cent cords x = box[0][0] - delt x
                                cent_cords_y = box[0][1] - delt_y
                                cent_cords_vec = [cent_cords_x, cent_cords_y]
                                cent_cords = np.vstack((cent_cords, cent_cords_vec))
                                cv2.circle(open_cv_image, (int(cent_cords[count][0]), \
                                         int(cent_cords[count][1])), 40, (B * 255, G * 2
55, R * 255), 2)
                                #cv2.drawContours(open_cv_image,[box],0,(0,0,255),2)
        closest_neighbours = find_neighbours(cent_cords, count)
        # number of horizontal neighbours
        hor_neigh_count = 0
        # number of vertical neighbours
        ver_neigh_count = 0
        # attributes of horizontal neighbours
        hor_neigh = []
        # attributes of vertical neighbours
        ver_neigh = []
        # recording the pix scale implied by the horizontal and vertical spacing of cit
y blocks
        pix_scale_hor = []
        pix_scale_ver = []
        # recording the angles of connections between neighbouring points
        angle_hor = []
        angle_ver = []
        good_rect_log = np.zeros(count)
        # iterating over starting point
        for s_p in xrange(1,count):
                # iterating over both first to fourth closest neighbour
                for neighbour in xrange(0,4):
                        if((closest_neighbours[s_p,neighbour,2] > 75) & \
                                (closest_neighbours[s_p,neighbour,2] < 84)):</pre>
                                good_rect_log[s_p] = 1
                                pix_scale_hor.append(closest_neighbours[s_p,neighbour,2
] \
                                         / horizontal_spacing)
                                angle = math.atan(closest_neighbours[s_p,neighbour,1] \
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/ closest\_neighbours[s\_p,neighbour,0])

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                                angle = math.degrees(angle)
                                angle_hor.append(angle)
                                hor neigh count += 1
                        if((closest_neighbours[s_p,neighbour,2] > 88) & \
                                (closest_neighbours[s_p,neighbour,2] < 94)):</pre>
                                good_rect_log[s_p] = 1
                                pix_scale_ver.append(closest_neighbours[s_p,neighbour,2
] / \
                                 vertical spacing)
                                angle = math.atan(closest_neighbours[s_p,neighbour,0] /
                                        closest_neighbours[s_p,neighbour,1])
                                angle = math.degrees(angle)
                                angle_ver.append(-angle)
                                ver_neigh_count += 1
        pix_scale_omni = np.asarray(pix_scale_hor + pix_scale_ver)
        angle_omni = np.asarray(angle_hor + angle_ver)
        angle_mean, angle_std_err, pix_scale_mean, pix_scale_std_err = \
                output(pix_scale_omni, angle_omni, \
                hor_neigh_count, ver_neigh_count)
        # if -p is given as a flag output is printed
        if(flag == '-p'):
                print(cent_cords)
                print('%d rectangles drawn') % count
                #filename = ('./vision/adaptive_test_kern_%d.png') % kern_dims
                #image = Image.fromarray(open_cv_image)
                #image.save(filename)
                #print('image saved to disk')
                #cv2.imshow('city block centres', thresh_perm)
                cv2.imshow('city block centres', open_cv_image)
                # loading array into image and saving
                image_out = Image.fromarray(np.uint8(open_cv_image))
                image_out.save('./images/rect_fit.png')
                if cv2.waitKey(0) & 0xff == 27:
                    cv2.destroyAllWindows()
        # angle_mean, angle_std_err, pix_scale_mean, pix_scale_std_err, count, cent_cor
ds
        # good_rect_log
        #data_out = [angle_mean, angle_std_err, pix_scale_mean, pix_scale_std_err, coun
t, cent_cords, good_rect_log]
        return angle_mean, angle_std_err, pix_scale_mean, pix_scale_std_err, count, cen
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t\_cords, good\_rect\_log