

Supplementary Material

Towards individual-based pollination ecology: Automatic tracking of life histories of individual flowers

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# -*- coding: utf-8 -*-
"""
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### Multi flower tracking algorithm ###
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Detections are in the format: x_{min} , y_{min} , x_{max} , y_{max}

Max distance: Threshold for maximum distance between point and track before initiation of a new track is forced

Max dissapeared: Max number of frames a track can be disappeared before the track is terminated and a new point in that area are considered new tracks

Running mean: Number of previous frame on which to calculate track position (to calculate distance between point and track). If set to one, position is centroid of previous track, if above one, means of x and y are calculated.

The centroid tracking approach was based on: <https://pyimagesearch.com/2018/07/23/simple-object-tracking-with-opencv/> (Accessed on 2022-06-10)

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### Pseudocode of tracking algorithm: ###
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-Go to next frame
-If the frame has no detections
  -Add one to disappeared counter for any existing tracks
-If the frame has detections
  -If we are not tracking objects
    -Initiate tracks for all points
  -If we are tracking objects, calculate all pairwise distances (If running mean above 1, we use the mean x and y position of the track, else we just use position in the previous frame)
    -For all distances below max distance threshold, try to associate points and tracks (shortest distance from point to object get associated and removed, then the second shortest distance etc.)
    -For points with distances above max distance
      -Initiate new tracks
      -Add one to track id counter
  -For points that did not get associated to an exisiting track
    -Initiate new tracks
    -Add one to track id counter
  -Add frame tracking data to result dataframe for final output
  -Update running mean dictionary with new data
    -For tracks that are at the running mean threshold, remove oldest position and add current
  -For tracks that did not get a point associated to them, add one to disappeared counter
  -For tracks that exceed disappeared threshold, remove from active tracking storage
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import pandas as pd # Replace with cudf if performance is too slow?
from collections import OrderedDict
import numpy as np
from scipy.spatial import distance as dist
import matplotlib.pyplot as plt
import time
from statistics import mean
from datetime import datetime
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br = "\n" # Line break for use in code
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# ===== SETTINGS =====
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verbose = True # Set to True if you want tracking process printed to screen and False if not
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# ===== PROGRAM =====
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class tracker():
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Suppl. 1 Python code for flower tracking algorithm

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def __init__(self, max_disappeared, max_distance, running_mean_threshold, results_filename, frames, detections, verbose):
    self.nextObjectID = 0 # Counter for object ids
    self.objects = OrderedDict() # Dictionary. objectID is the key, centroid is the content
    self.means = OrderedDict() # Dictionary to keep track of running means of object coordinates

    self.disappeared = OrderedDict() # Keeps track of how long an objectID has been lost

    self.detections = detections # Store parameters for use in the class
    self.max_disappeared = max_disappeared
    self.max_distance = max_distance
    self.running_mean_threshold = running_mean_threshold
    self.results_filename = results_filename
    self.frames = frames

    self.tracks = [] # Create a list for storing tracking results as we go

    with open(self.results_filename, 'a') as resultFile: # Write the header of the output file
        header = 'frame,filename,x_min,x_max,y_min,y_max,x_c,y_c,objectID\n'
        resultFile.write(header)

    if verbose:
        print("Here are the detections",br,self.detections,br)

### Functions for tracking ###
def store_tracking_results(self, frame, centroid, objectID):
    self.tracks.append([frame, centroid[0], centroid[1], objectID])
    if verbose:
        print(f'Object ID {objectID} with centroid {centroid} in frame {frame} stored.')

def write_tracks_file(self): # Write tracking data to the final result file
    starttime = time.time()

    with open(self.results_filename, 'a') as resultFile:
        for t in self.tracks:
            frame, x_c, y_c, objectID = t[0], t[1], t[2], t[3]
            filename = self.detections.loc[self.detections['frame'] == frame, 'filename'].iloc[0]
            x_min = self.detections.loc[((self.detections['frame'] == frame) & (self.detections['x_c'] == x_c) & (self.detections['y_c'] == y_c)), 'x_min'].iloc[0]
            x_max = self.detections.loc[((self.detections['frame'] == frame) & (self.detections['x_c'] == x_c) & (self.detections['y_c'] == y_c)), 'x_max'].iloc[0]
            y_min = self.detections.loc[((self.detections['frame'] == frame) & (self.detections['x_c'] == x_c) & (self.detections['y_c'] == y_c)), 'y_min'].iloc[0]
            y_max = self.detections.loc[((self.detections['frame'] == frame) & (self.detections['x_c'] == x_c) & (self.detections['y_c'] == y_c)), 'y_max'].iloc[0]

            resultFile.write(f'{frame},{filename},{x_min},{x_max},{y_min},{y_max},{x_c},{y_c},{objectID}{br}')
        endtime = time.time()
        print(f'Writing done. That took {round(endtime-starttime, 4)} seconds. {br}File saved as: {self.results_filename}{br}')

def get_frame_detections(self, frame): # Get the detections from the current frame
    block = self.detections.loc[self.detections['frame'] == frame]
    frame_detections = block[['x_c', 'y_c']] # We just need the centroid, so we'll grab that and return it
    return frame_detections

def register(self, frame, centroid): # Initiate a new track
    if verbose:
        print(f'Registering point with centroid {centroid} in frame {frame}')
    self.objects[self.nextObjectID] = [centroid] # Set the new centroid as content for the new objectID in the Objects dictionary
    self.means[self.nextObjectID] = centroid

    self.disappeared[self.nextObjectID] = 0 # Set number of times the new object has disappeared to zero.

    self.length_dict = {key: len(value) for key, value in self.objects.items()} # For storing how many frames a track has been tracked (to check against running mean setting)
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Suppl. 1 Python code for flower tracking algorithm

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self.store_tracking_results(frame, centroid, self.nextObjectID)

self.nextObjectID += 1 # Add 1 to the objectID counter so it's ready for the next point

if verbose:
    print(f'Current objects: {br}{self.objects}')

def deregister(self, objectID): # Deregister object by deleting it from the objects dict and removing the associated counter from the disappeared dict.
    del self.objects[objectID]
    del self.disappeared[objectID]
    del self.means[objectID]

def update_object(self, objectID, centroid): # Updating the dictionary storing the object centroids

    if verbose:
        print(f'Received in update {br} Object id: {objectID} {br}Centroid: {centroid}{br}')

    if len(self.objects[objectID]) < self.running_mean_threshold:
        if verbose:
            print(f'Length ({len(self.objects[objectID])}) is less than running mean threshold ({self.running_mean_threshold})')
            print(f'Appending {[centroid]} to {self.objects[objectID]}')
        self.objects[objectID].append(centroid)

    if len(self.objects[objectID]) == self.running_mean_threshold:
        if verbose:
            print(f'Length {len(self.objects[objectID])} of {self.objects[objectID]} is equal to running mean threshold.')
            print(f'Deleting first item in {self.objects[objectID]} ({self.objects[objectID][0]}) and appending {centroid}')
        del self.objects[objectID][0]
        self.objects[objectID].append(centroid)

def update_means(self): # Calculate the new means and store
    for key, value in self.objects.items():
        if len(value) > 1:
            c_m = [mean([i[0] for i in value]), mean([i[1] for i in value])]
        if len(value) == 1:
            c_m = value[0]
        self.means[key] = c_m
    if verbose:
        print(f'Updated means dictionary{br}Current mean dict:{br}{self.means}')

def return_tracks_webapp(self): # Only for webapp that is not currently active
    return self.tracks

### Tracking algorithm ###
def track(self, frame): # Start tracking
    frame_detections = self.get_frame_detections(frame) # Get the detections for the current frame
    if verbose:
        print(f'FRAME {frame}. Contains {len(frame_detections)} points.')

    # If the frame has no detections
    if frame_detections.empty: # we will add 1 to disappeared for all objects that are being tracked.
        for objectID in list(self.disappeared.keys()): # loop over any existing tracked objects and mark them as +1 in disappeared
            #print("Object id in disappeared: ", objectID)
            self.disappeared[objectID] += 1

        if self.disappeared[objectID] > self.max_disappeared: # Deregister points that have been disappeared longer than max disappeared threshold
            self.deregister(objectID)

    return self.objects # Return since there is nothing to update

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# If the frame has detections
inputCentroids = frame_detections[['x_c', 'y_c']].values.tolist() # we'll grab the centroid coordinates and convert to a list

if verbose:
    print(f'Input centroids: {br}{inputCentroids}')

if not self.objects: # if Objects is empty, we are currently not tracking any objects, so we'll take the input centroids and register each of them
    for i in range(0, len(inputCentroids)):
        self.register(frame, inputCentroids[i])

if verbose:
    print("Not tracking objects. Initiated tracking on the current points")
    print(f'Current objects:{br}{self.objects}')

else: # We are already tracking objects, so let's see if we can associate any current frame detections with objects that are being tracked.
    objectIDs = list(self.means.keys()) # Store the object IDs and their centroids
    objectCentroids = list(self.means.values())

    print("Object IDs: ", objectIDs)
    print("Object centroids: ", objectCentroids)
    D = dist.cdist(objectCentroids, inputCentroids) # Calculate distances between new points and existing tracks.
    if self.max_distance != 0: # If the max_distance has been set to 0, we'll ignore the next step. (Otherwise 0 would force new tracks for each point).
        D[D > self.max_distance] = np.nan # Set the distance to NA for the pairs that have distance above the threshold we have set. This will force the initiation of new tracks for these points.

    if verbose:
        print("We are tracking existing objects.")
        print(f'Current object ids: {objectIDs}')
        print(f' Current object centroids:{br}{objectCentroids}')
        print(f'Input centroids from current frame:{br}{inputCentroids}')
        print(f'Here\'s the distance matrix:{br}{D}')

    objectIndexes = list(range(0, len(D))) # Grab a list of the object indexes (rows of the D matrix).
    inputIndexes = list(range(len(D[0]))) # Grab a list of the input indexes (columns of the D matrix).

    for c in range(len(D[0])): # Loop over the input centroids
        if not np.isnan(D).all(): # Continue if there a still distance values left in the matrix (i.e. not all NA)
            result = np.unravel_index(np.nanargmin(D, axis=None), D.shape) # Find the row,column index of the lowest distance in the matrix
            D[result[0], :] = np.nan # Set the row and column for this element as NA (since the input and object has now been associated and cannot be used again)
            D[:, result[1]] = np.nan

            objectIndexes.remove(result[0]) # Remove the object index from the list since it has now been used
            inputIndexes.remove(result[1]) # Remove the object index from the list since it has now been used

            self.update_object(objectIDs[result[0]], inputCentroids[result[1]]) # Update the object with the new centroid coordinates
            self.store_tracking_results(frame, inputCentroids[result[1]], objectIDs[result[0]]) # And store the tracking information

        else: # All elements in the distance matrix are NA.
            if verbose:
                print("Association based on distance done. Now dealing with points that were not associated.")
            pass

    self.update_means() # Update the dictionary containing the running means of the points

    for o in objectIndexes: # We'll add 1 for the objects that were not associated with a point in the current frame (in dictionary (disappeared) containing the number of frames the tracks have been lost).

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Suppl. 1 Python code for flower tracking algorithm

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objectID = objectIDs[o]
self.disappeared[objectID] += 1
if self.disappeared[objectID] > self.max_disappeared: # Deregister any tracks that have been disappeared for more
frames than the max disappeared threshold.
    self.deregister(objectID)
    if verbose:
        print(f'Deregistering object {objectID}')

for i in inputIndexes: # And we'll initiate new tracks for the points in the current frame that were not associated with an
existing track.
    if verbose:
        print(f'Registering point {i} with the centroid {inputCentroids[i]}')
    self.register(frame, inputCentroids[i])

# ===== RUN =====

# currentTime = datetime.now() # Use this if you need to time-stamp result file
# currentTime=('%02d-%02d-%02d'%(currentTime.hour,currentTime.minute,currentTime.second))

# ===== Run a single round of tracking =====
# t = tracker(4, 500, 5, frames) # Instantiate the class instance and pass in the threshold for max_disappeared and the list of
frames.
# starttime = time.time()
# for f in frames: # Loop over frames and track
#     t.track(f)
# endtime = time.time()
# print(f'Tracking done. That took {round(endtime-starttime, 3)} seconds. That is {round((endtime-starttime)/len(frames), 3)}
seconds per frame.')
# t.write_tracks_file()
# =====

# ===== Plot stuff =====
# plt.scatter(detections['x_c'], detections['y_c'], c = detections['frame'])
# plt.plot(detections['x_c'],detections['y_c'])

# tracks = pd.read_csv(resultFilename) # Careful not to write several times to same file!
# print(tracks)
# print(tracks.objectID.unique())
# print(f'Number of tracks found: {len(tracks.objectID.unique())}')

# =====

### END OF SCRIPT ###

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Suppl. 2: Results of the filtering algorithm for filtering distances (eps) 50-500 with a stepsize of 50 for the three series.

Eps	Total mismatches	Tracks kept
50	158	165
100	103	147
150	71	124
200	47	100
250	30	69
300	9	49
350	0	28
400	0	18
450	0	16
500	0	11