

{ Google Convention Guide
[PEP-8] on python programming }

Car Tracking Algorithm

{ Can this code 'track' multiple
objects ? }

{ Can this code work on live system
by only changing 'get-frames' method ? }

- what if this code is to be used in online system? live video feed?



counter = 0
interval =

FirstImagePoints
= given Points
= initialized - PPs

[# of images, Width, height, channels]

Algorithm 1 Tracking

1: **Input Parameters:**

2: AllImages: (List of names of all images we want to perform tracking on.)

3: BoundingBox: (A box around car.) *frame?*

4: IntervalSize: (Interval Size to calculate forward and backward flow error.)

5: GeometricThreshold: (Threshold to filter points on base of geometric distance.)

6: ForwardZeroFlowThreshold: (Threshold to filter points on base of zero flow while moving forward.)

7: BackwardZeroFlowThreshold: (Threshold to filter points on base of zero flow while moving backward.)

8: DescriptorThreshold: (Threshold to filter points on base of descriptor matching.)

9: **Output:**

10: TrackedGoodPoints: (The list of list of good points which we track)

11: **My Algorithm:**

12: FinalGoodPoints = []

13: ImageCounter ← 0

14: loop: on len(AllImages)-1

15: Images ← GetFrames(AllImages, ImageCounter, IntervalSize)

16: FirstImagePoints ← GetPoints(BoundingBox, Images[0], method = 'qCorners')

→ KLT corners

17: OpticalFlowForward, TrackedFirstImagePoints

GetForwardFlow(Images, FirstImagePoints, ForwardZeroFlowThreshold)

18: OpticalFlowBackward, TrackedFirstImagePoints

GetBackwardFlow(Images, OpticalFlowForward, TrackedFirstImagePoints, BackwardZeroFlowThreshold)

19: GoodPoints ← GetGoodPoints(TrackedFirstImagePoints, OpticalFlowForward, OpticalFlowBackward, ssd)

20: TrackedGoodPts ← TrackGoodPoints(Images, GoodPoints)

21: FinalGoodPoints.append ← TrackedGoodPts

22: Flow, FlowPoints ← CalculateFlow(Images[ImageCounter], Images[ImageCounter], GoodPoints)

23: RansacFlow = GetRansac(Flow)

24: BoundingBox ← UpdatebyRansacFlowvector

25: **if** ImageCounter == len(AllImages) - 2 **then**

26: GoodPointsList ← FlowPoints

27: FinalGoodPoints.append ← GoodPointsList

28: ImageCounter ++

29: **endloop:** on len(AllImages)

30: **return** FinalGoodPoints

being called
= # of images

images
image path

One image will be read multiple times

Complete
time = # of images $\times K \times O(\text{Image Read time})$

Error handling

Ideal
time = # of images $\times O(\text{Image Read time})$



Algorithm 2 GetFrames

- 1: **Input Parameters:**
- 2: *ListofImagesName*: (List contains the path to every image in ascending order)
- 3: *FirstImageIndex*: (starting index to read images from list.)
- 4: *IntervalSize*: (Specifies how many frames to read)
- 5: **Output:**
- 6: *Images*: (List of Images.)
- 7: *Status*: (1 for successful and -1 in case of no frame found)

time based

8: **My Algorithm:**

error(' -ve indices')

```

9: if FirstImageIndex or IntervalSize < 0 then return Images, status ← -1
10: else
11:   a ← FirstImageIndex + IntervalSize
12:   loop:
13:     if FirstImageIndex < a then
14:       if FirstImageIndex > len(ListofImagesName) then return
15:       else
16:         Img ← read(ListofImagesName[FirstImageIndex]) →
17:         Images.append ← Img
18:         FirstImageIndex++
19:     Endofloop:
20:     if len(Images) == 0 then return Images, status ← -1
21:   else
22:     return Images, status ← 1

```

$K \times O(\text{Image Read time})$

{ interval size }
{ \times image size }



Algorithm 3 GetPoints *(Initialization)*

```

1: Input Parameters:
2: BoundingBox: (This contains the starting and ending index of box.  $[xmin, ymin, xmax, ymax]$ .)
3: Image: (Image of which we want to get points.)
4: Method: (Method which specify, how to get Image Points)
5: Output:
6: Points: (List of points in range of given box)
7: Status: (1 for successful and -1 in case of no points found)

8: My Algorithm:

9:  $Status = -1$ 
10: if  $Method == "grid"$  then
11:    $Points \leftarrow \text{points between}(xmin, ymin) \text{ and } (xmax, ymax)$ 
12:   return  $Points, status \leftarrow 1$ 
13: if  $Method == "gCorners"$  then
14:    $patch = im[ymin : ymax, xmin : xmax]$ 
15:    $corners \leftarrow \text{goodFeaturesToTrack}(patch, 1000, 0.0001, 1, \text{useHarrisDetector} =$ 
      $True, k = 0.001$ )
16:    $corners \leftarrow corners.reshape((-1, 2))$ 
17:    $corners[:, 0], corners[:, 1] \leftarrow corners[:, 0] + xmin, corners[:, 1] + ymin$ 
18:   return  $Points \leftarrow corners, status \leftarrow 1$ 

```

Resolution

{numpy arrays}

Zero flow
till 'K'
frames

K \neq # of frames
for PB tracking

Single call
for multi-objects

Algorithm 4 GetForwardFlow

1: **Input Parameters:**

2: *Images:* (List of Images on which we want to compute forward flow.)

3: *FirstImageFeaturePoints:* (This list contains the features points on first image, from where we want to compute forward flow.)

4: *ThresholdZeroFlow:* (This value specify the threshold value to filter background points from object points)

5: **Output:**

6: *FlowPointsAtLastImage:* (Flow points at the First frame.)

7: *FirstImageFeaturePoints:* (Given features point, which we want to track at the first image)

8: *Status:* (1 for successful and -1 in case of no points found after zero flow filter)

9: **My Algorithm:**

10: $Status = -1$

11: $a \leftarrow 0$

12: loop: till $len(Images)$:

13: $a < len(Images)$:

14: $next \leftarrow a++$

15: $currentImage-Gray \leftarrow cv2.cvtColor(Images[a], cv2.COLOR_BGR2GRAY)$

16: $nextImage-Gray \leftarrow cv2.cvtColor(Images[next], cv2.COLOR_BGR2GRAY)$

17: $ForwardPoints, st, err \leftarrow cv2.calcOpticalFlowPyrLK(currentImage-Gray, nextImage-Gray, FirstImageFeaturePoints.astype(np.float32), None, **lk_params)$

18: **if** $a \leftarrow (len(Images) - 2)$ **then**

19: $ForwardPoints, GivenFirstImagePoints$

$\leftarrow CalculateZeroFlow(ForwardPoints, GivenFirstImagePoints, ThresholdZeroFlow)$

20: **endofloop**

21: **if** $len(ForwardPoints) == 0$ **then return**

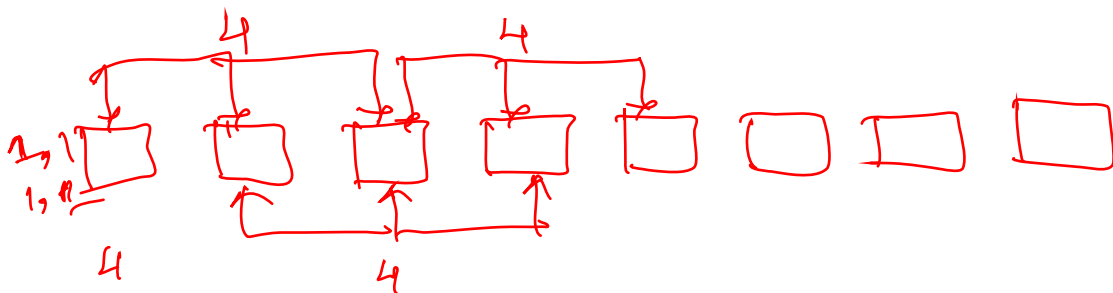
$ForwardPoints, FirstImageFeaturePoints, Status \leftarrow -1$

22: **else**

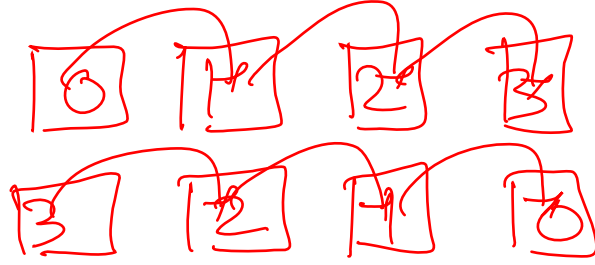
23: **return** $ForwardPoints, FirstImageFeaturePoints, Status \leftarrow 1$

ideal
time

color2gray conversion
 $= O(\text{one image conversion time}) \times \# \text{ of images}$



interval = 3



Isn't this same as Forward Flow

Algorithm 5 GetBackwardFlow

1: **Input Parameters:**

2: *Images:* (List of Images on which we want to compute Backward flow.)

3: *FirstImageFeaturePoints:* (This list contains the features points on last image, from where we want to com

4: *ThresholdZeroFlow:* (This value specify the theshold value to filter background points from object points)

5: **Output:**

6: *FlowPointsAtLastImage:* (Flow points at the last frame. We get from ForwardFlow method)

7: *FirstImageFeaturePoints:* (Given features point, which we want to track at the first image)

8: *Status:* (1 for successful and -1 in case of no points found after zero flow filter)

9: **My Algorithm:**

10: $Status = -1$

11: $a \leftarrow \text{len}(\text{Images}) - 1$

12: *loop:* till $\text{len}(\text{Images})$:

13: $a > 0$:

14: $next \leftarrow a - -$

15: $currentImage-Gray \leftarrow \text{cv2.cvtColor}(\text{Images}[a], \text{cv2.COLOR_BGR2GRAY})$

16: $nextImage-Gray \leftarrow \text{cv2.cvtColor}(\text{Images}[next], \text{cv2.COLOR_BGR2GRAY})$

17: $ForwardPoints, st, err \leftarrow \text{cv2.calcOpticalFlowPyrLK}(currentImage - Gray, nextImage-Gray, FirstImageFeaturePoints.astype(np.float32), None, **lk_params)$

18: **if** $a \leftarrow 1$ **then**

19: $ForwardPoints, FirstImageFeaturePoints$

$\leftarrow \text{CalculateZeroFlow}(ForwardPoints, GivenFirstImagePoints, ThresholdZeroFlow)$

20: *endofloop*

21: **if** $\text{len}(ForwardPoints) == 0$ **then** **return**

$ForwardPoints, FirstImageFeaturePoints, Status \leftarrow -1$

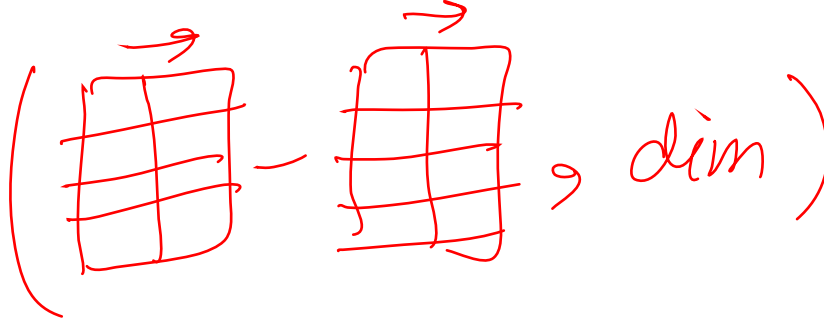
22: **else**

23: **return** $ForwardPoints, FirstImageFeaturePoints, Status \leftarrow 1$

tracked points-in-k

init-points-in-y

= 
norm



Algorithm 6 CalculateZeroFlow

- 1: **Input Parameters:**
- 2: *LastImagePoints*: (Points at the last frame after optical flow.)
- 3: *FirstImagePoints*: (Given Points at the first image)
- 4: *Threshold*: (Threshold to filter points on base of zero flow)
- 5: **Output:**
- 6: *FilteredLastImagePoints*: (Last Image point after zero flow filter)
- 7: *FilteredFirstImagePoints*: (First Image point after zero flow filter)

8: **My Algorithm:**

- 9: *FilteredLastImagePoints* = []
- 10: *FilteredFirstImagePoints* = []
- 11: $t \leftarrow 0$
- 12: loop: on $\text{len}(\text{FirstImagePoints})$
- 13: **if** $\text{np.linalg.norm}(\text{LastImagePoints}[a] - \text{FirstImagePoints}[a])$
 Threshold **then**
- 14: *FilteredLastImagePoints.append* $\leftarrow \text{LastImagePoints}[t]$
- 15: *FilteredFirstImagePoints.append* $\leftarrow \text{FirstImagePoints}[t]$
- 16: $t++$
- 17: endloop: on $\text{len}(\text{FirstImagePoints})$
- 18: **return** *FilteredLastImagePoints*, *FilteredFirstImagePoints*

vectorize

Algorithm 7 GetGoodPoints*(Apply all filters)*

```

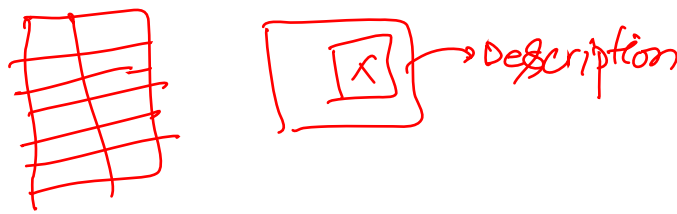
1: Input Parameters:
2: FirstImagePoints: (Features Points at start of image)
3: ForwardPoints: (Given Points tracked to given interval size.)
4: BackwardPoints: (Given forward Points tracked back to first image.)
5: Images: (Set of images of given interval size, on which we want to track points.)
6: GeomaticThreshold: (Threshold to filter points on base of geomatic distance.)
7: DescriptorThreshold: (Threshold to filter points on base of descriptor SSD or Dot Product)
8: Method: (Specify the method to use to compare descriptor)
9: Output:
10: GoodPoints: (Return Good features points which pass criteria of filters. )
11: Status: (Return -1 if no good feature point found otherwise 1.)

12: My Algorithm:

13: varGoodPoints = []
14: varForwardPoints = []
15: FinalGoodPoints = []
16: Status  $\leftarrow$  -1
17: loop: till len(BackwardPoints)
18:   i  $\leftarrow$  0
19:   dist  $\leftarrow$  np.linalg.norm(BackwardPoints[i] - FirstImagePoints[i])
20:   if dist  $\leq$  GeomaticThreshold then
21:     varGoodPoints.append  $\leftarrow$  FirstImagePoints[i]
22:     varForwardPoints.append  $\leftarrow$  ForwardPoints[i]
23:     i ++
24:   Endloop: len(BackwardPoints)
25:   orb  $\leftarrow$  cv2.ORB_create()
26:   FirstImageGray  $\leftarrow$  GrayImageofImages[First]
27:   LastImageGray  $\leftarrow$  GrayImageofImages[Last]
28:   keyPointsGood  $\leftarrow$  KeyPoints(varGoodPoints)
29:   keyPointsFarword  $\leftarrow$  KeyPoints(varFarwordPoints)
30:   Goodkp, GoodptsDes  $\leftarrow$  orb(FirstGrayImage, keyPointsGood)
31:   Farwordkp, FarwordptsDes  $\leftarrow$  orb(LastGrayImage, keyPointsFarword)
32:   if Method == "DotProduct" then
33:     loop: len(varGoodPoints)
34:       t = 0
35:       gdptdes  $\leftarrow$  NormalizedGoodptsDes
36:       forwarddes  $\leftarrow$  NormalizedFarwordptsDes
37:       product  $\leftarrow$  dot(forwarddes, gdptdes)
38:       if product < "DescriptorThreshold" then
39:         FinalGoodPoints.append  $\leftarrow$  varGoodPoints[t]
40:         t ++
41:       Endloop: len(varGoodPoints)
42:   if len(FinalGoodPoints) == 0 then return
   FinalGoodPoints, Status  $\leftarrow$  -1
43:   else
44:     return FinalGoodPoints, Status  $\leftarrow$  1

```

*Vectorize**this method is not mathematically correct*



```

1: if Method == "SSD" then
2:   loop:len(varGoodPoints)
3:   t = 0
4:   gdptdes ← NormalizedGoodptsDes
5:   forwarddes ← NormalizedFarwordptsDes
6:   ssddistance ← np.linalg.norm(forwarddes - gdptdes)
7:   if ssddistance < "DescriptorThreshold" then
8:     FinalGoodPoints.append ← varGoodPoints[t]
9:     t ++
10:   Endloop:len(varGoodPoints)s
11:   if len(FinalGoodPoints) == 0 then return
      FinalGoodPoints, Status ← -1
12:   else
13:     return FinalGoodPoints, Status ← 1

```

Vectorize

Algorithm 8 TrackGoodPoints

1: **Input Parameters:**

2: *Images:* (List of Images on which we want to compute forward flow.)

3: *GoodPoints:* (This list contains the good features points on first image, from where we want to compute forward flow.)

4: **Output:**

5: *TrackedPoints:* (List of good points tracked in given images. The size of list is equal to 2*size of images by 2.)

6: **My Algorithm:**

7: TrackedGoodPoints = []

8: GoodPointsList ← Reshape goodpoints in 2 by length list. At zero index all x-coordinate and at 1 all y-coordinate of good points.

9: TrackedGoodPoints.append ← (GoodPointsList[0])

~~10: TrackedGoodPoints.append ← (GoodPointsList[1])~~

11: a ← 0

12: loop: till len(Images):

13: a < len(Images) :

14: next ← a ++

15: currentImage-Gray ← cv2.cvtColor(Images[a], cv2.COLOR_BGR2GRAY)

16: nextImage-Gray ← cv2.cvtColor(Images[next], cv2.COLOR_BGR2GRAY)

17: ForwardPoints, st, err ← cv2.calcOpticalFlowPyrLK(currentImage - Gray, nextImage-Gray, FirstImageFeaturePoints.astype(np.float32), None, **lk_params)

18: GoodPointsList ← ForwardPoints

19: TrackedGoodPoints.append ← (GoodPointsList[0])

20: TrackedGoodPoints.append ← (GoodPointsList[1])

21: endofloop

return TrackedGoodPoints

(compute flow per frame for each 'good-points')



$\left\{ \begin{array}{l} \text{---} = \text{extract SingleFrame Flow} \\ \text{box_flow} = \text{RANSAC}(\text{---}) \end{array} \right\}$

box $\left\{ \begin{array}{l} \rightarrow \text{Median} \\ \rightarrow \text{RANSAC} \end{array} \right\}$

Algorithm 9 CalculateFlow

1: **Input Parameters:**

2: *CurrentImage*: (Current Image from which we want to calculate flow)

3: *NextImage*: (Very next image to current image)

4: *GoodPoints*: (Features points on current image to computer flow.)

5: **Output:**

6: *Flow*: (This list contain difference of flow of points from current image to next image)

7: *ForwardPoints*: (This list contain flow of points from current image to next image)

8: **My Algorithm:**

Don't we have it from previous step?

9: *currentImage-Gray* \leftarrow *cv2.cvtColor*(*CurrentImage*, *cv2.COLOR_BGR2GRAY*)

10: *nextImage-Gray* \leftarrow *cv2.cvtColor*(*NextImage*, *cv2.COLOR_BGR2GRAY*)

11: *ForwardPoints*, *st*, *err* \leftarrow *cv2.calcOpticalFlowPyrLK*(*currentImage-Gray*, *nextImage-Gray*, *FirstImageFeaturePoints.astype(np.float32)*, *None*, ***lk_params*)

12: *a* \leftarrow 0

13: *loop*: till *len(ForwardPoints)*

14: *x* \leftarrow *FlowPoints*[*a*][0] - *GoodPoints*[*a*][0]

15: *y* \leftarrow *FlowPoints*[*a*][1] - *GoodPoints*[*a*][1]

16: *Flow.append* \leftarrow (*x*, *y*)

17: *endofloop*:

18: **return** *Flow*, *ForwardPoints*

} Vectorize





Median

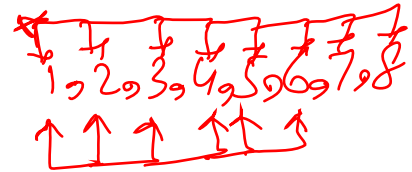
(whether to get flow of best hypothesis or median flow of all inliers)

Algorithm 10 GetRansac

- 1: **Input Parameters:**
- 2: *Flow*: (Flow of good points, we get as output of calculate flow method)
- 3: *Threshold*: (Threshold to filter inliers from outliers)
- 4: **Output:**
- 5: *FlowPoint*: (Return a single point which has the most number of inliers.)

6: My Algorithm:

- 7: *Inliers* = []
- 8: *NormalizedFlowPoints* \leftarrow *Normalizedto1*
- 9: *loop*: till *len(NormalizedFlowPoints)*
- 10: *tempInliers* = []
- 11: *i* \leftarrow *NormalizedFlowPoints*
- 12: *product* \leftarrow *np.dot(NormalizedFlowPoints, i)*
- 13: *loop*: till *len(product)*
- 14: *j* \leftarrow *product*
- 15: **if** *j* \geq *threshold* **then**
- 16: *tempInliers.append* \leftarrow *j*
- 17: *endloop:Product*
- 18: **if** *len(tempInliers)* \geq *len(Inliers)* **then**
- 19: *Inliers* \leftarrow *tempInliers*
- 20: *FlowPoint* \leftarrow *Flow[i]*
- 21: *endofloop:NormalizedFlowPoints*
- 22: **return** *FlowPoint*



$\frac{\# \text{ of points}}{\# \text{ of hypothesis}} \times \# \text{ of points}$

$$8 \times 8 \sim 64$$

$$\left\{ \begin{array}{l} 1000 \times 1000 \sim 10^6 \\ 100 \times 1000 \end{array} \right.$$

- { ① Random Sample

{ ② Evaluate hypotheses'

{ 2a) compute distance

{ 2b) compare

{ ③ Check if this is the best hypothesis