

Optical Flow

Chao Duan, Xuan Yang

1. Basic idea
2. Correlation-based optical flow
3. Differential-based and Dense optical flow

Optical Flow

Basic idea

image1



image2

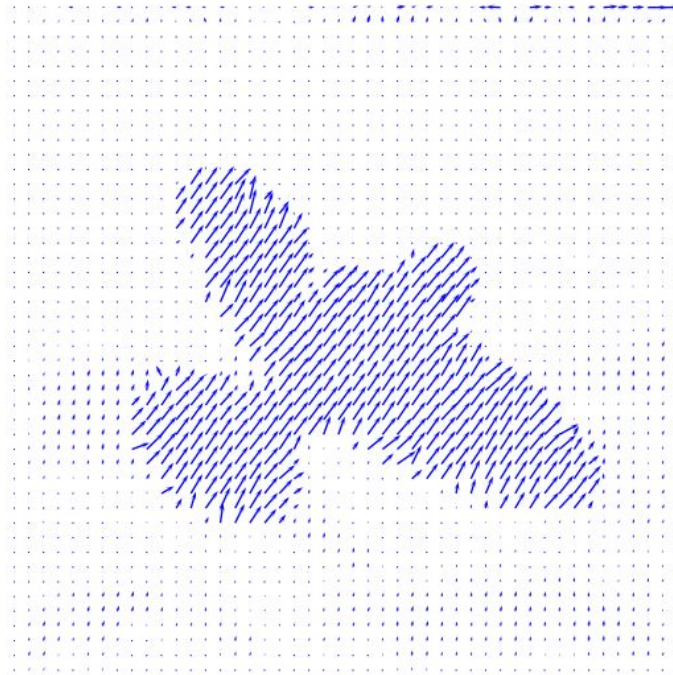


(original image1 and image2 are captured from [1])

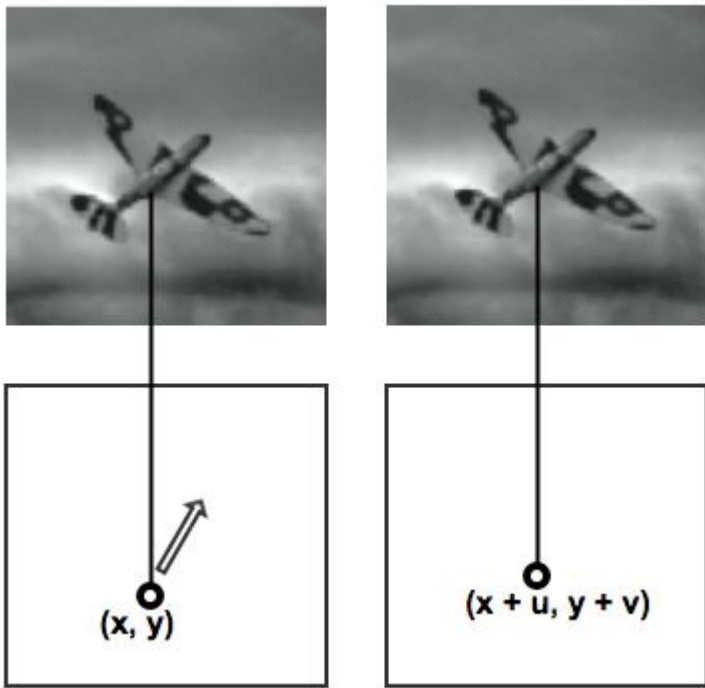
image2-1



Motion detecting result



Basic idea



Important assumption

1. Constant brightness
2. Similar velocity for a block
3. Small displacement

Correlation-based



Correlation



(x, y)

$[(x-d, x+d), [y-d, y+d]]$

$(2d + 1)^2$ possible I_2

i	71	68	63
	66	65	89
	101	82	60
	j		

I_1

Differences measure methods:

SAD(Sum of absolute differences)

$$\sum_{(i,j) \in W} |I_1(i, j) - I_2(dx + i, dy + j)|$$

ZSAD(Zero – mean Sum of absolute differences)

$$\sum_{(i,j) \in W} |I_1(i, j) - \bar{I}_1(i, j) - I_2(dx + i, dy + j) + \bar{I}_2(dx + i, dy + j)|$$

SSD(Sum of squared differences)

$$\sum_{(i,j) \in W} (I_1(i, j) - I_2(dx + i, dy + j))^2$$

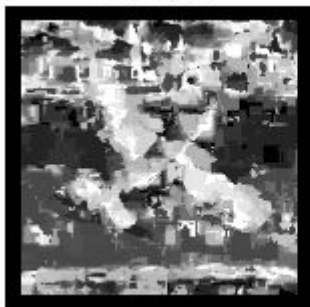
ZSSD(Zero – mean Sum of squared differences)

$$\sum_{(i,j) \in W} (I_1(i, j) - \bar{I}_1(i, j) - I_2(dx + i, dy + j) + \bar{I}_2(dx + i, dy + j))^2$$

Correlation

No Gaussian filter:

SAD 5.48s



ZSAD 7.74s



SSD 4.87s

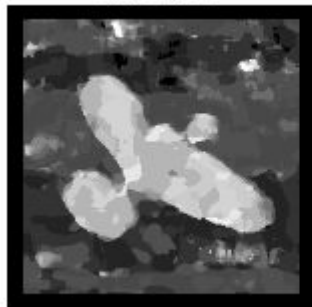


ZSSD 8.51s

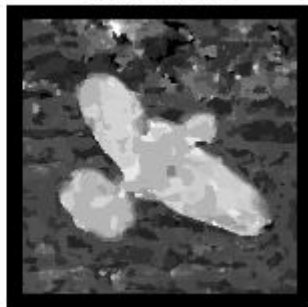


With Gaussian filter:

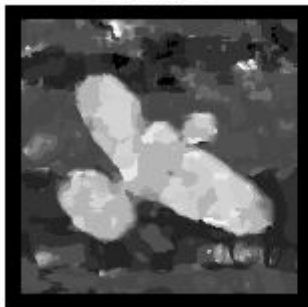
SAD 4.89s



ZSAD 7.92s



SSD 4.55s



ZSSD 8.37s



Correlation

Similarity measure methods:

CC(Cross correlation)

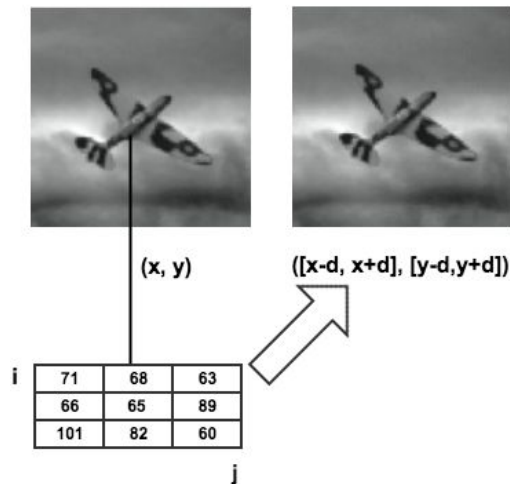
$$\sum_{(i,j) \in W} I_1(i,j) I_2(dx + i, dy + j)$$

NCC(Normalized Cross correlation)

$$\frac{\sum_{(i,j) \in W} I_1(i,j) I_2(dx + i, dy + j)}{\sqrt{\sum_{(i,j) \in W} I_1(i,j)^2 \sum_{(i,j) \in W} I_2(dx + i, dy + j)^2}}$$

ZNCC(Zero – mean Normalized Cross correlation)

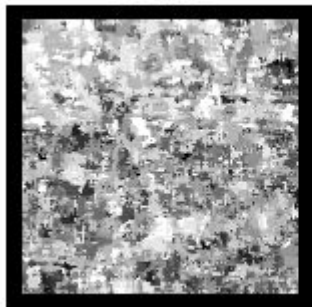
$$\frac{\sum_{(i,j) \in W} (I_1(i,j) - \bar{I}_1(i,j)) (I_2(dx + i, dy + j) - \bar{I}_2(dx + i, dy + j))}{\sqrt{\sum_{(i,j) \in W} (I_1(i,j) - \bar{I}_1(i,j))^2 \sum_{(i,j) \in W} (I_2(dx + i, dy + j) - \bar{I}_2(dx + i, dy + j))^2}}$$



Correlation

No Gaussian filter:

CC 6.23s



NCC 58.10s



ZNCC 19.08s

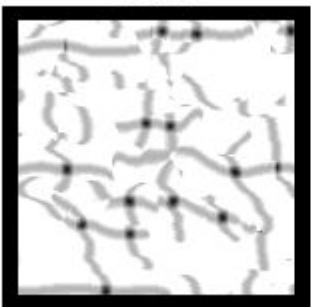


ZNCC (speedup) 10.67s

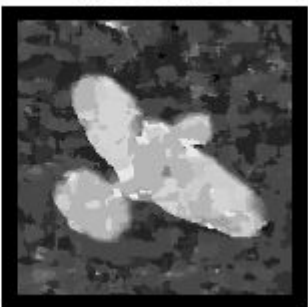


With Gaussian filter:

CC 5.06s



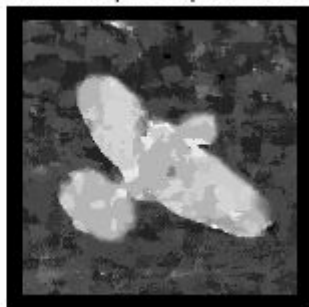
NCC 54.73s



ZNCC 18.02s



ZNCC (speedup) 11.13s



Belive me I am flying!

image1

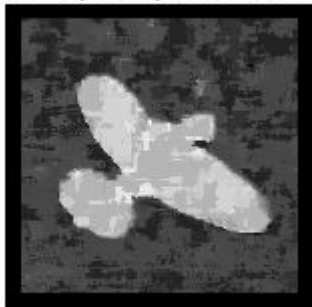


image2

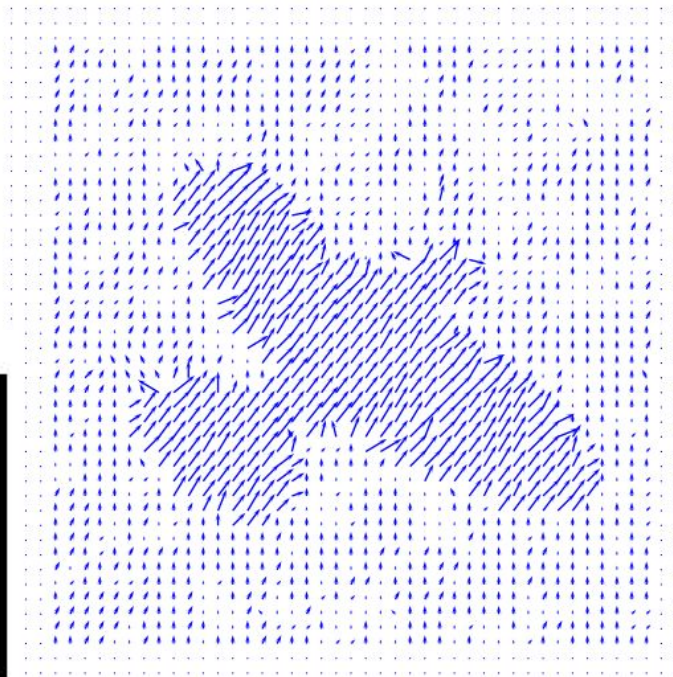
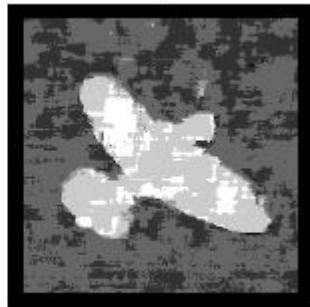


ZNCC(speedup) cost:10.29s

N

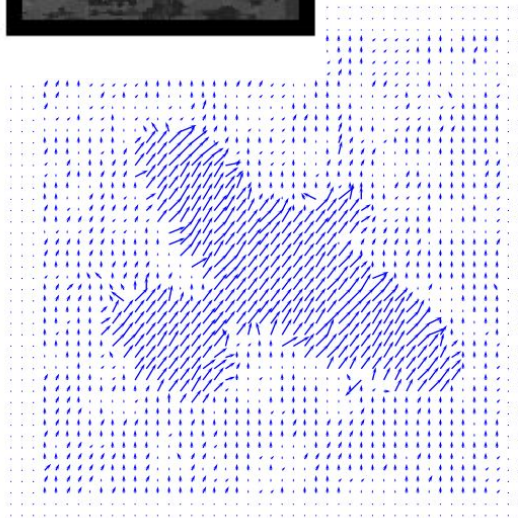
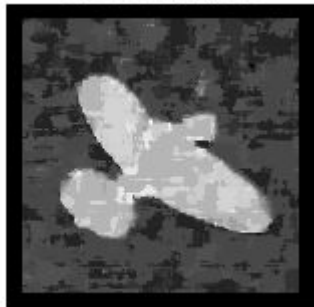


V

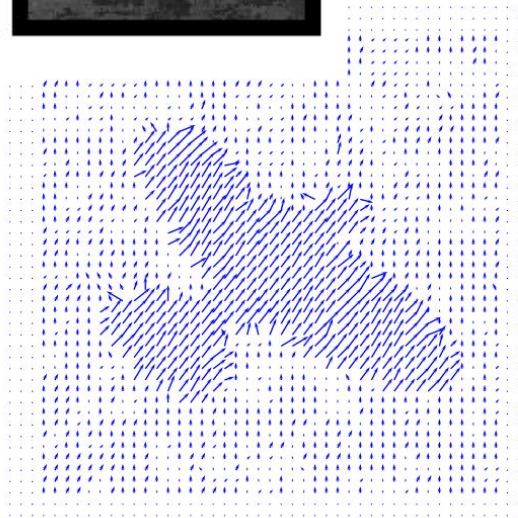


Comparison

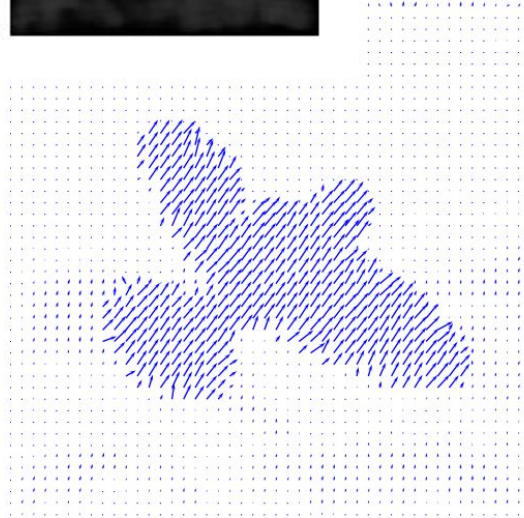
ZSSD cost:8.31s



ZNCC(speedup) cost:10.29s



Farneback(opencv) cost:0.02s



Differential-based

& Dense optical flow

Differential-based Derive

Image brightness constancy[5]:

➡ from a short interval t_1 to t_2 , while an object may change position, the reflectivity and illumination will remain constant.

$$f(x + \Delta x, y + \Delta y, t + \Delta t) \approx f(x, y, t)$$

➡ According to Taylor series expansion:

$$f(x + \Delta x, y + \Delta y, t + \Delta t) = f(x, y, t) \frac{\partial}{\partial x} \Delta x + \frac{\partial}{\partial y} \Delta y + \frac{\partial}{\partial z} \Delta z + h.o.t$$

$$\nabla I \bullet + I_t = 0$$

HORN & SCHUNCK METHOD[6]

use masks to calculate f_x , f_y , f_t :

$$\begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}$$

mask1

$$\begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix}$$

mask2

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

mask3

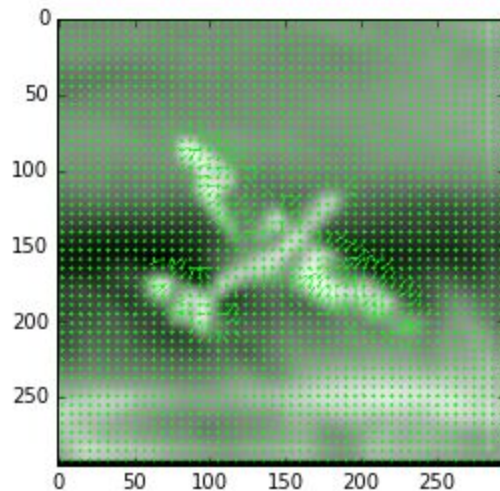
$$\begin{bmatrix} -1 & -1 \\ -1 & -1 \end{bmatrix}$$

mask4

$$e_s = \iint ((u_x^2 + u_y^2) + (v_x^2 + v_y^2)) dx dy$$

$$e_c = \iint (I_x u + I_y v + I_t)^2 dx dy$$

minimize $e = \lambda e_c + e_s$



LUCAS & KANADE METHOD[6]

$$\min \sum_i (f_{xi}u + f_{yi}v + f_t)^2$$

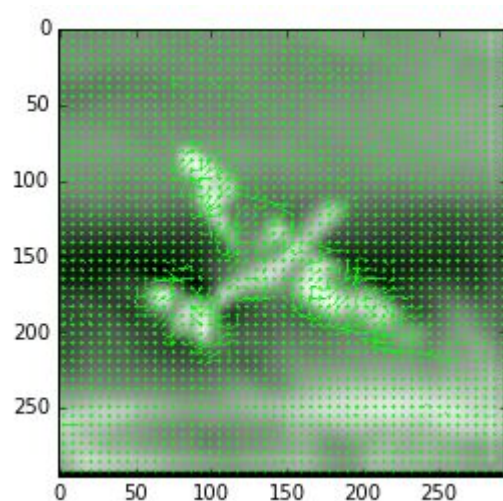
$$\sum (f_{xi}u + f_{yi}v + f_{ti})f_{xi} = 0$$

$$\sum (f_{xi}u + f_{yi}v + f_{ti})f_{yi} = 0$$

$$\sum f_{xi}^2 u + \sum f_{xi}f_{yi}v = -\sum f_{xi}f_{ti}$$

$$\sum f_{xi}f_{yi}u + \sum f_{yi}^2 v = -\sum f_{yi}f_{ti}$$

$$\begin{bmatrix} \sum f_{xi}^2 & \sum f_{xi}f_{yi} \\ \sum f_{xi}f_{yi} & \sum f_{yi}^2 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} -\sum f_{xi}f_{ti} \\ -\sum f_{yi}f_{ti} \end{bmatrix}$$



Gunnar Farneback method

Dense optical flow

_____ A optimal method to track all the pixels in an image

Farneback method uses Polynomial Expansion to approximate the neighbors of a pixel[7].

Demo with Dense techniques

➡ Extract the moving object from the static background[8]

➡ Gaussian filter

➡ Get background with threshold

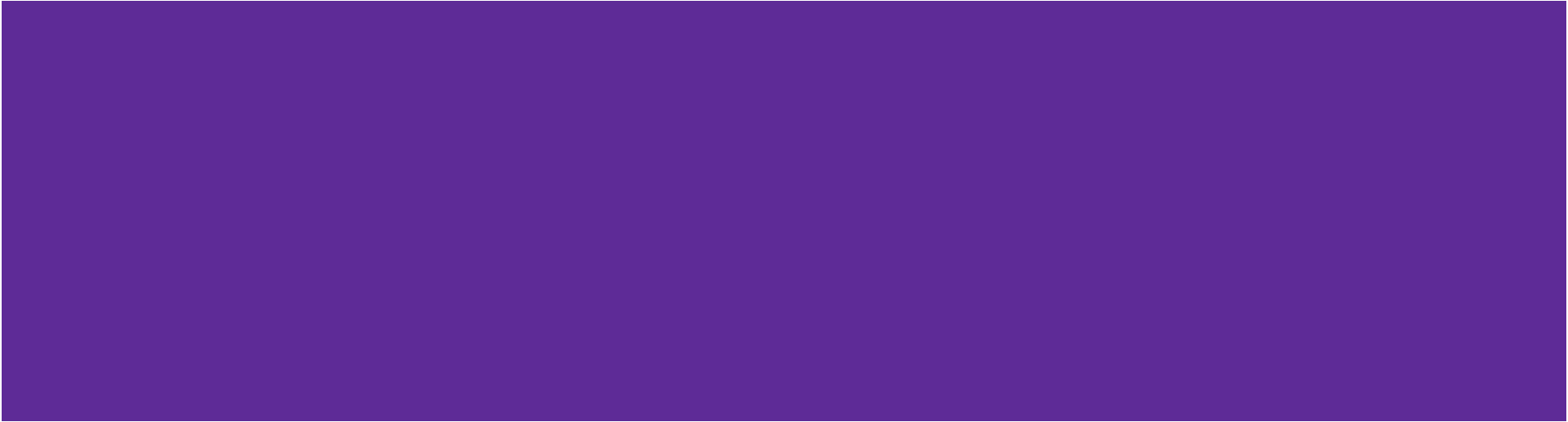
➡ Track the moving object

➡ findContours() in opencv

➡ Draw the rectangle edge of the moving object

Speed up with numba!

Thanks



Reference

- [1] Nixon M. Feature extraction & image processing[M]. Academic Press, 2008.
- [2] Giachetti, A., Matching Techniques to Compute Image Motion, Image Vision Comput., 18(3), pp. 247–260, 2000
- [3] Correlation based similarity measures-Summary, <https://siddhantahuja.wordpress.com/tag/normalized-cross-correlation/>
- [4] Zero Mean Normalized Cross-Correlatio, <https://martin-thoma.com/zero-mean-normalized-cross-correlation/#tocAnchor-1-2>
- [5] O'Donovan, Peter. "Optical flow: Techniques and applications." The University of Saskatchewan (2005).
- [6] Coarse-to-fine Optical Flow, <http://eric-yuan.me/coarse-to-fine-optical-flow/>
- [7] The Gunnar-Farneback optical flow, <https://www.safaribooksonline.com/library/view/opencv-essentials/9781783984244/ch07s04.html>
- [8] SuganyaDevi, K., N. Malmurugan, and R. Sivakumar. "Efficient foreground extraction based on optical flow and SMED for road traffic analysis." *International Journal of Cyber-Security and Digital Forensics (IJCSDf)* 1.3 (2012): 177-182.