EECS 442 Discussion

09/27/2017

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Announcements

HW 1 is due tomorrow.

Some of the problem needs a submission of original .m files together with the report. Pack them into a single file.

Comment your code and make graders' life easier.

• Quiz tomorrow.

Covers Tuesday's lecture.

Short survey at the end of the discussion.

Provide your feedback and help me do better ©

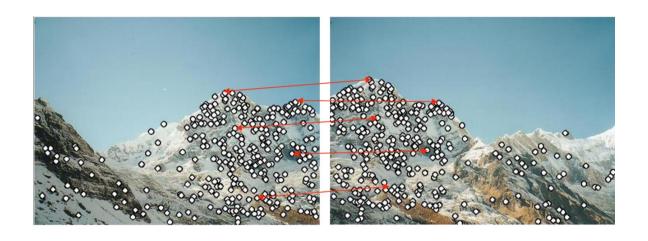
Today's Topics

- Structure tensor and Harris operator
- Application of feature detection: image stitching









- 1. Find feature points in both images. -> structure tensor Reduction
- 2. Match the feature points into pairs. -> matching algorithm, e.g.,
- Hungarian Matching

 3. Use pairs to align two images. -> homography Estimation

I
$$(x, y) \rightarrow (x + u, y + v)$$

 $I(x + u, y + v) = I(x, y) + \frac{\partial I(x, y)}{\partial x}u + \frac{\partial I(x, y)}{\partial y}v + \cdots - \cdots$
 $E = \sum_{(x,y) \in W} \left(I(x + u, y + v) - I(x, y)^{2} \right)$
 $\Rightarrow \sum_{(x,y) \in W} \left(I(x + u, y + v) - I(x, y)^{2} \right)$
 $\Rightarrow \sum_{(x,y) \in W} \left(I(x, y) + \frac{\partial I(x, y)}{\partial x}u + \frac{\partial I(x, y)}{\partial y}v - I(x, y) \right)^{2}$
 $= 2 \left(I(x, y) + \frac{\partial I(x, y)}{\partial x}u + \frac{\partial I(x, y)}{\partial y}v - I(x, y) \right)^{2}$
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$$H = \begin{bmatrix} h_{11} & h_{12} \\ h_{21} & h_{21} \end{bmatrix}$$

$$det \begin{bmatrix} H - \lambda I \end{bmatrix} = det \begin{bmatrix} h_{11} - \lambda & h_{12} \\ h_{21} & h_{22} - \lambda \end{bmatrix} = 0$$

$$(h_{11} - \lambda) (h_{21} - \lambda) - h_{21}h_{22} = 0$$

$$\lambda^{2} - (h_{11} + h_{21})\lambda + h_{11}h_{22} - h_{21}h_{22} = 0$$

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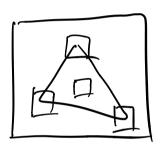
$$\lambda^{2} = \frac{1}{2} (h_{11} + h_{21} + h_{21} + h_{21}h_{21} = 0$$

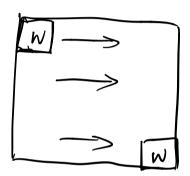
$$\lambda^{2} = \frac{1}{2} (h_{11} + h_{21} + h_{21} + h_{21}h_{21} = 0$$

$$\lambda^{2} = \frac{1}{2} (h_{11} + h_{21} + h_{21} + h_{21}h_{21} = 0$$

But large/small

1. Local feature





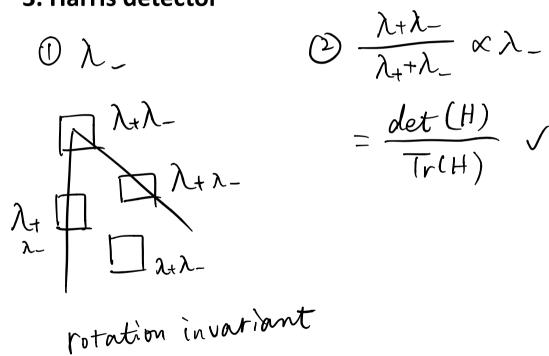
2. Kernel operator

• Ix, Iy don't change when moving the window. => Precalculated by using kernel In HW, pad zeros to handle boarders.

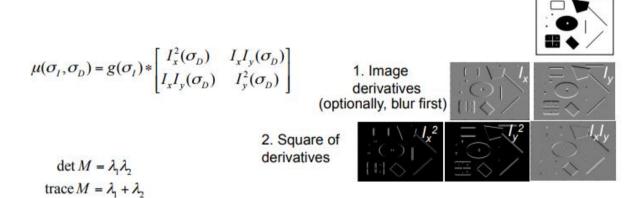
con "same"

Also possible to do summation with kernel.

3. Harris detector



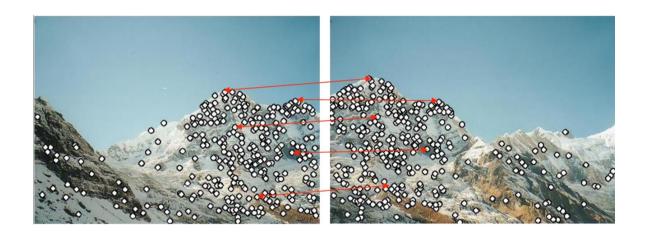
Harris Example



Cornerness function – both eigenvalues are strong
 Compute f



4. Non-maxima suppression



- 1. Find feature points in both images. -> structure tensor V
- 2. Match the feature points into pairs. -> matching algorithm, e.g., $\sqrt{}$ Hungarian
- 3. Use pairs to align two images. -> homography

Recall line fitting

y = Ax

\[
\begin{align*}
\(y = A \times \)
\(\frac{y_1}{y_n} = \begin{bmatrix} x_1 & \\ \frac{y_1}{y_n} & \\ \f