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Mosaicing of Images with Few Textures and Strong Illumination changes: Application to Gastroscopic Scenes

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Scientific challenges

- Registration of images acquired under complex conditions combining non-textured scenes with significant illumination changes between images and specular reflections.
- Mosaicing of gastroscopic image sequences: nonlinear relationship between pixel positions in the 2D mosaic due to non-planar and non-static surfaces (homographies are unusable).

Proposed variational OF method







Gastroscopic images affected by specular reflections (SR)

$$E_{data}(I_s, I_t, \boldsymbol{u}) = \sum_{x \in \Omega} \theta_x \left\| \boldsymbol{D}(P_{I_s}(x)) - \boldsymbol{D}(P_{I_t}(x + \boldsymbol{u}_x)) \right\|_2^2.$$

- Local/patch region illumination change model [1] $P_{I_t}(x + u_x) = a_x P_{I_s}(x) + b_x$ where $a_x, b_x \in \mathbb{R}$, and $a_x > 0$.
- Descriptor **D** is called an illumination invariant when $D(P_I(x)) = D(a_x P_I(x) + b_x) \qquad (6)$ $\forall a_x > 0, b_x \in \mathbb{R}.$

General form of illumination-invariant patch-based descriptors

Contributions

- Efficient descriptor-based variational optical flow (OF) method (appropriate for scenes with few textures, strong illumination changes between images and specular reflections).
- 2. Proposal of a general form of illuminationinvariant patch-based descriptors: the illumination invariance is proven both theoretically and experimentally.
- 3. Accurate mosaicing method: OF fields are directly used to map the pixels from input images to the 2D mosaic coordinate system.

References

[1] D. H. Trinh, W. Blondel, and C. Daul, "A general form of illumination-invariant descriptors in variational optical flow estimation," *in IEEE Int. Conf. on Image Processing (ICIP)*, Beijing, China, Sept. 2017.

[2] S. Ali, C. Daul, E. Galbrun, and W. Blondel, "Illumination invariant optical flow using neighborhood descriptors," *Computer Vision and Image Understanding*, vol. 145, pp. 95–110, 2016.

and results of SR regions detection using [5]. The size and shapes of SR regions vary strongly between images.

SR detection [5]:

• $\tilde{I}_x(R, G, B) = \frac{\min(R_x, G_x, B_x)}{\max(R_x, G_x, B_x)} I_x(R, G, B)$ • $\tilde{I}(R, G, B) \xrightarrow{\text{convert } RGB \text{ to } XYZ} I(X, Y, Z) \text{ and } \mathcal{Y} = \frac{Y}{X+Y+Z}$ If $Y_x > \mathcal{Y}_x$ then x is considered as a SR pixel. *Denote by \mathcal{S}_I the set of detected SR pixels in an image I.

 $S = dilation(S_{I_s}, se) \cup dilation(S_{I_t}e)$ with *se* a structural element having a size of 7x7 pixels. The pixels which belong to *S* will not be involved in the OF computation process.



Descriptor-based variational OF model:

$$\begin{split} \min_{\boldsymbol{u}} \left[E_{reg}(\boldsymbol{u}) + \lambda E_{data}(I_{s}, I_{t}, \boldsymbol{u}) \right] \\ \text{where} \\ E_{reg}(\boldsymbol{u}) &= \sum_{x \in \Omega} \sum_{x' \in \mathcal{N}_{x}} \theta_{x} \theta_{x'} w_{x'}^{x'} \| \boldsymbol{u}_{x} - \boldsymbol{u}_{x'} \|_{1}, \\ \text{with} \\ w_{x}^{x'} &= \exp\left(\frac{-\|\mathbf{x} - \mathbf{x}'\|_{2}^{2}}{2\sigma_{1}^{2}} + \frac{-\|\mathbf{L}(\mathbf{x}) - \mathbf{L}(\mathbf{x}')\|_{2}^{2}}{2\sigma_{2}^{2}} \right), \\ \theta_{x} &= \begin{cases} 0 \ if \ x \in \mathcal{S} \\ 1 \ if \ x \notin \mathcal{S} \end{cases}, \\ \text{and} \end{cases}$$

$$\mathbf{D}(P_{I}(x_{0})) = \begin{bmatrix} \operatorname{sgn}(M_{1} \otimes P_{I}(x_{0})) \\ \operatorname{sgn}(M_{2} \otimes P_{I}(x_{0})) \\ \vdots \\ \operatorname{sgn}(M_{n} \otimes P_{I}(x_{0})) \end{bmatrix} M_{i} = \begin{bmatrix} \alpha_{i,4} & \alpha_{i,3} & \alpha_{i,2} \\ \alpha_{i,5} & \alpha_{i,0} & \alpha_{i,1} \\ \alpha_{i,6} & \alpha_{i,7} & \alpha_{i,8} \end{bmatrix}$$

$$\alpha_{i,0} + \alpha_{i,1} + \dots + \alpha_{i,8} = 0$$

$$M_{i} \otimes P_{I}(x_{0}) = \sum_{j=0}^{8} \alpha_{i,j} I(x_{j}) \text{ and } \operatorname{sgn}(v) = \begin{cases} 1 & if \quad v > 0 \\ 0 & otherwise \end{cases}$$

$$M_{i} \otimes (\alpha_{x_{0}} P_{I}(x_{0}) + b_{x_{0}}) = \alpha_{x_{0}} M_{i} \otimes P_{I}(x_{0}) + b_{x_{0}} \sum_{j=0}^{8} \alpha_{i,j} \\ = \alpha_{x_{0}} M_{i} \otimes P_{I}(x_{0}), \forall i. \end{cases}$$

$$(7)$$

$$\Rightarrow \operatorname{sgn}\left(M_{i} \otimes (\alpha_{x_{0}} P_{I}(x) + b_{x_{0}})\right) = \operatorname{sgn}(M_{i} \otimes P_{I}(x_{0})), \forall i. \end{cases}$$
This descriptor satisfies illumination-invariant condition (6).



Input: Image sequence: $\{I_1, I_2, \dots, I_N\} = \{I_{k_0}, I_{k_0-1}, \dots, I_1\} \cup \{I_{k_0}, I_{k_0+1}, \dots, I_N\}, k_0 = median(1,2,3, \dots, N); I_{mosaic} = I_{k_0}.$ (1) $\{I_{k_0}, I_{k_0+1}, \dots, I_N\}, k_0 = median(1,2,3, \dots, N); I_{mosaic} = I_{k_0}.$ Step 1: Mapping of the pixels of I_n to the coordinate system of I_{k_0} - Compute OF from I_{k_0} to I_n :
(2) $F_{k_0,n} = \sum_{i=k_0}^{n-1} OF_{i,i+1} \text{ if } n > k_0$ $F_{k_0,n} = \sum_{i=k_0}^{n+1} OF_{i,i-1} \text{ if } n < k_0.$ (8)
- Perform warping I_n into the common coordinate system using $F_{k_0,n}$ and linear interpolation to get the warped image I_n^{warped} .
Step 2: Blending: For n = 1 to N
(4) $I_{mosaic} \leftarrow \text{Blending}(I_{mosaic}, I_n^{warped});$ (9)

[3] M. Drulea and S. Nedevschi, "Motion estimation using the correlation transform," *IEEE Trans. on Image Processing*, vol. 22, no. 8, pp. 3260–3270, 2013.

[4] Yinlin Hu, Rui Song, and Yunsong Li, "Efficient coarse-to-fine patch match for large displacement optical flow," in *CVPR*, Boston, MA, USA. 2016, IEEE.

[5] O. E. Meslouhi, M. Kardouchi, H. Allali, T. Gadi, and Y. A. Benkaddour, "Automatic detection and inpainting of specular reflections for colposcopic images," *Central European Journal of Computer Science*, vol. 1, no. 3, pp. 341–354, 2011.

Experimental Results









Mosaic built with 22 images extracted from a gastroscopic video-sequence in the pyloric antrum region. Color discontinuities are due to viewpoint changes of the endoscope.



OF results of the CPM-Flow method [4] and the proposed OF method based on the RKII-descriptor.



Mosaicing result for a gastroscopic sequence of 58 images. Color discontinuities were not corrected to show the contribution of different images to the mosaic.

