Medical Image Processing for Interventional Applications Feature Detectors

Online Course – Unit 10 Andreas Maier, Sebastian Bauer, Frank Schebesch Pattern Recognition Lab (CS 5)













Topics

Feature Detectors

Initial Considerations

Harris Corner Detector

Summary

Take Home Messages

Further Readings

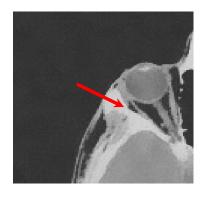






Feature Detectors

How to identify distinctive locations efficiently?



intensity profile

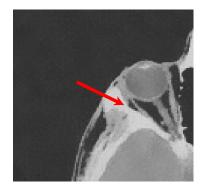




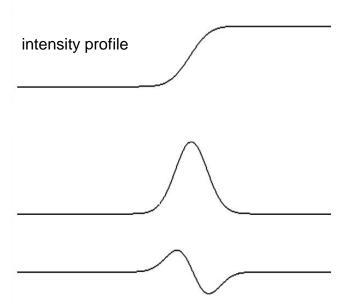


Feature Detectors

How to identify distinctive locations efficiently?



- First order derivatives
- Second order derivatives
- Structure tensor, Hessian matrix









Feature Detectors: Benchmark Study by Schmid, Mohr, and Bauckhage (2000)

Which one to choose?

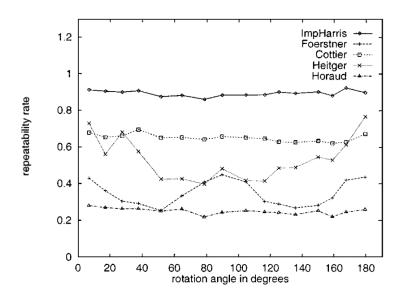


Figure 2: Rotation invariance

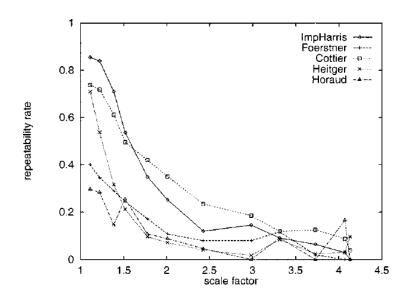


Figure 3: **Scale** invariance







Feature Detectors: Benchmark Study by Schmid, Mohr, and Bauckhage (2000)

Which one to choose?

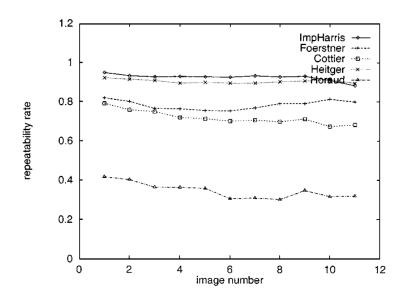


Figure 4: Illumination invariance

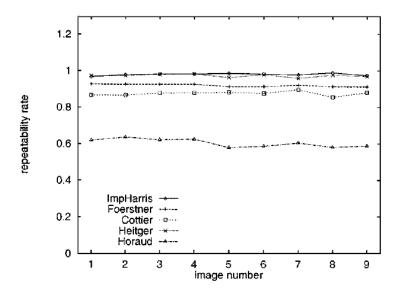


Figure 5: **Noise** invariance







Basic idea (Harris and Stephens, 1988): A corner point should have large intensity changes in all directions.

Gradient approximation:

$$\mathbf{g}(x,y) = \nabla f(x,y) = \begin{pmatrix} f_x(x,y) \\ f_y(x,y) \end{pmatrix}$$

Structure tensor (autocorrelation):

$$\mathbf{G}(x,y) = \sum_{i=-k}^{k} \sum_{j=-k}^{k} w(x,y) \mathbf{g}(x+i,y+j) \mathbf{g}^{T}(x+i,y+j)$$

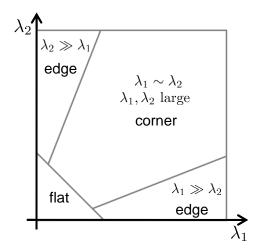
$$= \sum_{i=-k}^{k} \sum_{j=-k}^{k} w(x,y) \begin{bmatrix} (g_{x}(...))^{2} & g_{x}(...)g_{y}(...) \\ g_{x}(...)g_{y}(...) & (g_{y}(...))^{2} \end{bmatrix}$$







Eigenvectors and eigenvalues of structure tensor **G** describe predominant directions of the gradient:



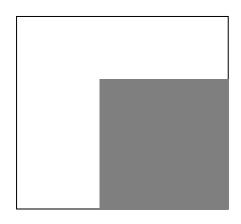


Figure 6: Schematic of relationship between feature categories and eigenvalues (left), example image with a corner, edges and flat areas (right)

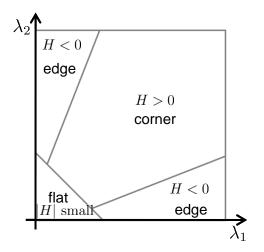






Eigenvectors and eigenvalues of structure tensor **G** describe predominant directions of the gradient:

$$H(x,y) = \det(\mathbf{G}(x,y)) - \nu \left(\operatorname{tr}(\mathbf{G}(x,y)) \right)^{2}$$



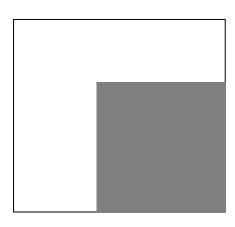


Figure 7: Structure described by value of *H* (left), example image with a corner, edges and flat areas (right)

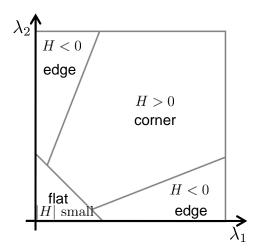






Eigenvectors and eigenvalues of structure tensor **G** describe predominant directions of the gradient:

$$H(x,y) = \det(\mathbf{G}(x,y)) - \nu \left(\operatorname{tr}(\mathbf{G}(x,y)) \right)^2 = \lambda_1 \lambda_2 - \nu (\lambda_1 + \lambda_2)^2$$



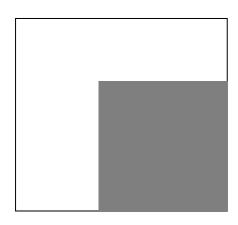


Figure 7: Structure described by value of *H* (left), example image with a corner, edges and flat areas (right)







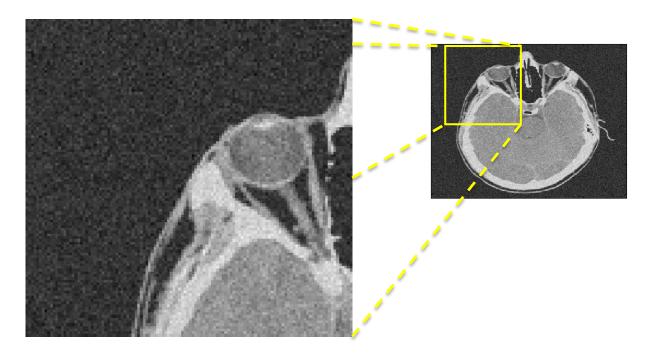


Figure 8: What about noise?







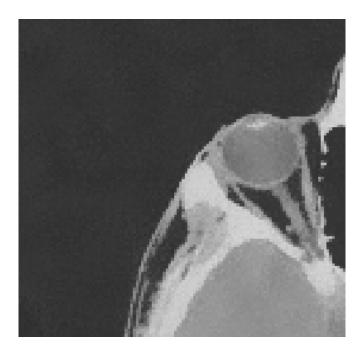


Figure 9: Edge-preserving denoising







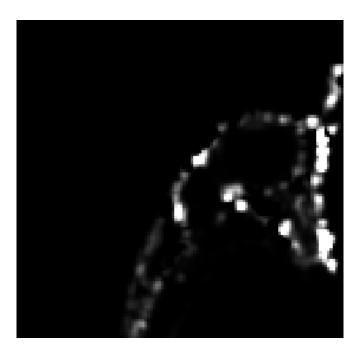


Figure 10: Corner response







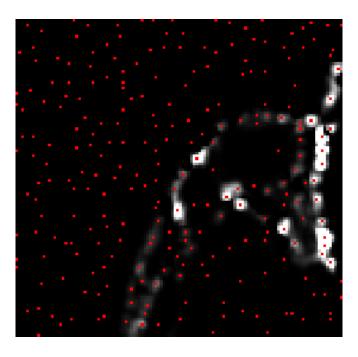


Figure 11: Corner localization, non-maximum suppression







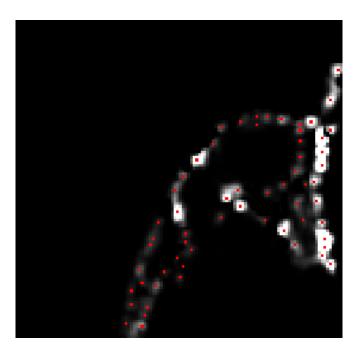


Figure 12: Corner selection







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Take Home Messages

- An analysis of the derivatives in an image yields important information to build a feature detector.
- Best choose features which are invariant to certain transformations.
- The Harris corner detector utilizes the structure tensor to determine image points to be corners, edges or rather part of flat and homogeneous areas.

Credits:

We acknowledge the contributions of F.F. Li, E. Angelopoulou, D. Lowe, and A. Berg for their material in units 9-14 (on feature detectors/descriptors).







Further Readings

- Cordelia Schmid, Roger Mohr, and Christian Bauckhage. "Evaluation of Interest Point Detectors". In: *International Journal of Computer Vision* 37.2 (June 2000), pp. 151–172. DOI: 10.1023/A:1008199403446
- Chris Harris and Mike Stephens. "A Combined Corner and Edge Detector". In: Proceedings of Fourth Alvey Vision Conference. 1988, pp. 147–152