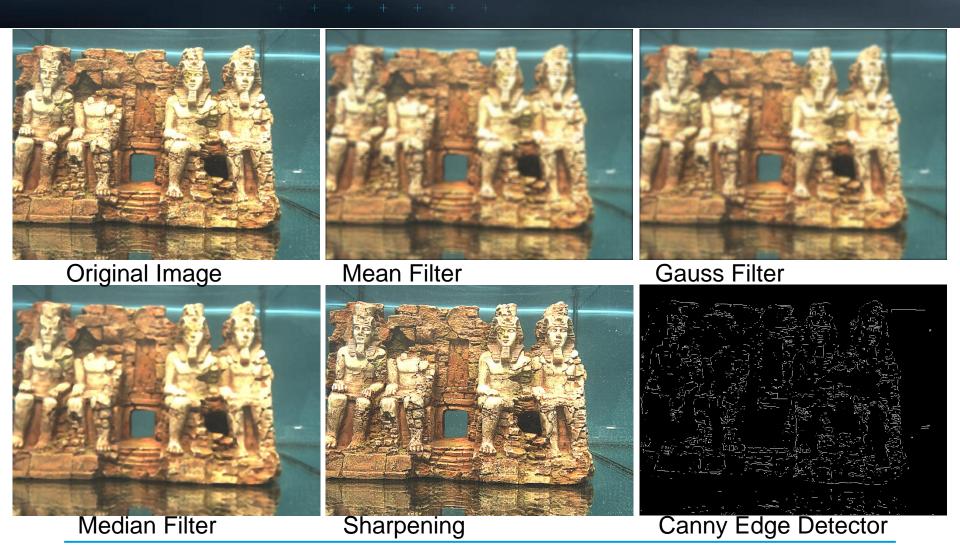
Outline



- Introduction
- Features and Feature Matching
- Geometry of Image Formation
- Calibration
- Structure from Motion
- Dense Stereo
- Conclusion

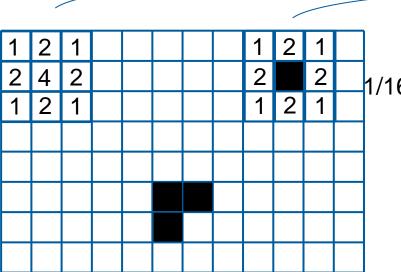








Gaussian Filter – Smoothing the Image



| | 1 | 2 | 1 |
|---|---|---|---|
| 6 | 2 | 4 | 2 |
| | 1 | 2 | 1 |

| | | - | |
|--|--|---|--|
| | | | |







Gradient Filter

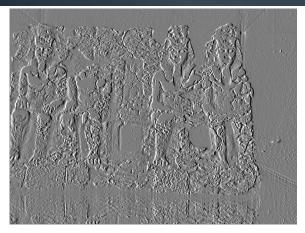


| -1 | 0 | 1 |
|----|---|---|
| -2 | 0 | 2 |
| -1 | 0 | 1 |

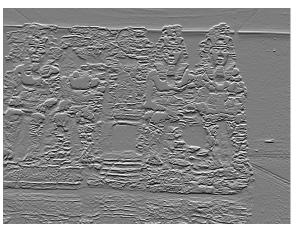
Sobel filter in x-direction

| -1 | -2 | -1 |
|----|----|----|
| 0 | 0 | 0 |
| 1 | 2 | 1 |

Sobel filter in y-direction



gradient x-direction

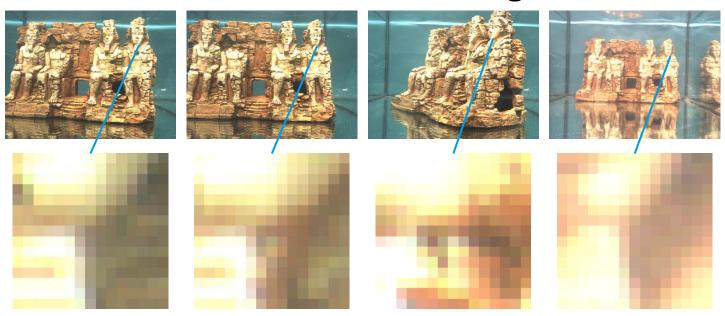


gradient y-direction





Feature Detection and Matching



Input: dense image sequences (video) or unordered, images (e.g. from the internet) Question: is there overlap between the images?

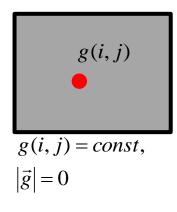


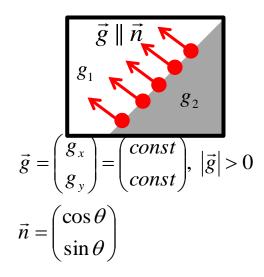


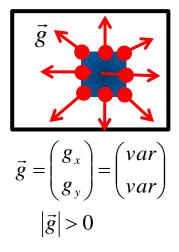
Feature Detection – Harris Corners

Images are composed of 3 basic primitives:

- Planar Patches: regions with constant gray values (zero gradient)
- Edges: regions with constant gradient vector (normal)
- Points (corners): regions with variable gradient











Feature Detection – Harris Corners

Corners, edges, and planes can locally be distinguished by computing the structure tensor J. J(i, j) computes information on the local structure at position (i, j) for a local image region S from the gradients:

- 1. Compute gradient images: $I_x(i,j) = H_{gx} * I(i,j), I_y(i,j) = H_{gy} * I(i,j)$
- 2. Point-wise products for the components of the structure tensor

$$I_{xx}(i, j) = I_{x}(i, j) \cdot I_{x}(i, j), \quad I_{yy}(i, j) = I_{y}(i, j) \cdot I_{y}(i, j), \quad I_{xy}(i, j) = I_{x}(i, j) \cdot I_{y}(i, j)$$

3. Local smoothing of the gradients

$$J_{xx} = G_{LP} * I_{xx}(i, j), \quad J_{yy} = G_{LP} * I_{yy}(i, j), \quad J_{xy} = G_{LP} * I_{xy}(i, j)$$

4. Computation of trace and determinant

$$J(i,j) = \begin{vmatrix} J_{xx} & J_{xy} \\ J_{xy} & J_{yy} \end{vmatrix}, \quad trace(J) = (J_{xx} + J_{yy}), \quad det(J) = J_{xx} \cdot J_{yy} - J_{xy}^2$$

HELMHOLTZ ASSOCIATION



Feature Detection – Harris Corners

Harris Corner Detector evaluates J:

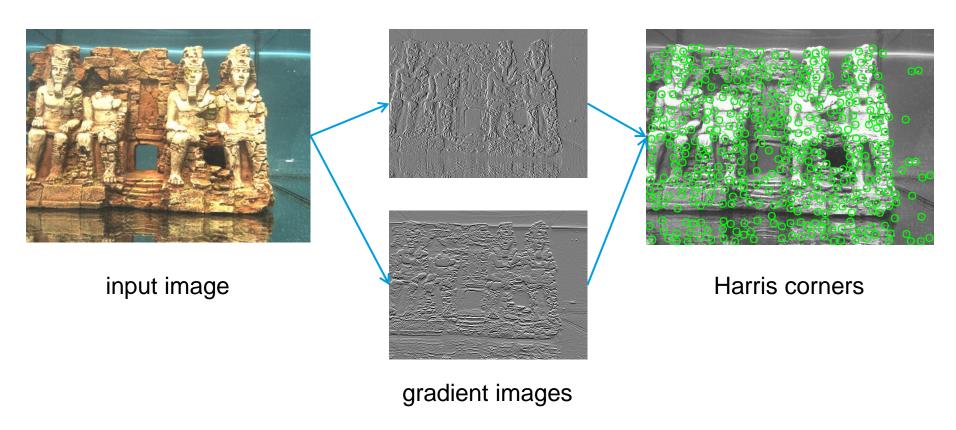
$$\begin{split} &\lambda_0, \lambda_1 Eigenwerte \ J \Rightarrow det(J) = \lambda_0 \lambda_1, \ trace(J) = \lambda_0 + \lambda_1 \\ &C_H = det(J) - k \cdot \left(trace(J)\right)^2, \quad 0 \leq k \leq = 0.25, \ k = 0.04 \\ &C_H \begin{cases} > t \ for \ edges \\ < t \ for \ corners \ and \ planes \end{cases} \end{split}$$

- C_H measures the 'cornerness'
- CH>t, t>0 potential corners
- select local maxima in a neighborhood (Non-Maxima-Suppression).



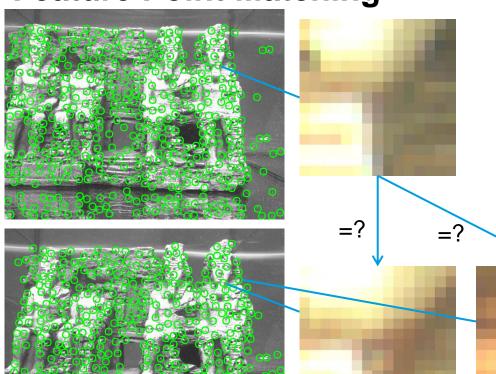


Gradients – Harris Corner Detector





Feature Point Matching



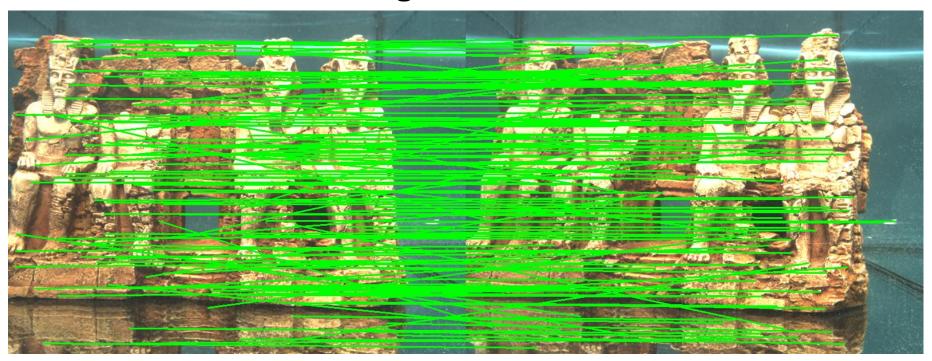
Pixel-wise patch comparison:

- SSD Sum of Squared Differences
- NCC Normalized Cross
 Correlation (invariant against brightness changes)





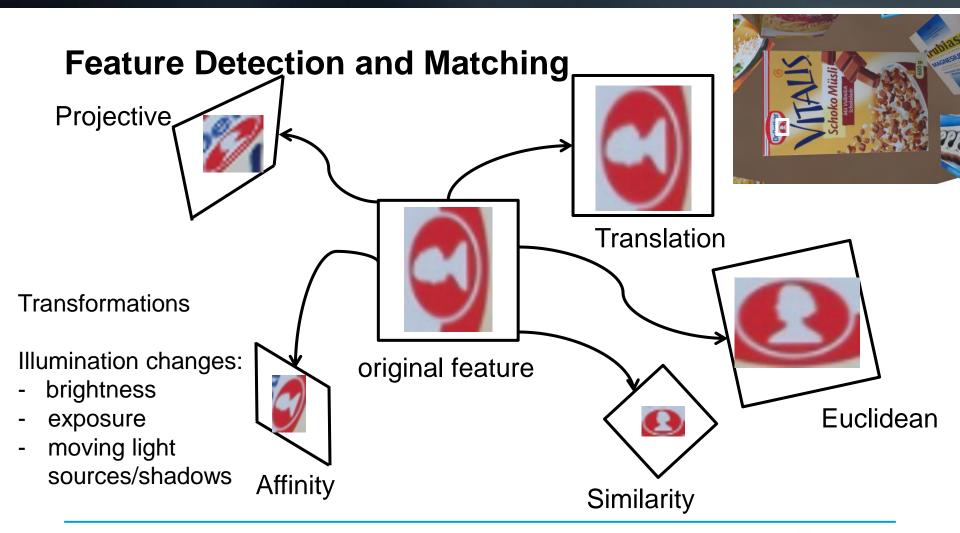
Feature Point Matching



Automatically detect hundreds or thousands of correspondences Problem: outliers











Feature Detectors & Descriptors

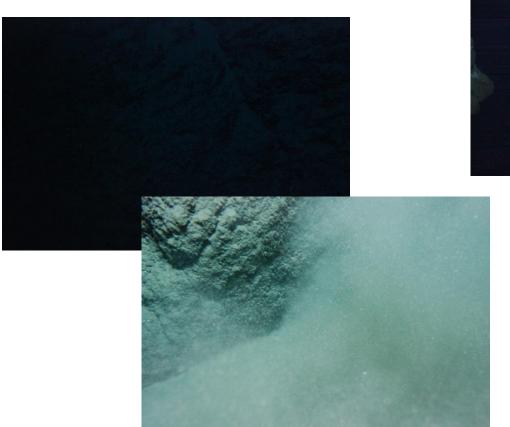
| Detector Class | Descriptor | Comparison |
|--|-------------|-----------------|
| Harris Corners | Image patch | SSD, NCC |
| Difference of Gaussian (DoG) | SIFT | Vector Distance |
| Maximally stable extremal regions (MSER) | e.g. SIFT | Vector Distance |
| many, many more | | |

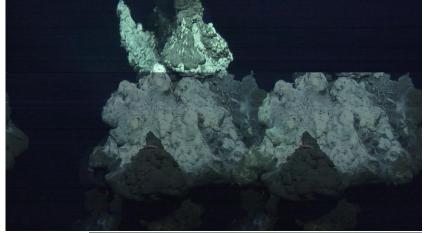
From now on: SIFT-features (out of scope in this tutorial)





Problematic Cases











References

- R. Szeliski, Computer Vision Algorithms and Applications, Springer 2011.
- C. Harris, M. Stephens. A combined corner and edge detector, Proceedings of the 4th Alvey Vision Conference, pages 147-151, 1988
- D. G. Lowe. Distinctive Image Features from Scale-Invariant Keypoints, International Journal of Computer Vision, 2004





Wrap up

- automated detection of feature points in all images
- match feature points between images
- different methods depending on baseline between images, expected rotation, translation, and scale
- illumination brightness, exposure

