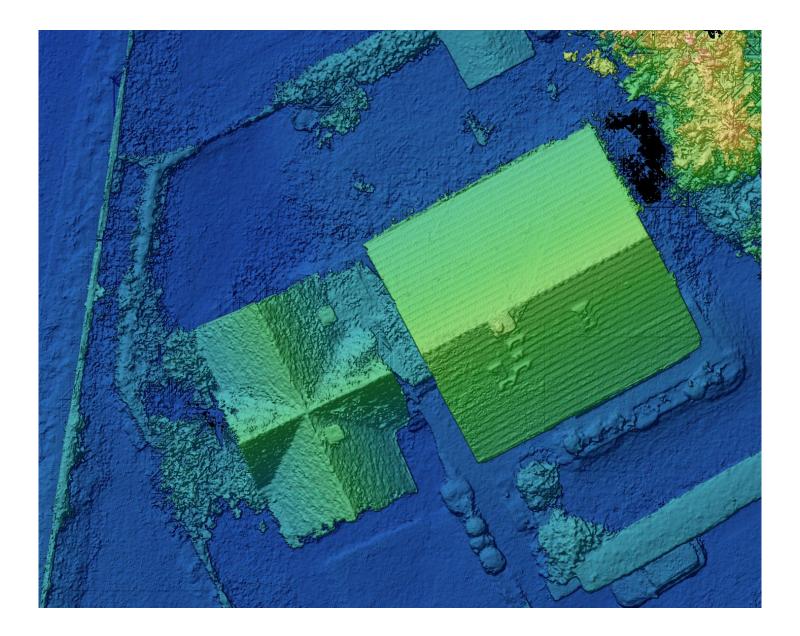
# Camera Drones Lecture – 3D data generation

Prof. Friedrich Fraundorfer

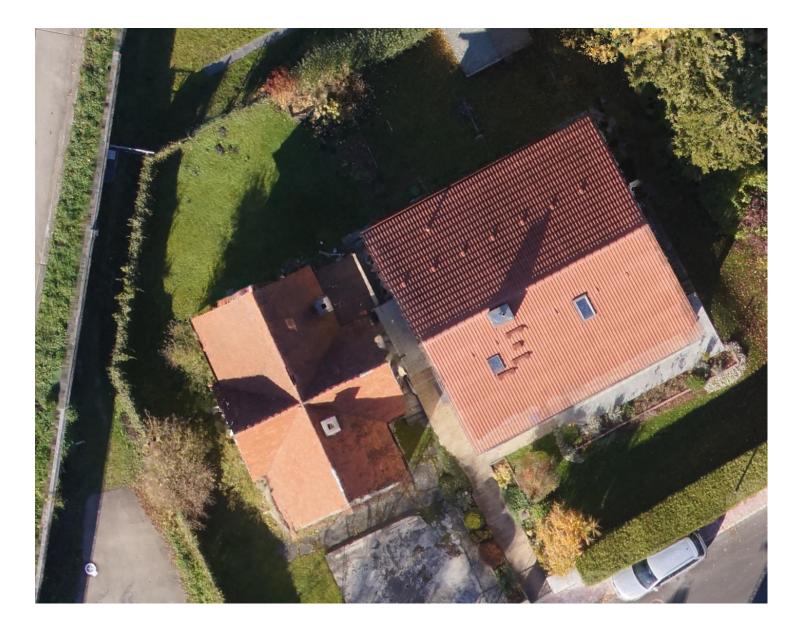
WS 2023

# Outline

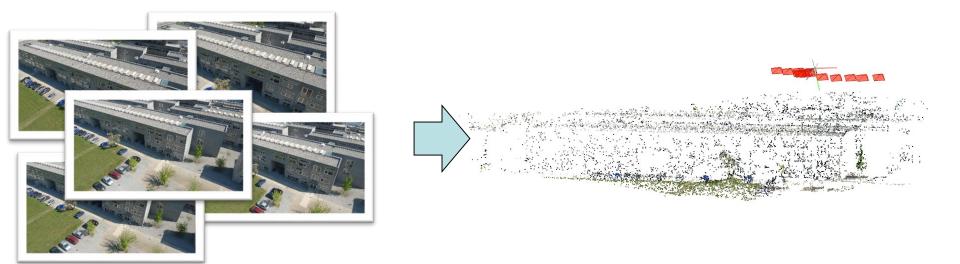
- Structure-from-Motion
  - SfM introduction
  - SfM concept
  - Feature matching
  - Camera pose estimation
  - Bundle adjustment
- Dense matching
- Data products (Orthophoto, DSM)



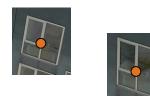
# Orthophoto



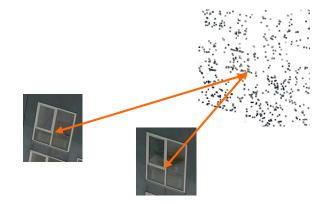
#### Structure-from-Motion (SfM) concept



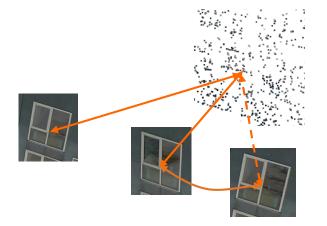
#### Structure-from-Motion (SfM) concept



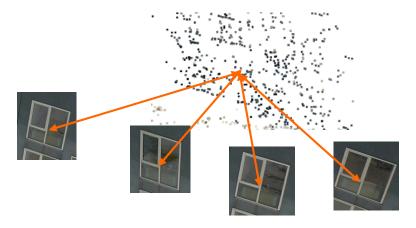
Initialize Motion  $(P_1, P_2 \text{ compatible with F})$ 



#### Initialize Structure (minimize reprojection error)

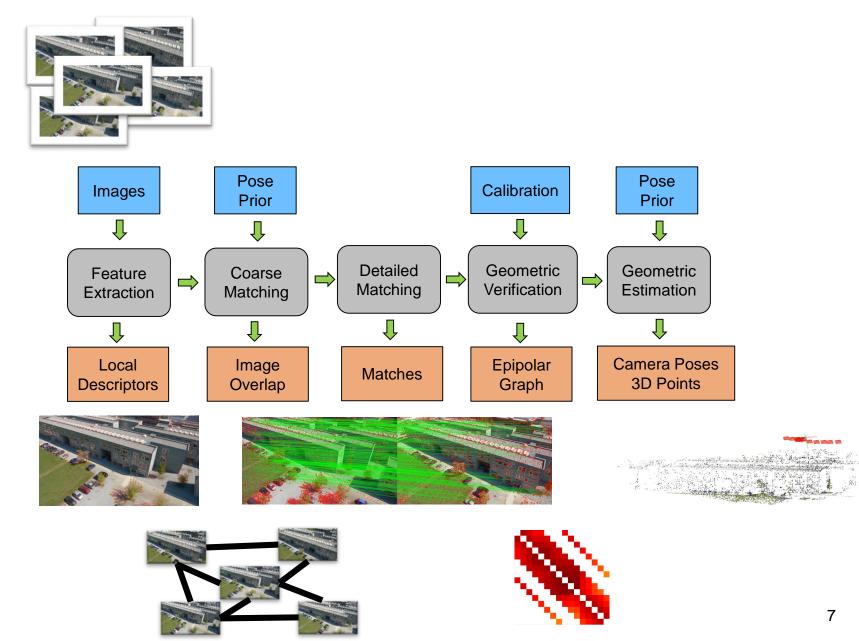


Extend motion (compute pose through matches seen in 2 or more previous views)

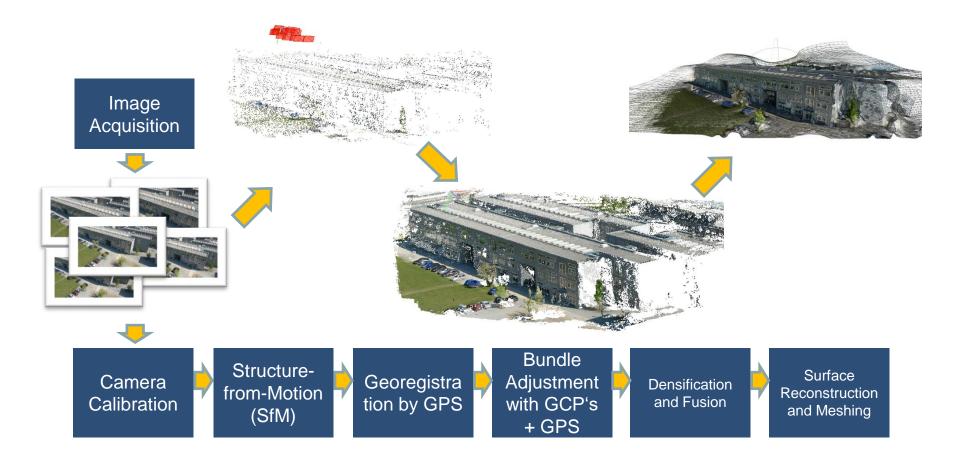


Extend structure (Initialize new structure, refine existing structure)

#### Structure-from-Motion (SfM) core pipeline

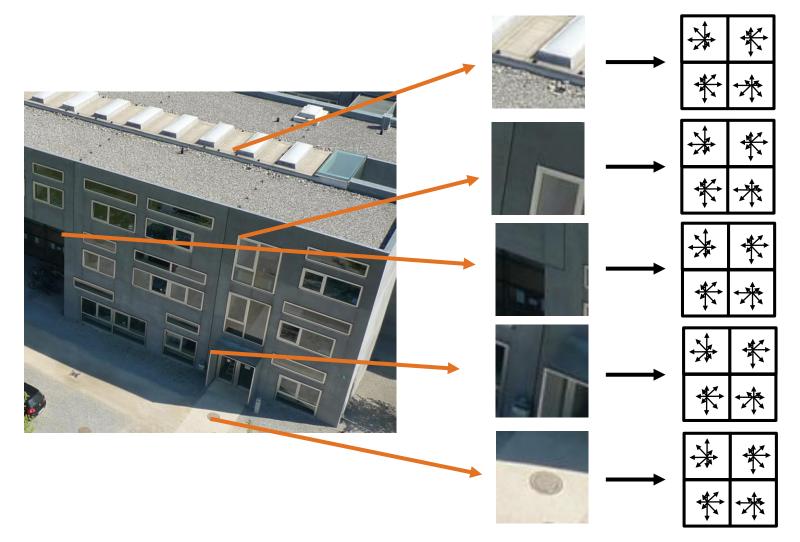


#### 3D data generation overall pipeline



#### Feature extraction

- SIFT features (best working features for matching right now)
- Each descriptor is a vector of length 128 (gradient histogram)



# **Coarse matching**

- Cluster similar images by similarity using visual words
- Will be used for speeding up exhaustive (nxn) matching

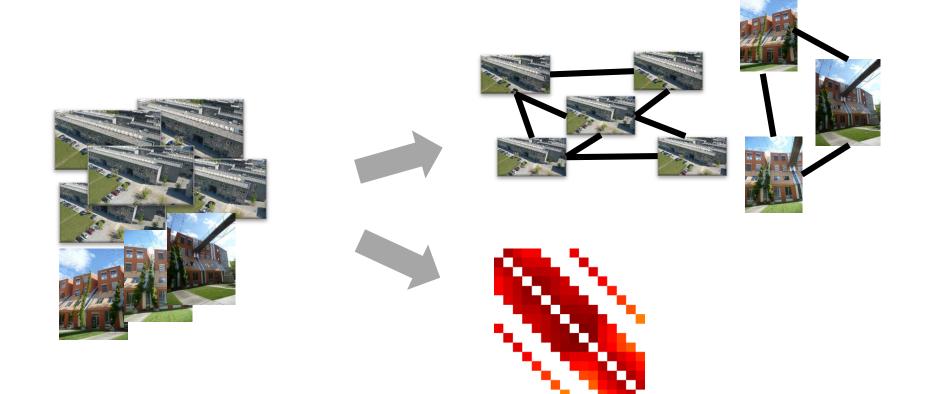
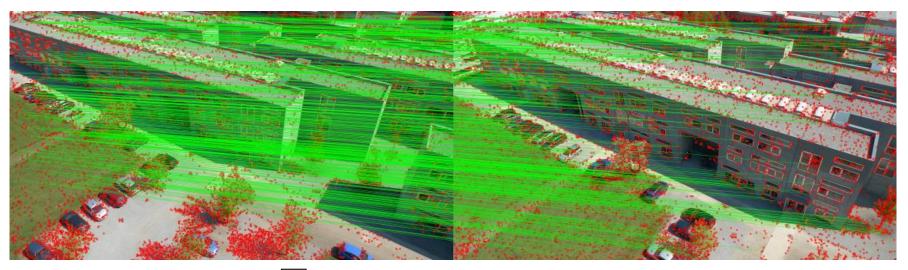
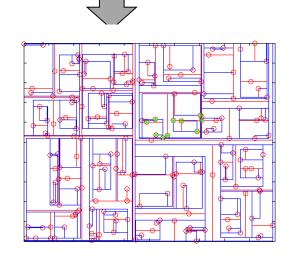


Image similarity

#### **Detailed matching**

Typically using NN-search with a Kd-tree





#### Epipolar graph

- Defines the sequential order for geometry processing
- Is a plot of the number of geometrically verified feature matches

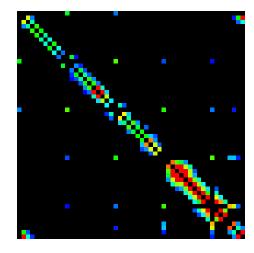
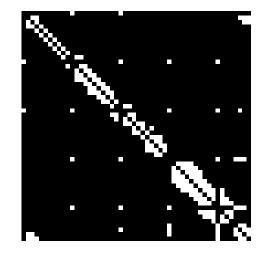


Image similarity



Epipolar graph

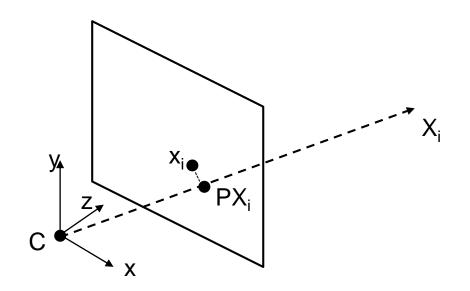
#### Geometry estimation

- Following the sequence ordering from the epipolar graph geometry is estimated for all images
- Geometry estimation is an alternating scheme:
  - Estimate camera pose of new images (position, rotation)
  - Triangulate new 3D data points seen in new image
  - Refinement by non-linear optimization (Bundle adjustment)

#### **Bundle adjustment**

- Levenberg-Marquard optimization of re-projection error
- Parameters are camera poses and all 3D points (millions of parameters to optimize!)

$$\min_{P_j, X_i} \left( \sum_i \sum_j \|x_{i,j} - P_j X_i\| \right)$$

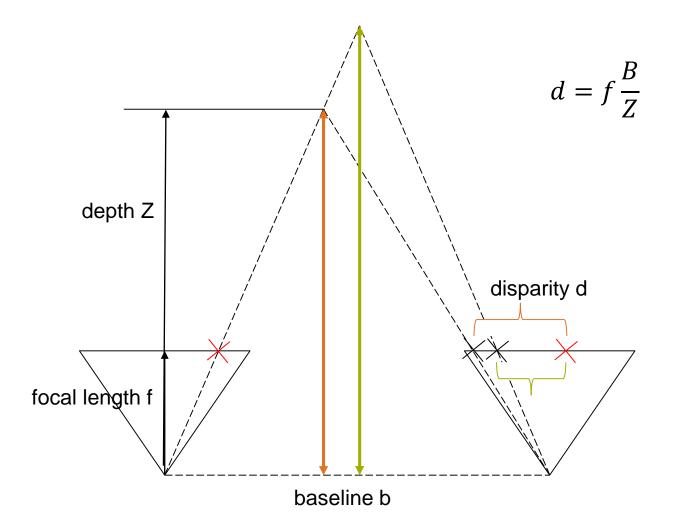


#### **Dense matching**

- SfM only gives sparse 3D data
- Only SIFT feature points are triangulated for most pixel no 3D data is computed
- Dense image matching computes a 3D point for every pixel in the image (1MP image leads to 1 million 3D points)
- Dense matching algorithms need camera poses as prerequisite

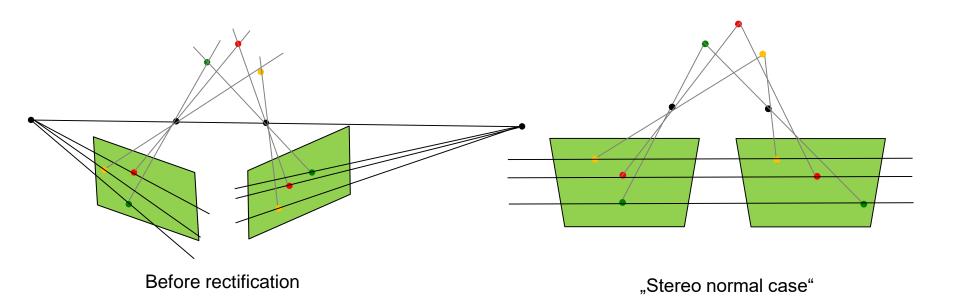
#### Geometric relation

- Stereo normal case
- Depth Z [m] can be computed from disparity d [pixel]

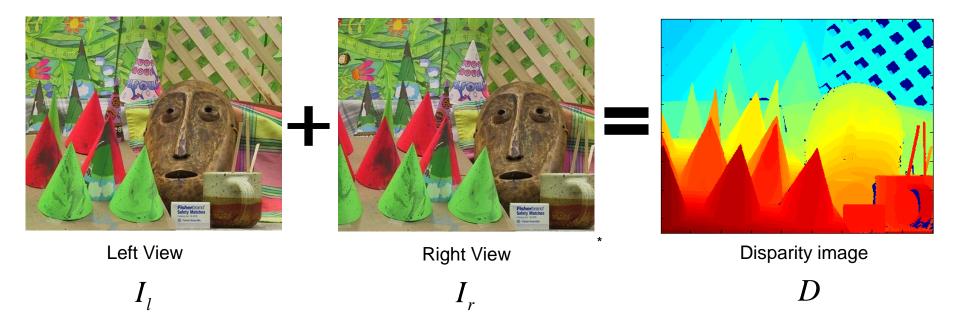


#### Rectification

- Image transformation to simplify the correspondence search
  - Makes all epipolar lines parallel
  - Image x-axis parallel to epipolar line
  - Corresponds to parallel camera configuration



#### Dense matching process



- Estimate disparity (depth) for all pixels in image left.
  - Evaluate correspondence measure for every possible pixel location on the line (e.g. NCC, SAD)
- Disparity d: Offset between pixel p in the left image and its correspondent pixel q in the right image.

#### **Census Transform**

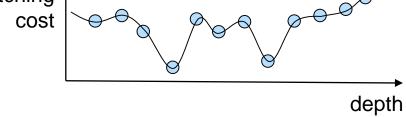
- A popular block matching cost
- Good robustness to image changes (e.g. brightness)
- Matching cost is computed by comparing bit strings using the Hamming distance (efficient)
- Bit strings encode if a pixel within a window is greater or lesser than the central pixel (0.. if center pixel is smaller, 1.. if center pixel is larger)

89	63	72	
67	55	64	00000011
58	51	49	

#### Dense matching process

matching image epipolar line Fisherbrand Salety Matches Fisherbrand Safety Matches matching **†** 000

reference image

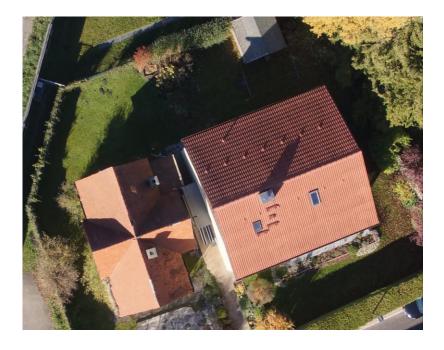


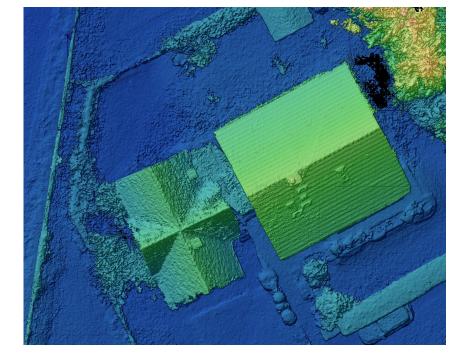
#### **Disparity selection**

- Single scanline based
  - Winner takes all (WTA)
    Select the disparity with the lowest cost (i.e. the highest similarity)
  - Scanline optimization (Dynamic programming)
    Select the disparities of the whole scanline such that the total (added up) costs for a scanline is minimal
- Global methods (Cost volume optimization)
  - Belief propagation
    Selects the disparities such that the total cost for the whole image is minimal
  - Semi-global Matching Approximates the optimization of the whole disparity image

#### Digital Surface Model (DSM)

1 height value per ground location

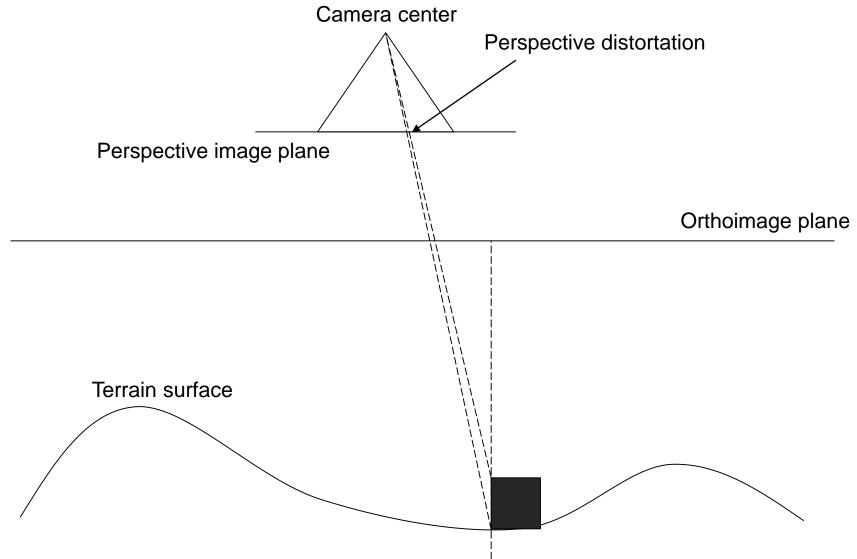




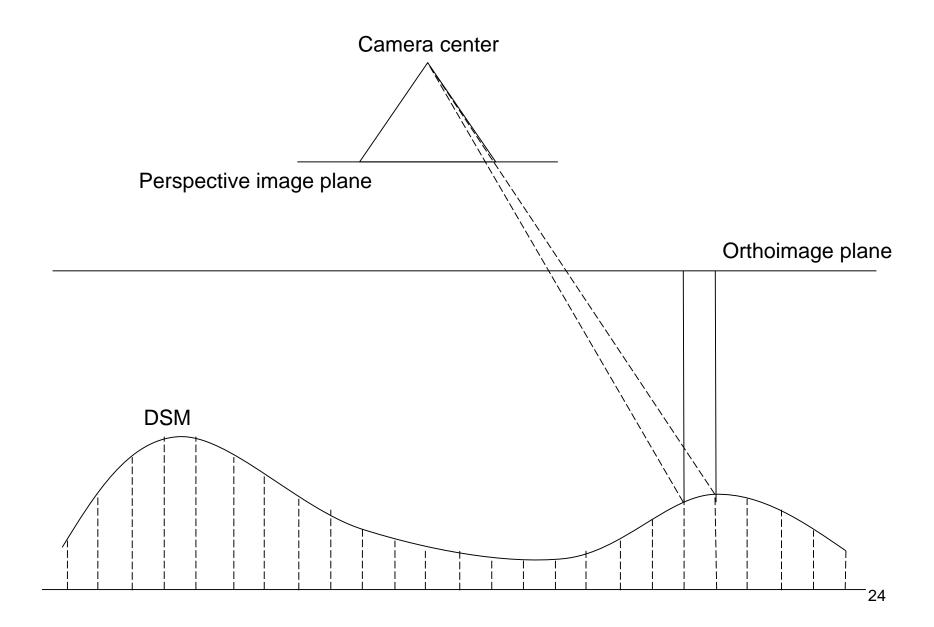
#### original image

digital surface model (color represents height)

#### Orthographic image projection

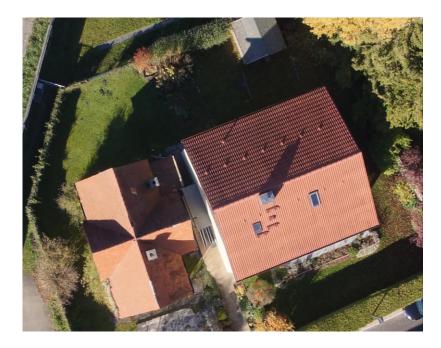


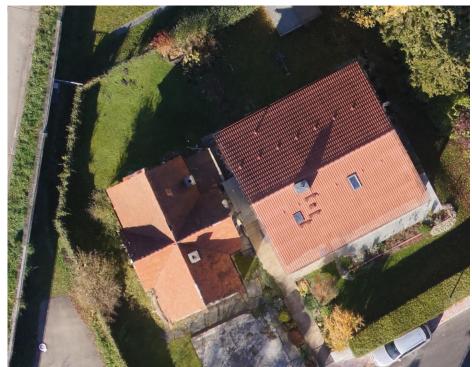
# Orthophoto generation



#### Orthophoto example

• A true orthophoto has all perspective effects removed





#### original image

#### true orthophoto