

# A Review

On Large Scale Visual Geo-Localization of Images in  
Mountainous Terrain

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Where this photo is taken?





Can we locate a photo by the skyline?



# About the Paper

- Authors
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- From
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- In
  - Computer Vision–ECCV 2012

# Introduction

- The idea is to
  - Obtain geo-localization
  - From a single image
- Useful for
  - Historical and forensic sciences
  - Documentation purposes
  - Photo organization for archival purposes
  - Intelligence applications
- It is
  - an automated approach for very large scale visual localization
  - that can efficiently exploit contours and geometric constraints



No GPS

*Strasse in Lauterbrunnen*

Wehrli (1920)

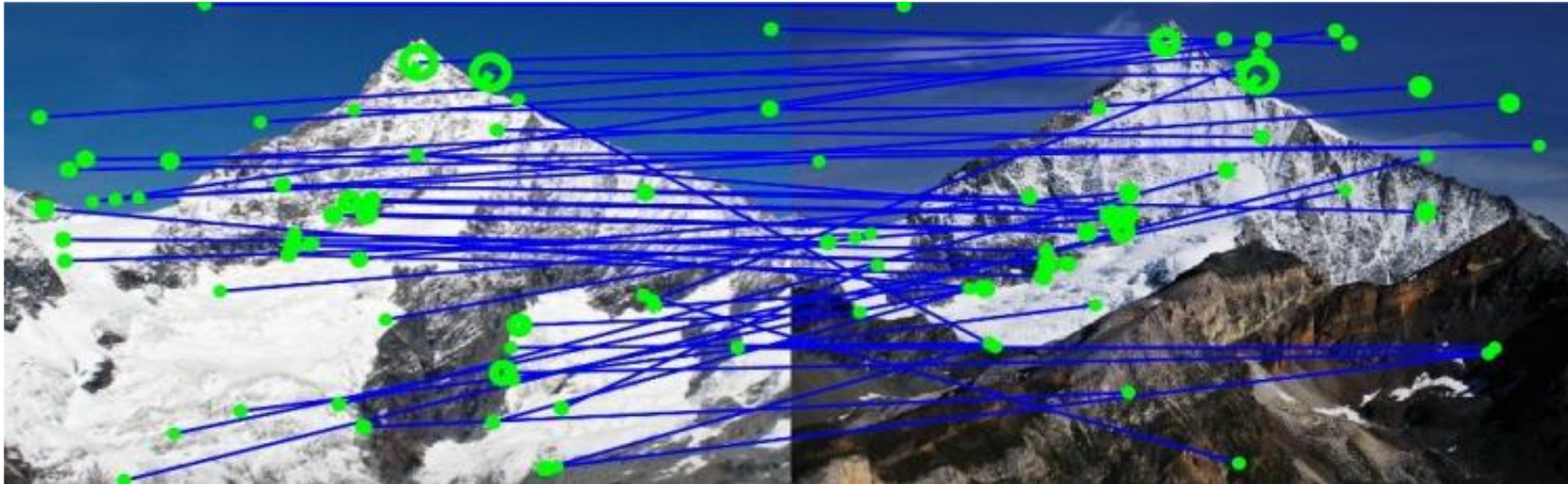


# Study

- Focuses on mountainous terrain
- Identify location of photo from shape of mountains on horizon
- 3D model of Switzerland is used
- Proposes an automated approach
- Dataset > 200 landscape query pictures with ground truths
- Validated in Switzerland, 40.000 km<sup>2</sup>
- 90% accuracy (within 1 km)
- Ultimate goal, given a DEM, localize **any** outdoor image in the **world**



# Motivation



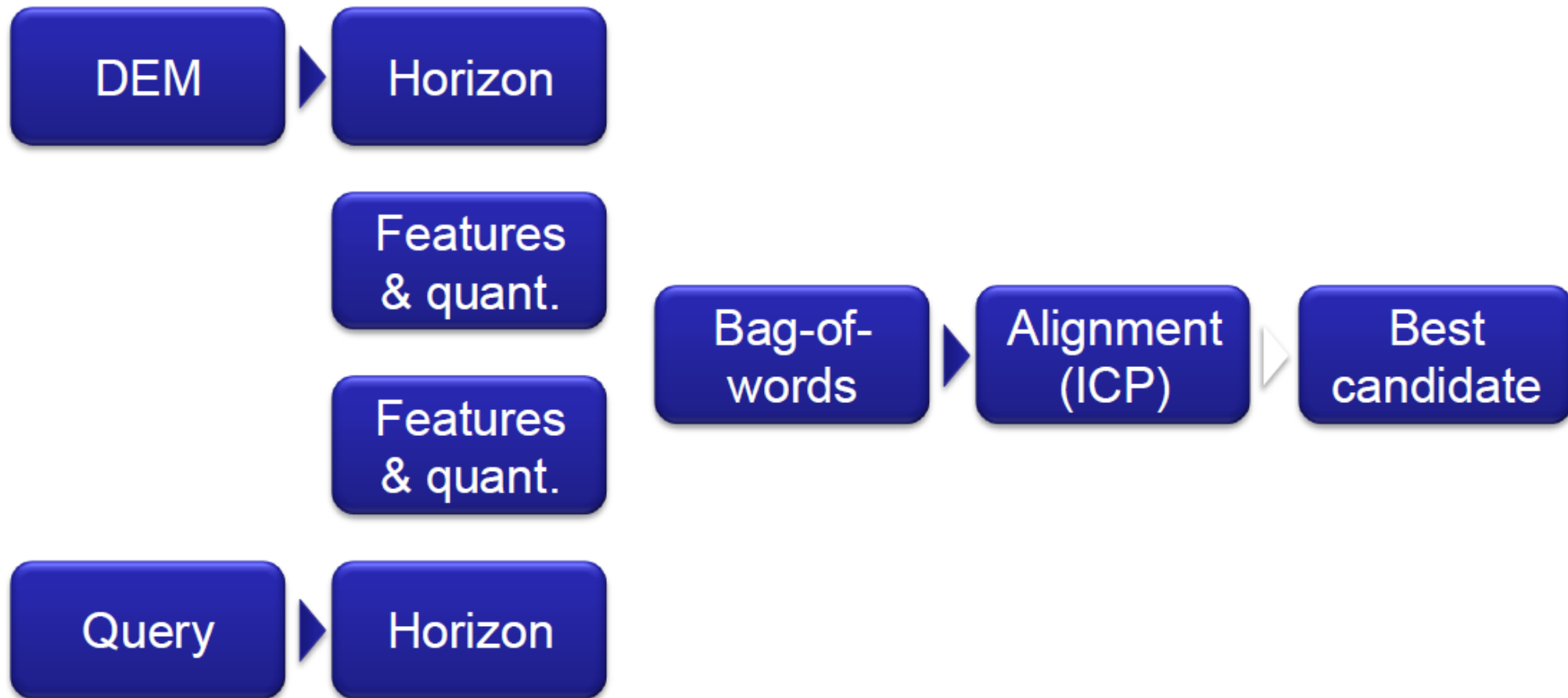
- Tremendous progress for cities
- But, natural environments are difficult
  - Vegetation, illumination, seasonal changes, snow cover, wheather conditions etc.
- Advantage, horizon remains very stable over long timespans



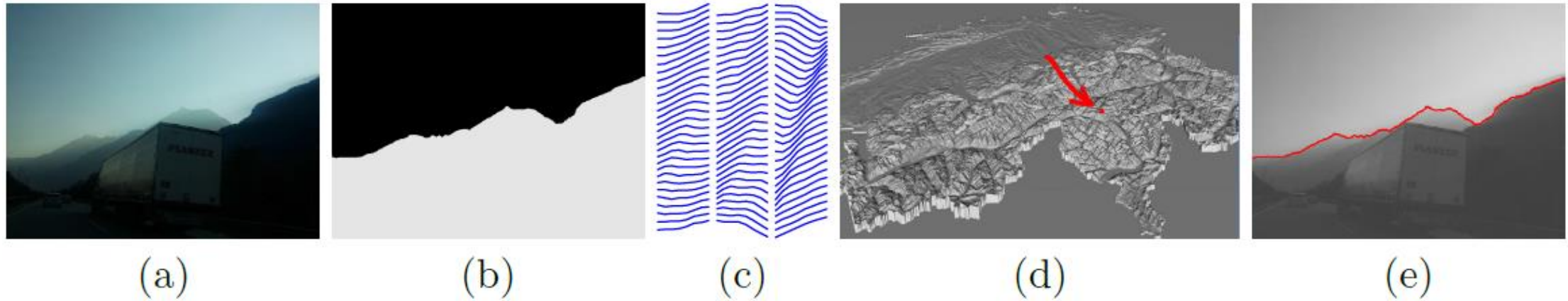
# Contribution

- A novel method for robust contour encoding
- Two different voting schemes to solve the large scale camera pose recognition from contours
  - First operates only in descriptor space
  - Second is a combined vote in descriptor and rotation space
- Works across the whole skyline, in contrast to previous work, that is matching only peaks
- Large scale, previous methods fail due to exhaustive search

# Overview



# Proposed Pipeline



- (a) Query image somewhere in Switzerland
- (b) Sky segmentation
- (c) Sample set of extracted  $10^\circ$  contourlets
- (d) Recognized geo-location in digital elevation model
- (e) Overlaid visible horizon at estimated position



# Recognition Approach

- Three position and three orientation parameters have to be estimated
- Assume that photo close to ground and people rarely twist camera
- Uses a representation that is robust w.r.t tilt of the camera
- Visible horizon of the DEM is extracted offline
- Differ from visual word based approaches by storing a viewing angle
- At query time, a sky segmentation technique is applied
- Extracted contour described by local contourlets and angular distance
- Find the most promising location and vote for the viewing direction

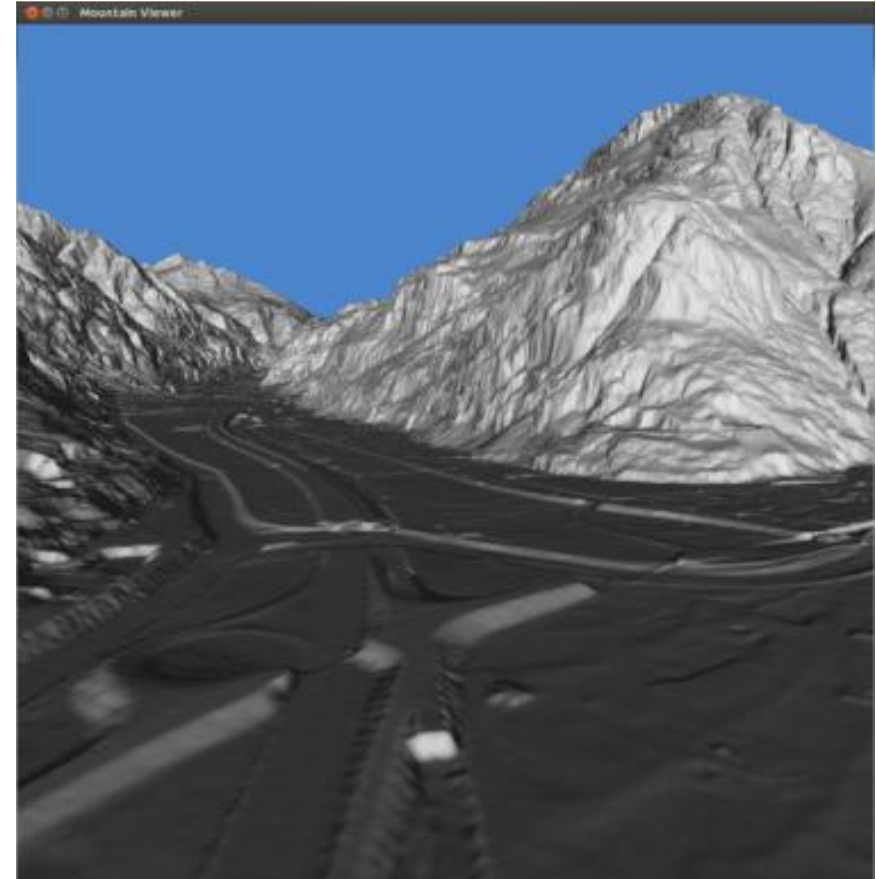
# Database Creation

- Digital elevation map (of Switzerland)
- 0.5 sample points per m<sup>2</sup>
- 0.5 - 8m vertical precision



# Database Creation (cont'd)

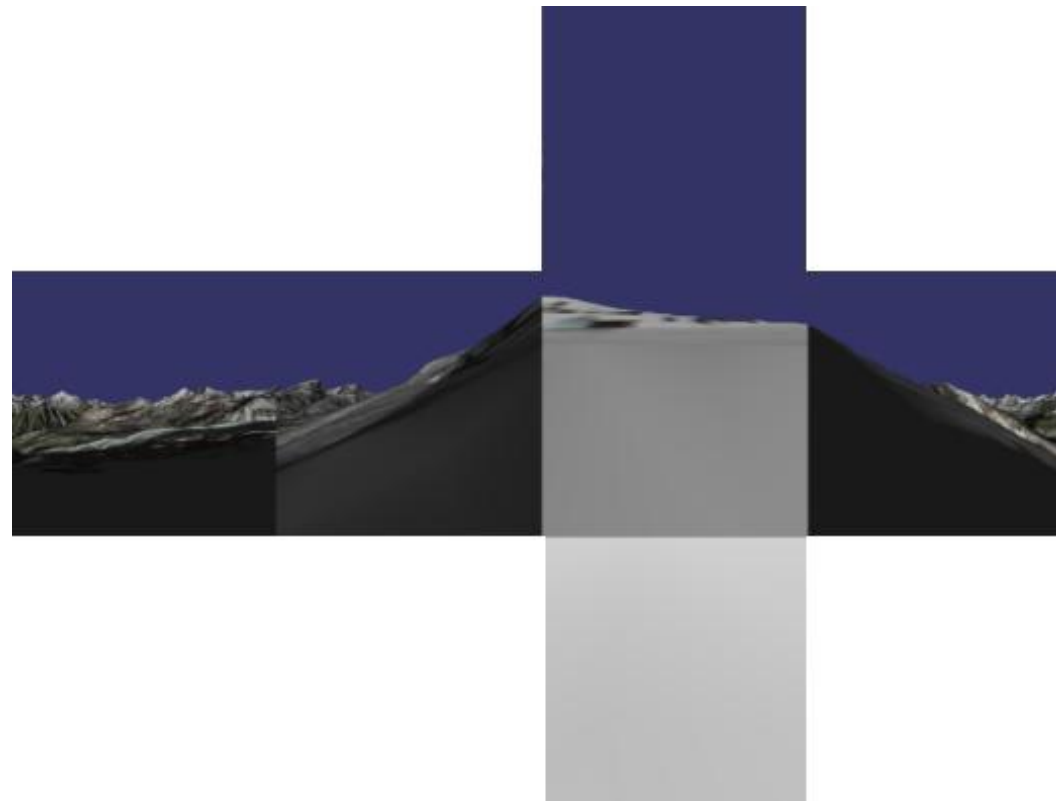
- Untextured 3D model
- Generated with Virtual Planet Builder
- Viewer based on Open Scene Graph





# Database Creation (cont'd)

- 3.5 million cubemaps generated
- Densely on  $\sim 100 \times 100 \text{m}$  grid
- Also extract contourlet
  - But with absolute viewing direction



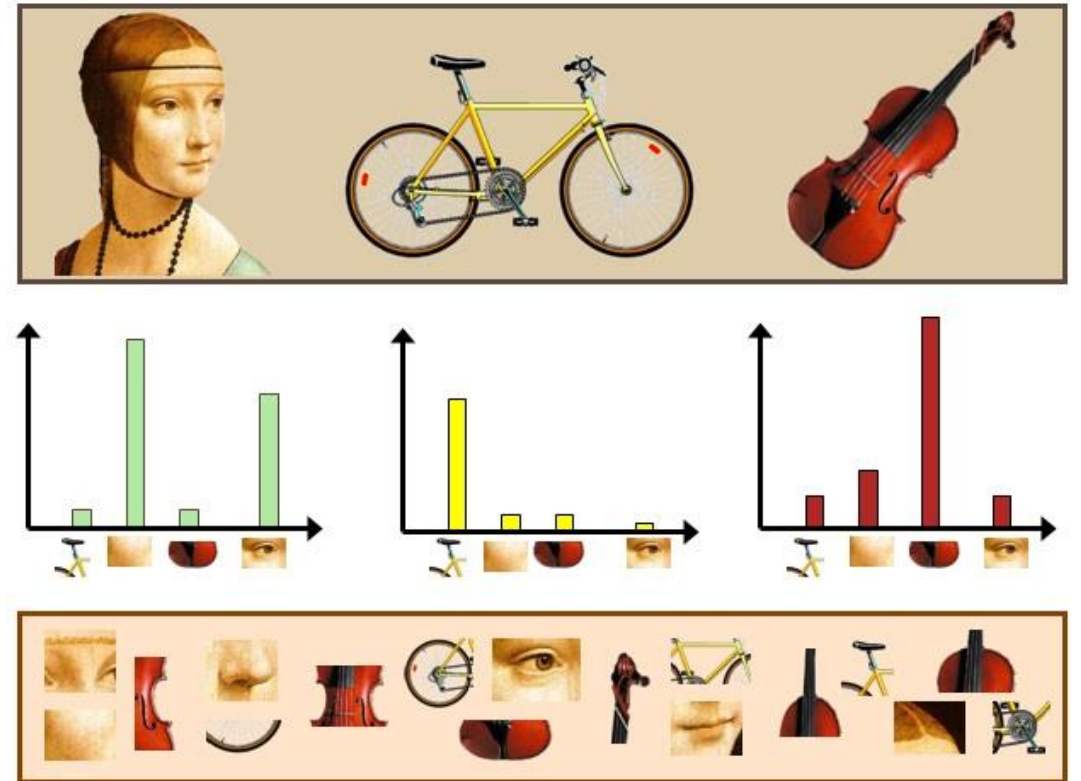
# Sky Segmentation



- De-hazing based on dark channel prior
- Structural constraints: **no sky below ground**
- A pixel being assigned to **sky** or **ground**
- Supports user-interaction

# Bag-of-Words Approach

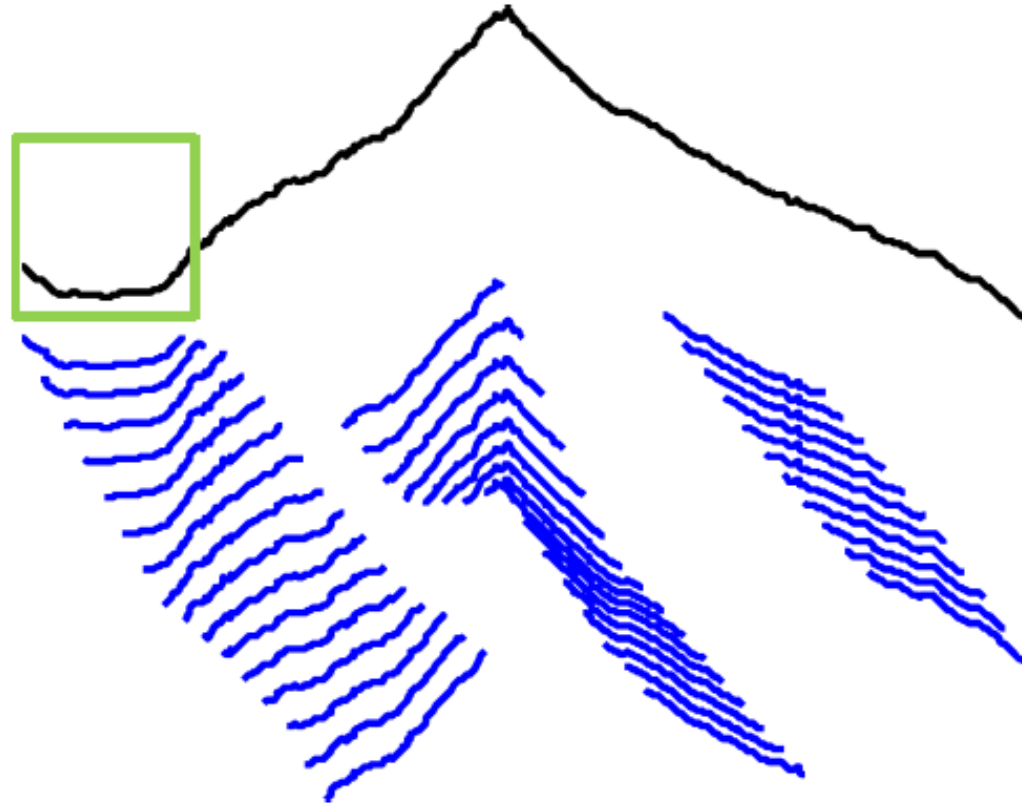
- Uses bag-of-words
  - an image classification method that treats image features as words.
  - a vector of occurrence counts of a vocabulary of local image features.
- Representing the bag of words
  - representation should allow all combinations of words
  - count how often each word occurs in the image
  - the resulting distribution (histogram) represents the image





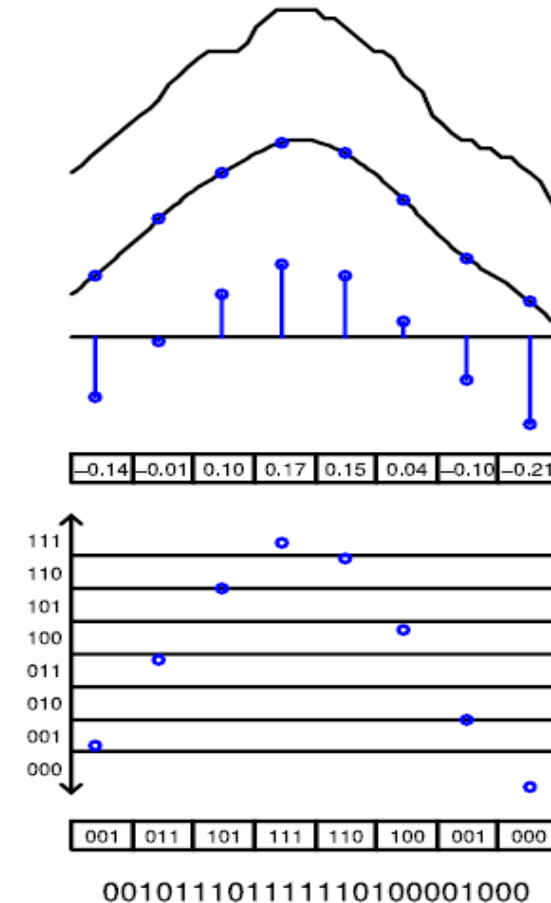
# Horizon Features

- Curvelets
  - Overlapping windows
  - $2.5^\circ / 10^\circ$  wide
  - Step  $1/8$  of width



# Horizon Features (cont'd)

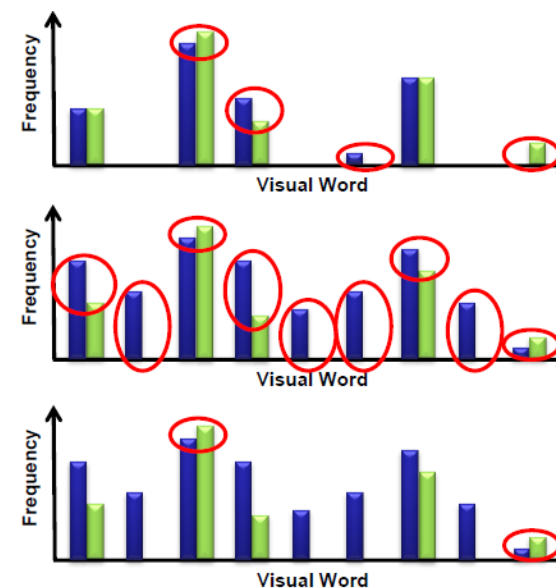
- Feature descriptor
  - Lowpass and subsample
  - Subtract mean and normalize
  - Each component quantized separately
  - Concatenate bin number to get visual word
- Contour word computation
  - raw contour
  - smoothed contour with n sampled points
  - sampled points after normalization
  - contourlet as numeric vector
  - each dimension quantized to 3 bits
  - contour word as 24-bit integer



# “Contains”-Semantics

- Standard bag-of-words
  - Penalizes all differences
  - “Equals”-semantics
- 10°-70° views is compared to 360° panoramas
- Only penalize places where query **exceeds** panorama
  - “Contains”-semantics

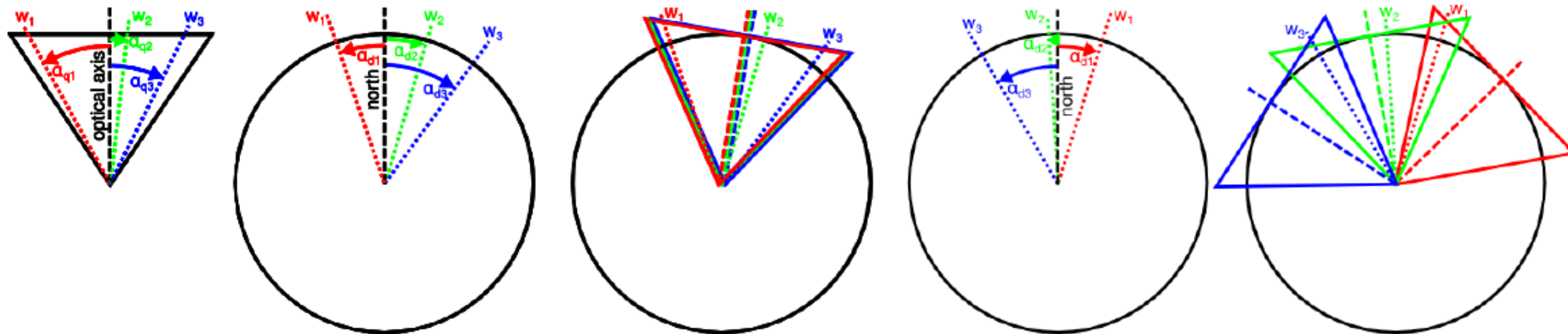
$$D^{E_w}(\tilde{\mathbf{q}}, \tilde{\mathbf{d}}) = \sum_i w_i |\tilde{q}_i - \tilde{d}_i|$$



$$D^C(\mathbf{q}, \mathbf{d}) = \sum_i w_i \max(q_i - d_i, 0)$$



# Voting for Direction

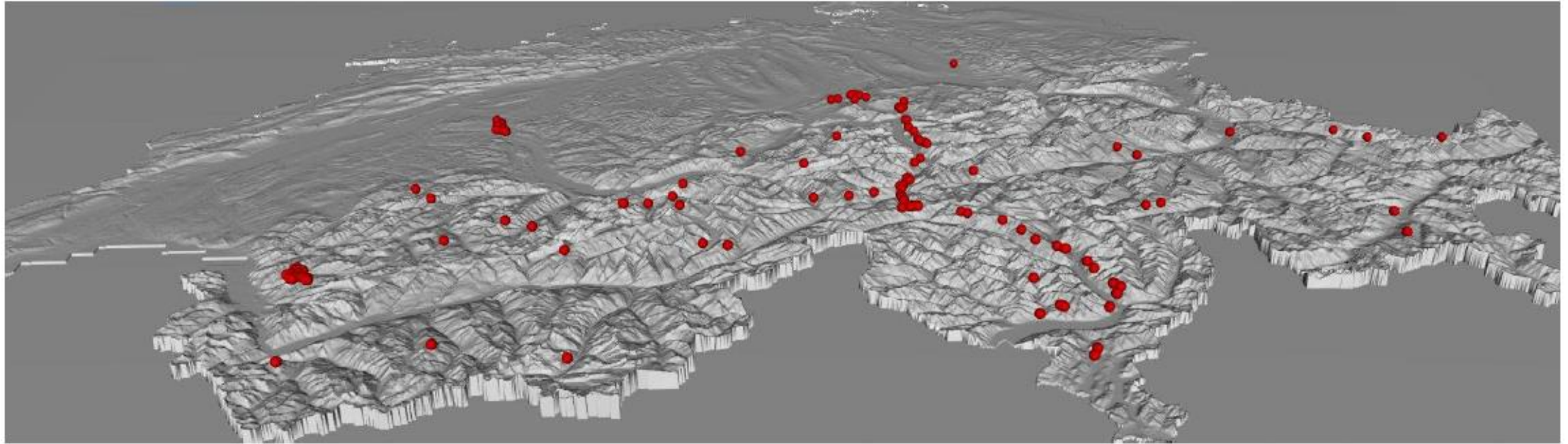


- Each correspondence implies a viewing direction
- Vote for location (panorama) **and** direction
- Enforces same relative order, not just proximity
- Built-in rough geometric verification

# Geometric Verification

- Verify top 1000 candidates
- Iterative Closest Points (ICP)
- **Alignment** of the two visible horizons
- ICP determines a full 3D rotation
- Sample azimuth
- Select best promising one for ICP
- ICP iteratively **reranks** and **moves** the best candidate(s) to the beginning of the short list

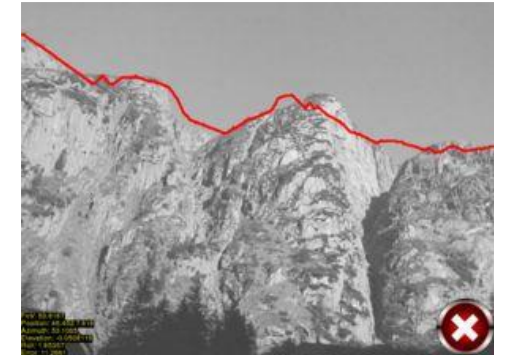
# Evaluation



- Oblique view onto Switzerland model
- The 213 query images' ground truth coordinates
- If location is within **1 km** radius of ground truth images

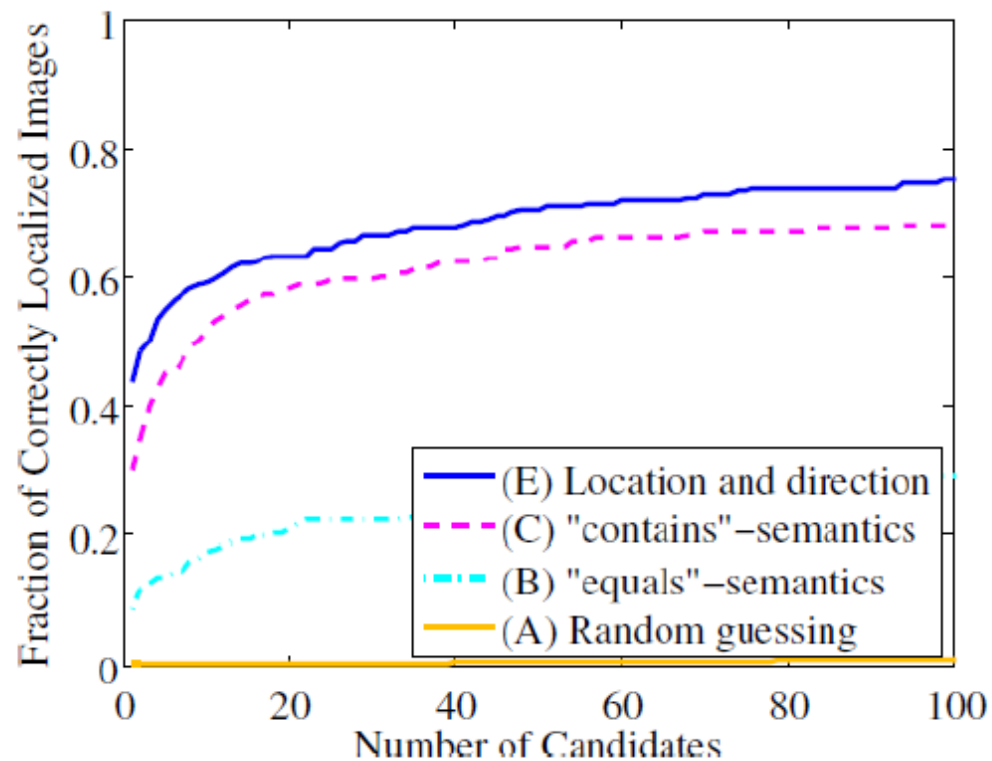


# Results (Qualitative)

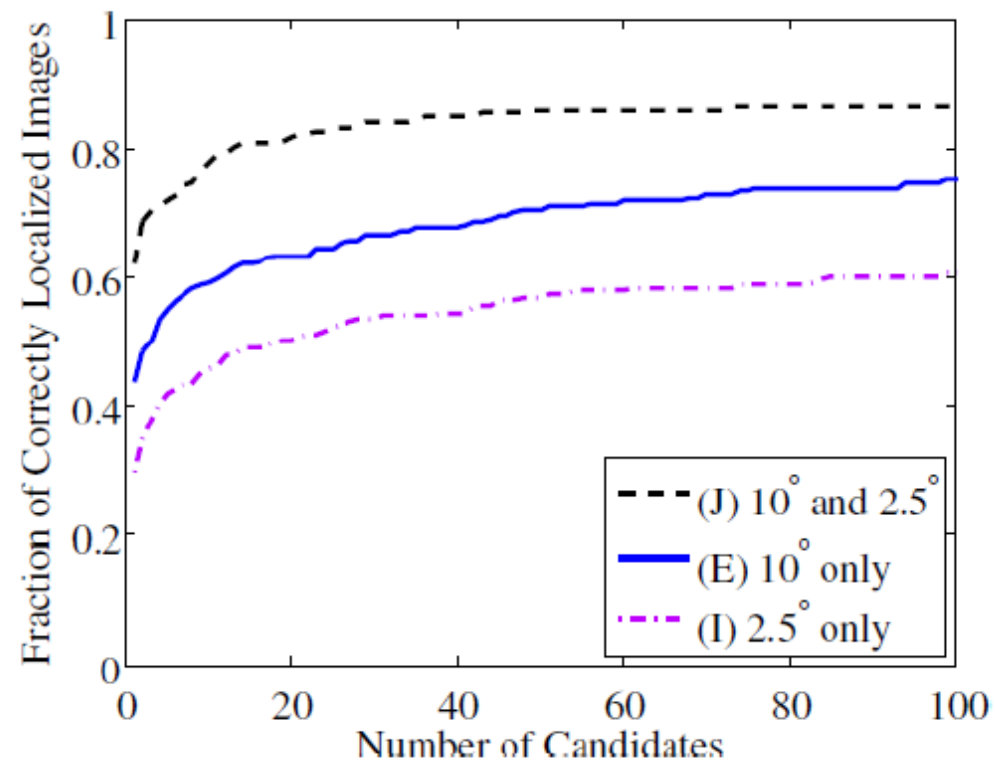


# Results (Quantitative)

## Voting Styles

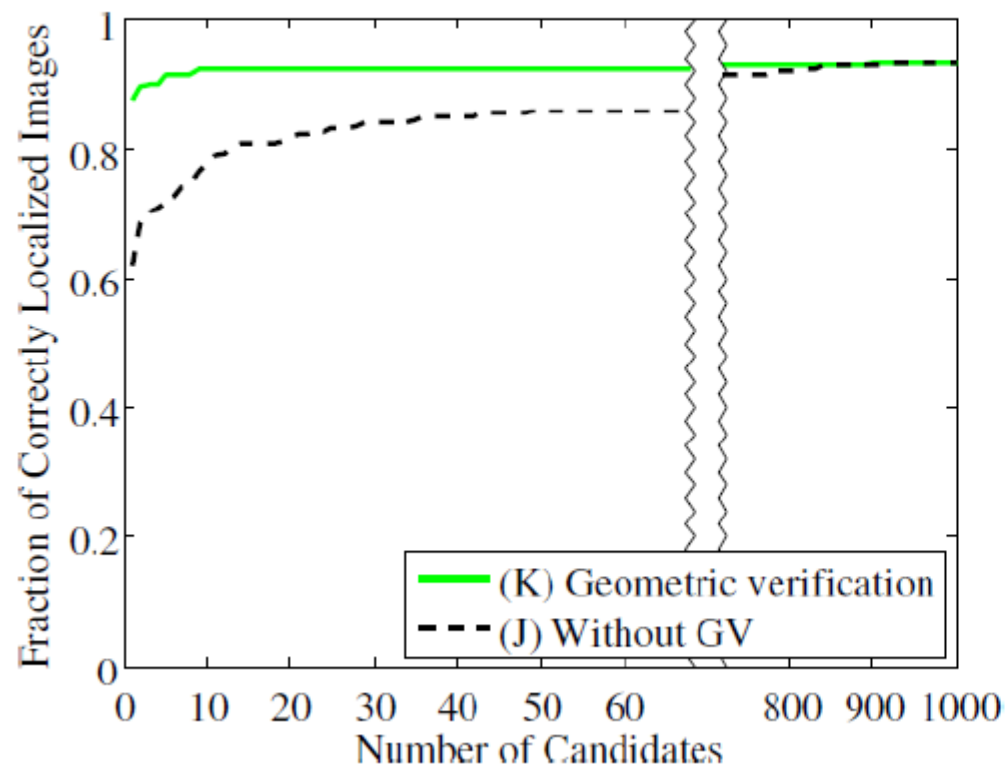


## Descriptor Width

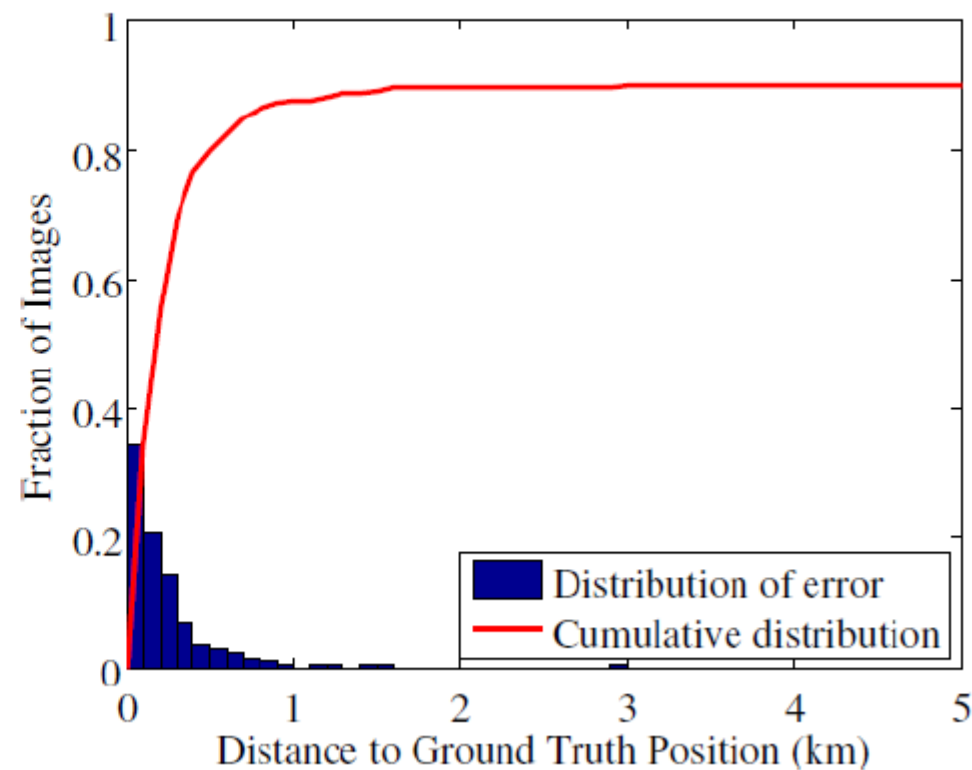


# Results (Quantitative) (cont'd)

## Geometric Verification



## Error Distribution



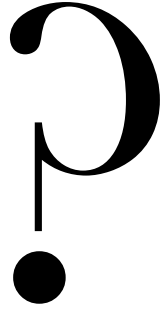
# Discussions

- Needs visible horizon
  - Complement existing methods
  - Geometric Verification does not deal with partial occlusions
- Assume picture taken “close” to ground
  - Can tolerate up to 100m elevation
- Assume that people rarely twist the camera relative to the horizon
  - Small roll
- Requires too much manual labeling, i.e. half of their evaluation data
- Localization takes 10 secs. per image, but no mention on the whole process
- Might exploit other cues, such as type of terrain



# Summary

- Large scale, whole country (Switzerland)
- Dense coverage
- Skyline is very discriminative
- Novel voting schemes
  - Large size differences (“contains”-semantics)
  - Built-in geometric verification (voting for direction)
- Works very well (88% success rate)
- With DEM, might be easily extended to other places



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