Homography

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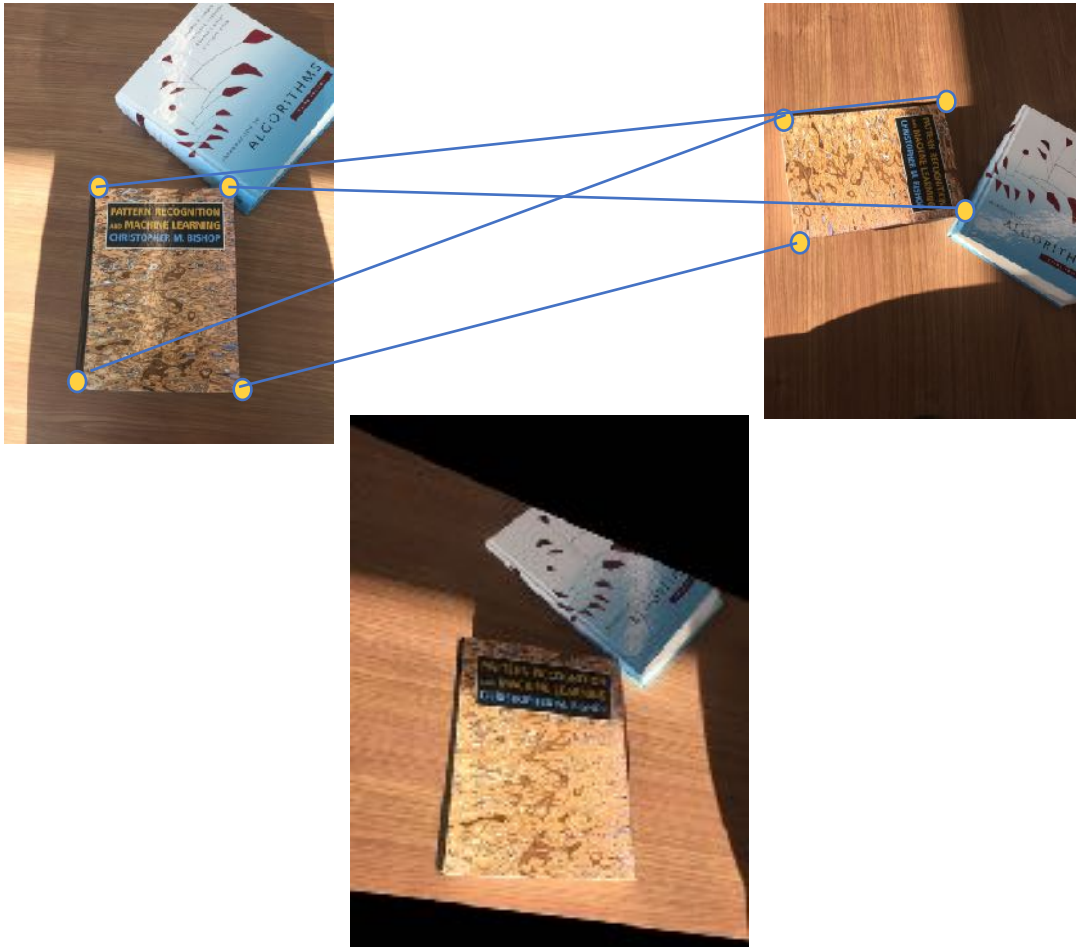
ALGORITHMS
PATTERN RECOGNITION
AND MACHINE LEARNING
CHRISTOPHER M. BISHOP

Homography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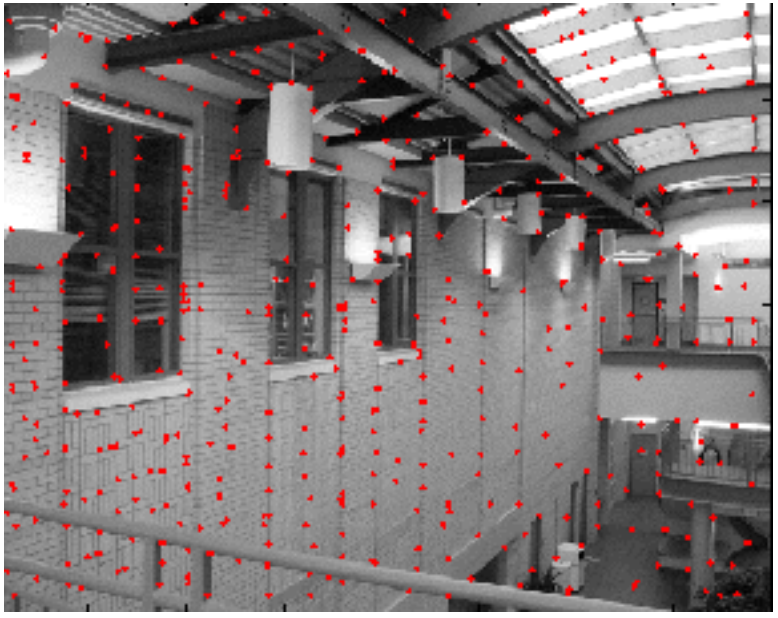
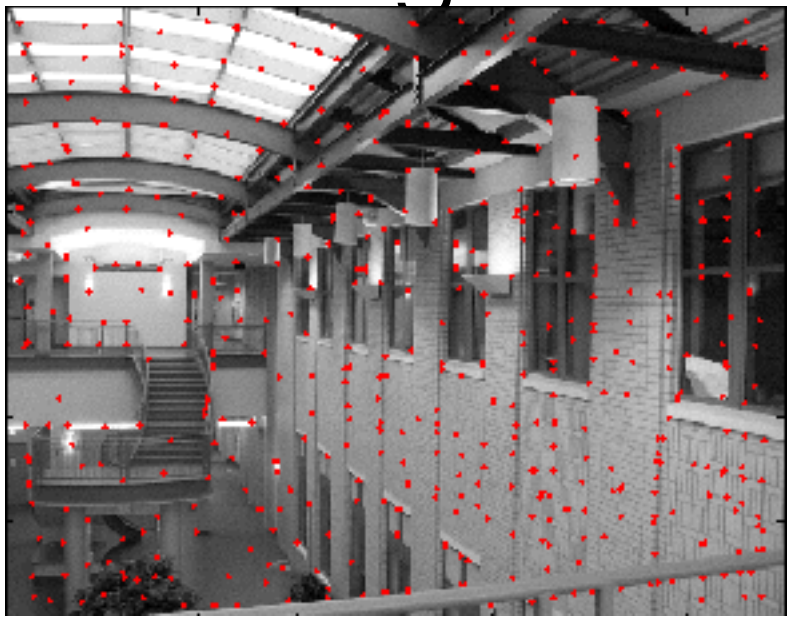
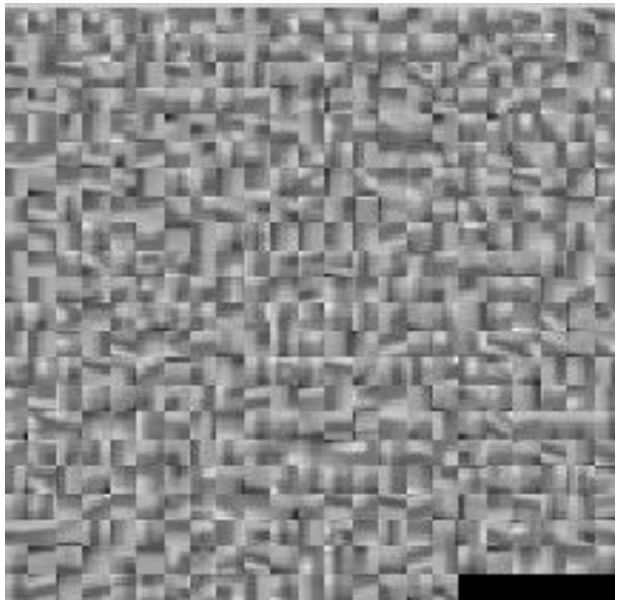
Feature matching



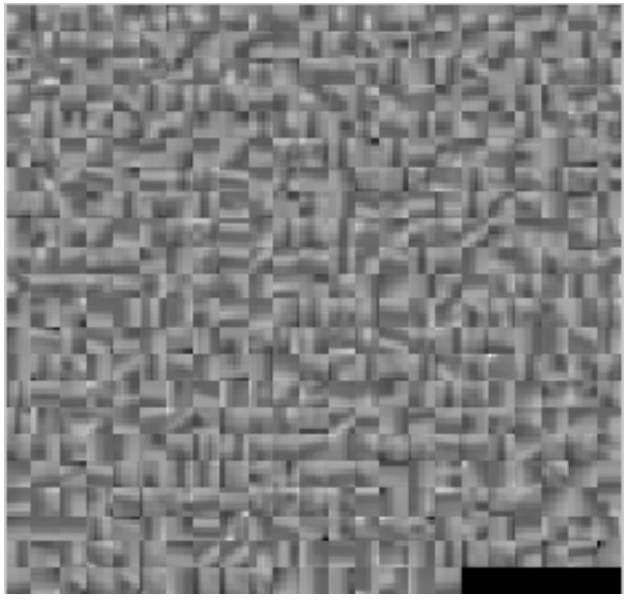


descriptors for left image feature points

descriptors for right image feature points

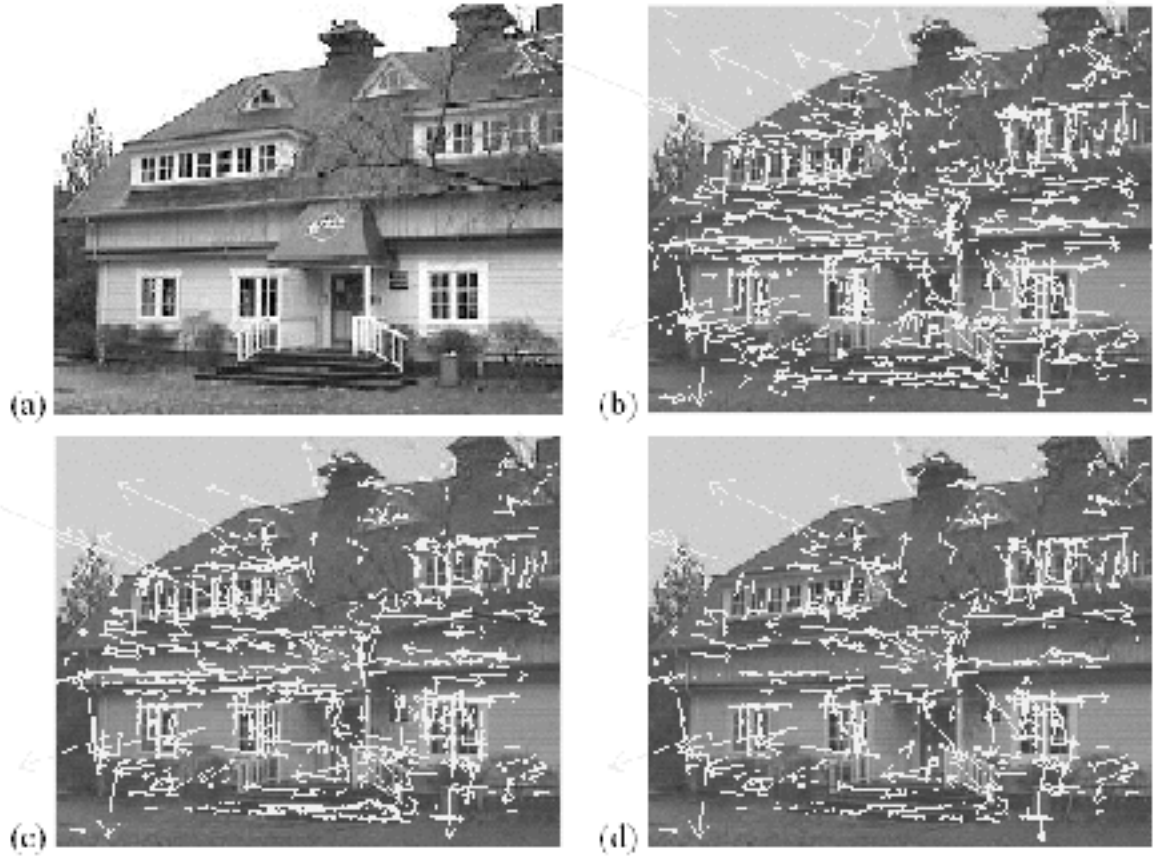


?



SIFT features

- Example



(a) 233x189 image
(b) 832 DOG extrema
(c) 729 left after peak value threshold
(d) 536 left after testing ratio of principle curvatures

Strategies to match images robustly

(a)Working with individual features: For each feature point, find most similar point in other image (SIFT distance)

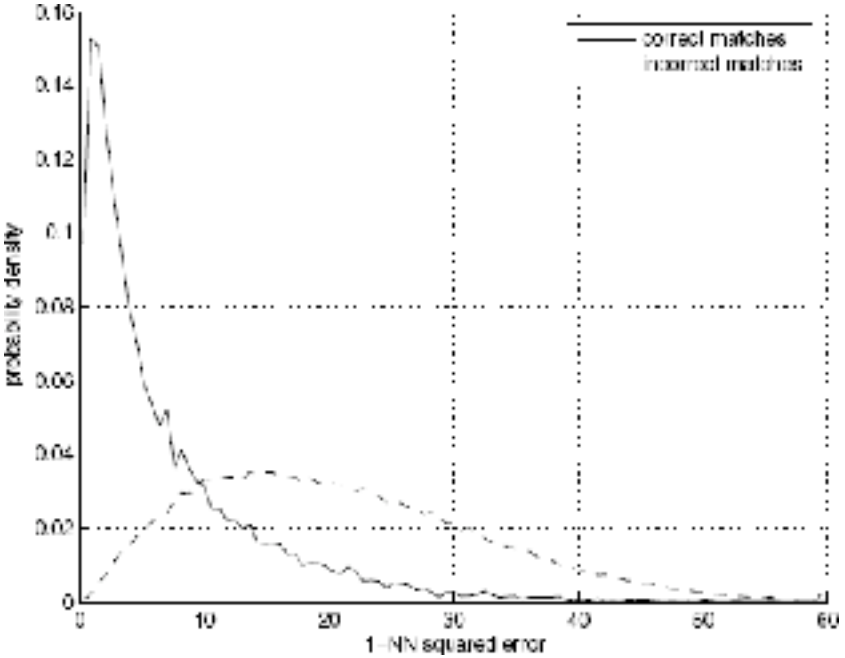
Reject ambiguous matches where there are too many similar points

(b)Working with all the features: Given some good feature matches, look for possible homographies relating the two images

Reject homographies that don't have many feature matches.

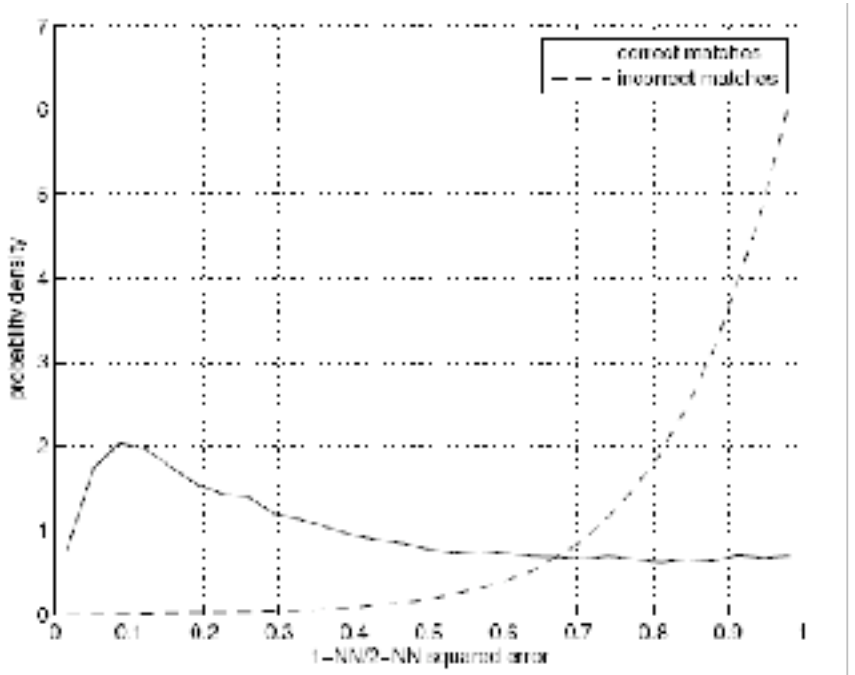
(a) Feature-space outlier rejection

- Let's not match all features, but only these that have "similar enough" matches?
- How can we do it?
  - SSD(patch1,patch2) < threshold
  - How to set threshold?
    Not so easy.

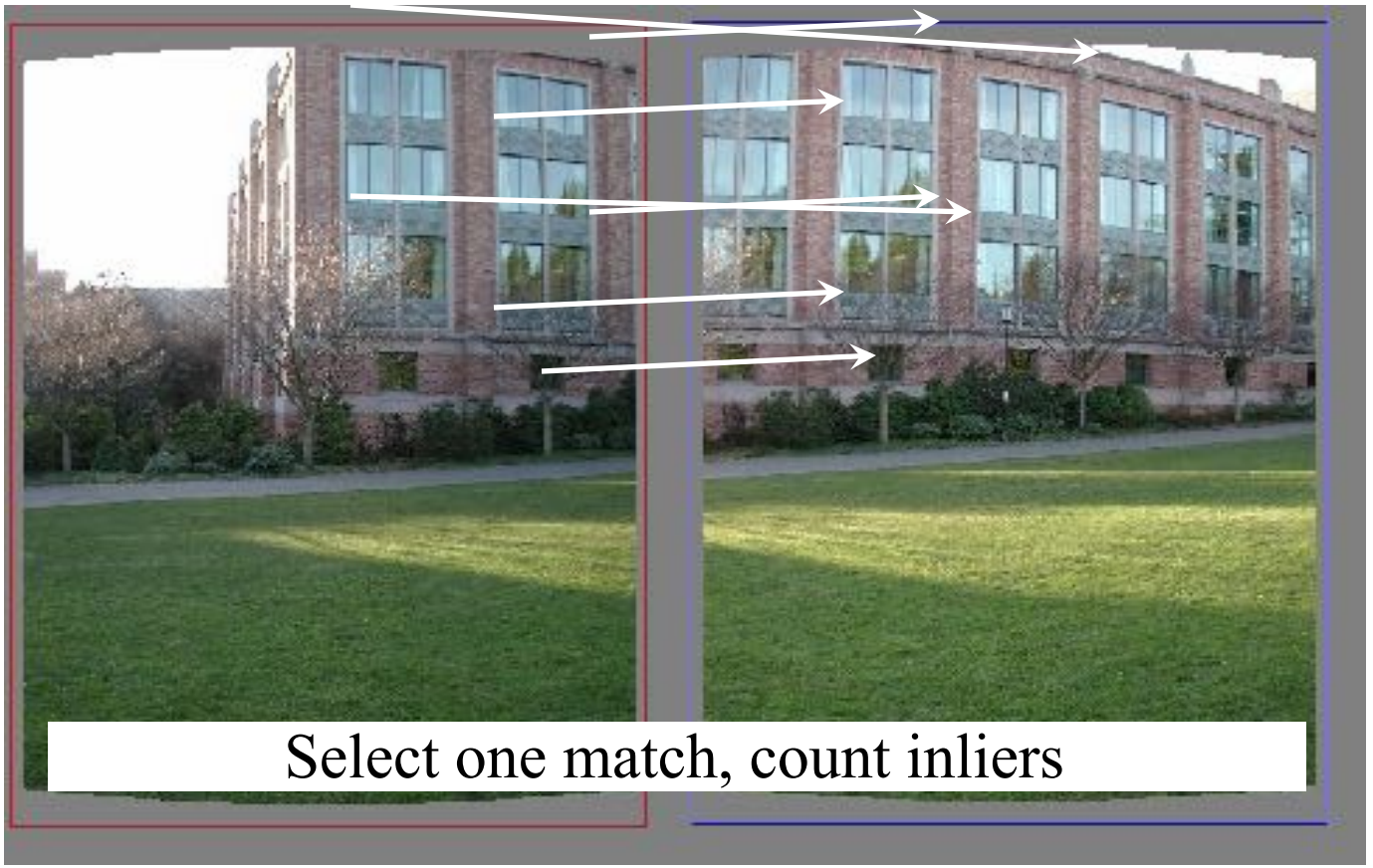


Feature-space outlier rejection

- A better way [Lowe, 1999]:
  - 1-NN: SSD of the closest match
  - 2-NN: SSD of the second-closest match
  - Look at how much better 1-NN is than 2-NN, e.g. 1-NN/2-NN
  - That is, is our best match so much better than the rest?



RAndom SAmple Consensus



RANSAC for estimating homography

RANSAC loop:

Select four feature pairs (at random)

Compute homography H (exact)

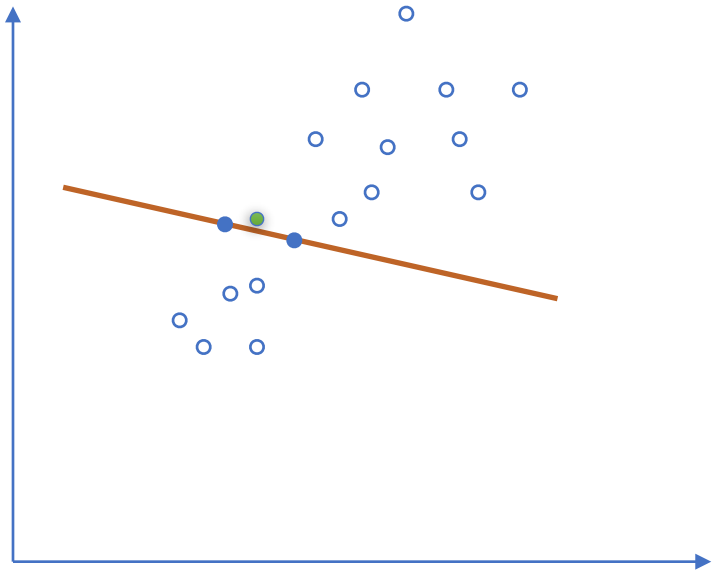
Compute inliers where $||p_i', H p_i|| < \varepsilon$

Keep largest set of inliers

Re-compute least-squares H estimate using all of the inliers

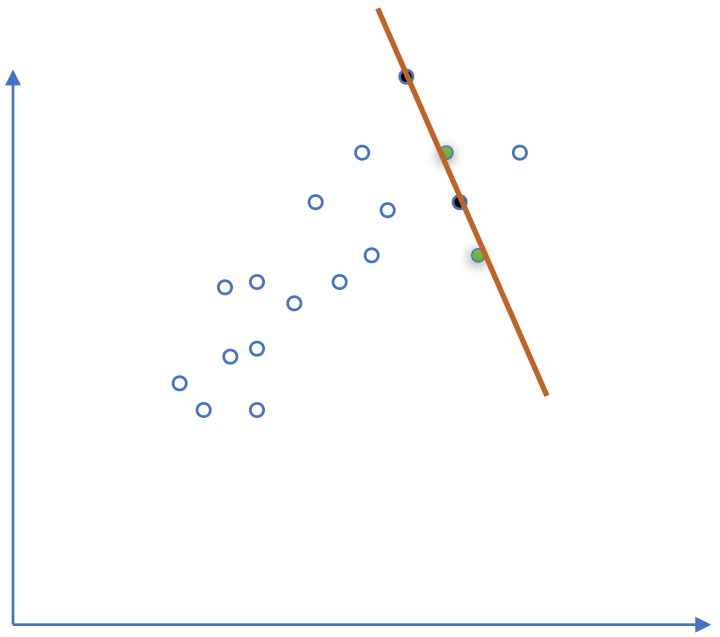
Fit a line

- Pick 2 points
- Fit line
- Count inliers (3 inliers)



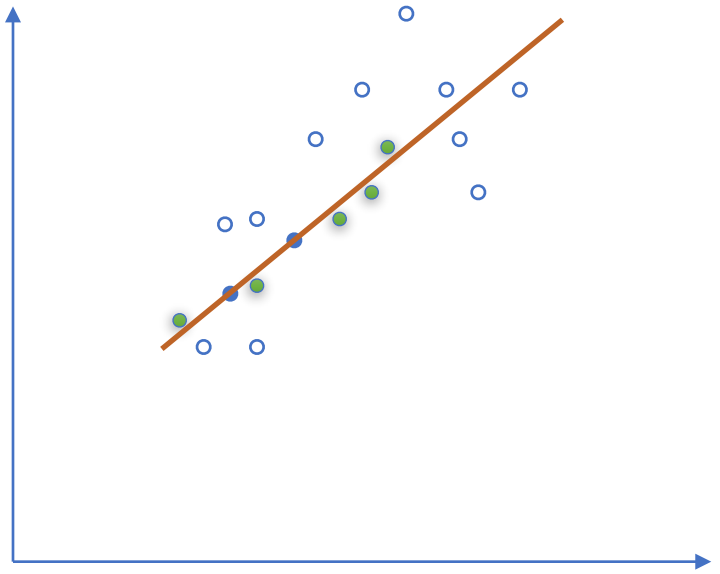
Fit a line

- Pick 2 points
- Fit line
- Count inliers (4 inliers)



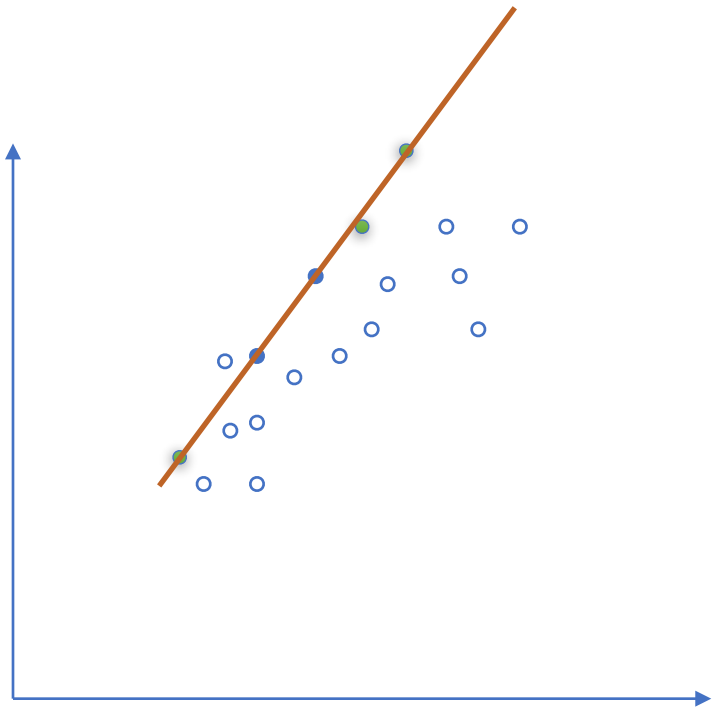
Fit a line

- Pick 2 points
- Fit line
- Count inliers (7 inliers)



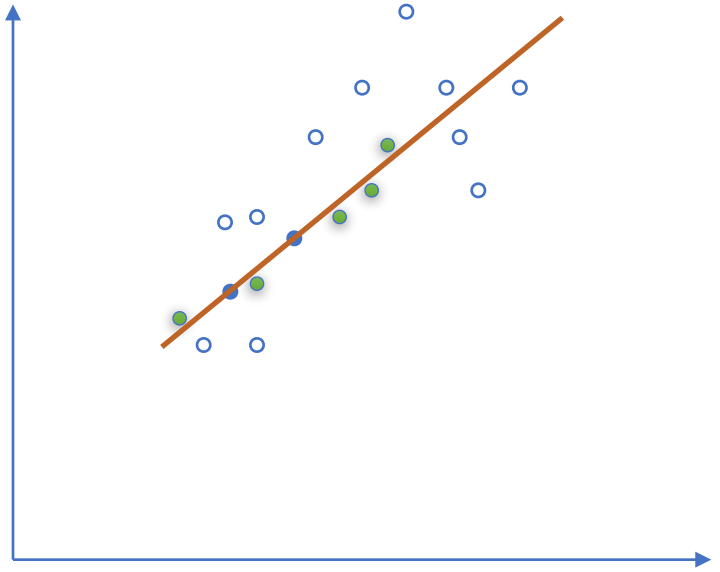
Fit a line

- Pick 2 points
- Fit line
- Count inliers (5 inliers)

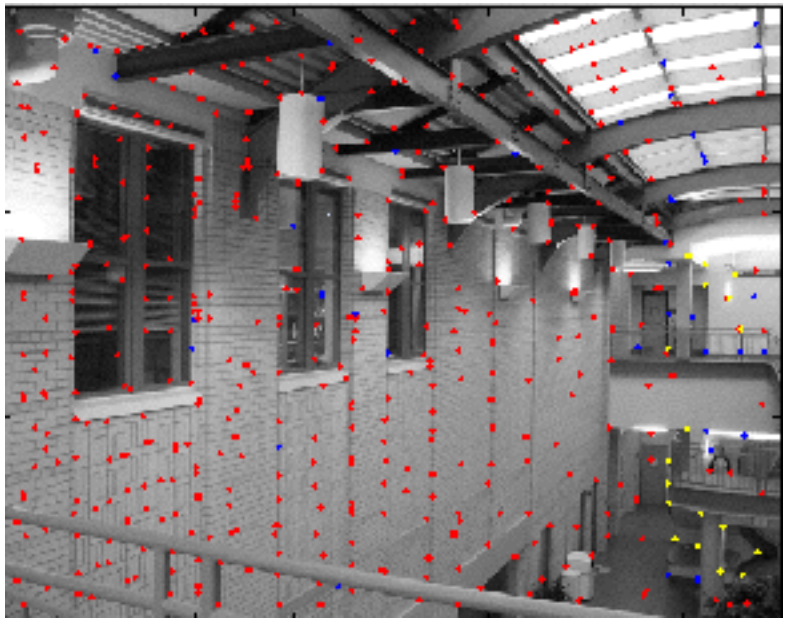


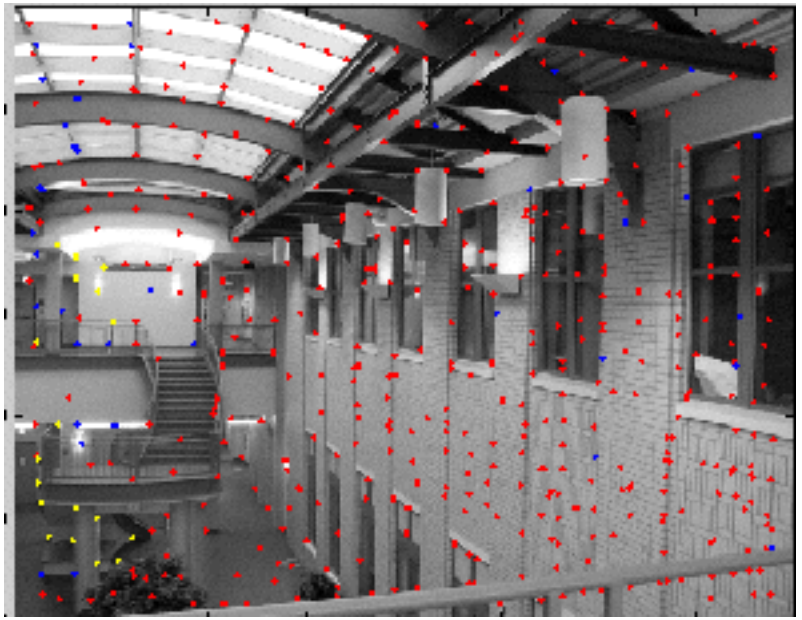
Fit a line

- Use biggest set of inliers
- Least squares fit



RANSAC





red:
rejected by 2nd nearest neighbor criterion
blue:
Ransac outliers
yellow:
inliers

RANSAC parameters

Decision Threshold: $|X' - HX| < \epsilon$

Number of trials: N

Size of the inlier set: M

Robustness

- Proportion of inliers in our pairs is g (for "g o o d")
- Our model needs P pairs
  - P=4 for homography
- Probability that we pick P inliers?

  g^p
- Probability that after N RANSAC iterations we have not picked a set of inliers?
  - $(1 - g^p)^N$

Robustness: example

- Proportion of inliers G=0.5
- Probability that we pick P=4 inliers?
  - $0.5^4 = 0.0625(6\,\%\,chance)$
- Probability that we have not picked a set of inliers?

$N = 100$ iterations

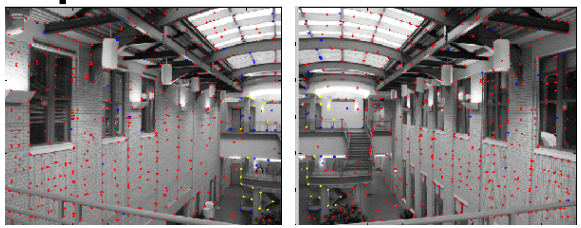
$(1 - 0.5^4)^{100} = 0.00157(1 \text{ chance in } 600)$

$N = 1000$ iterations

$(1 - 0.5^4)^{100} = 0.00157(1 \text{ chance in } 10^{28})$

1 chance in 1e28

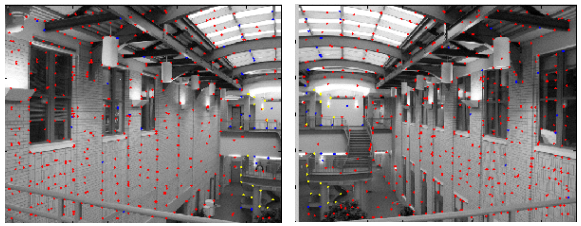
Robustness: example



- Proportion of inliers G=0.3
- Probability that we pick P=4 inliers?
  - 0.3^4=0.0081 (0.8% chance)
- Probability that we have not picked a set of inliers?
  - N=100 iterations:
    $(1-0.3^4)^{100}$=0.44 (1 chance in 2)
  - N=1000 iterations:

    1 chance in 3400

Robustness: example



- Proportion of inliers G=0.1
- Probability that we pick P=4 inliers?
  - 0.1^4=0.0001 (0.01% chances, 1 in 10,000)
- Probability that we have not picked a set of inliers?
  - N=100 iterations: $(1-0.1^4)^{100}$=0.99
  - N=1000 iterations: 90%
  - N=10,000: 36%
  - N=100,000: 1 in 22,000

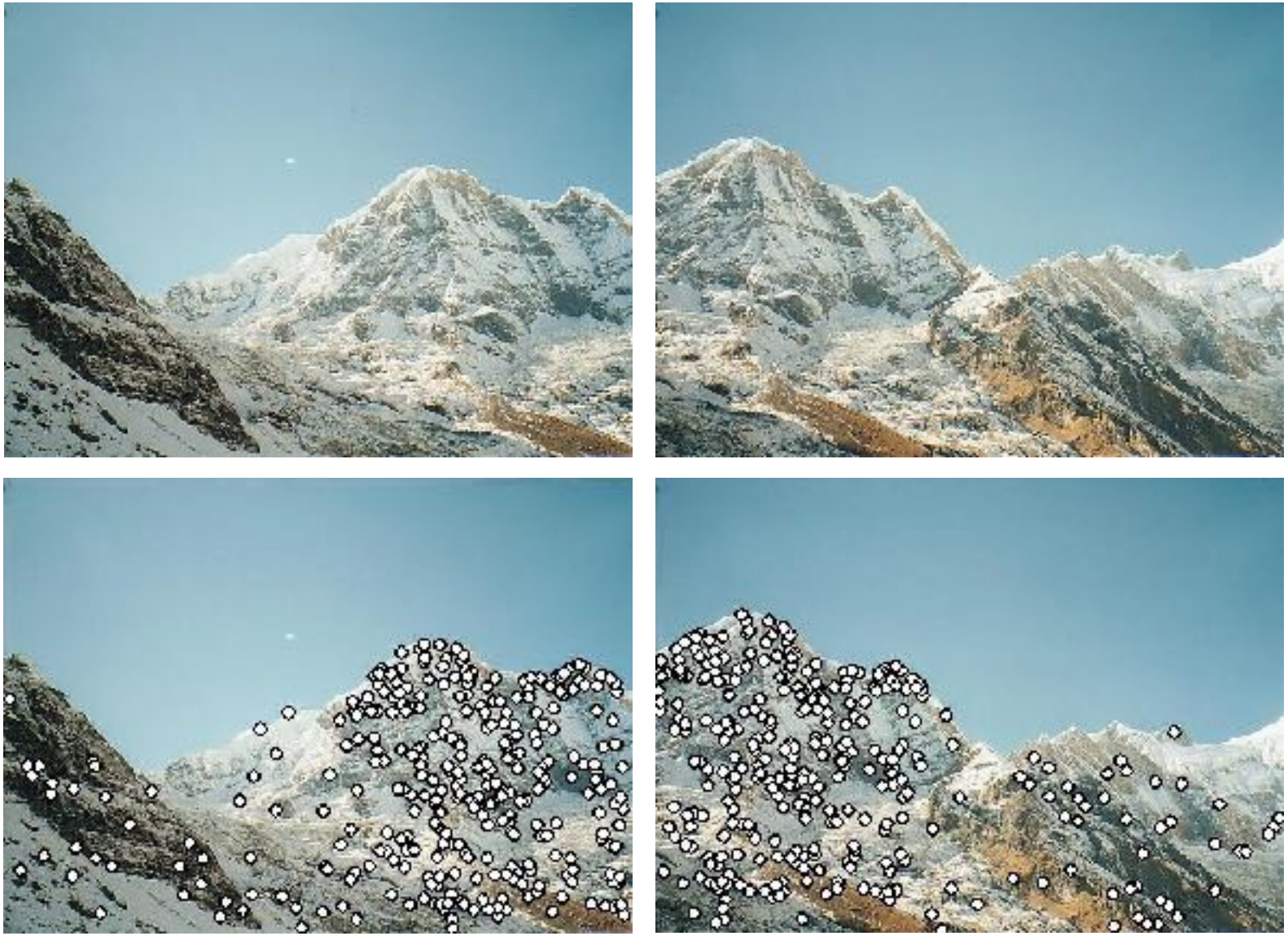
Robustness: conclusions

- Effect of number of parameters of model/ number of necessary pairs
  - Bad exponential
- Effect of percentage of inliers
  - Base of the exponential
- Effect of number of iterations
  - Good exponential

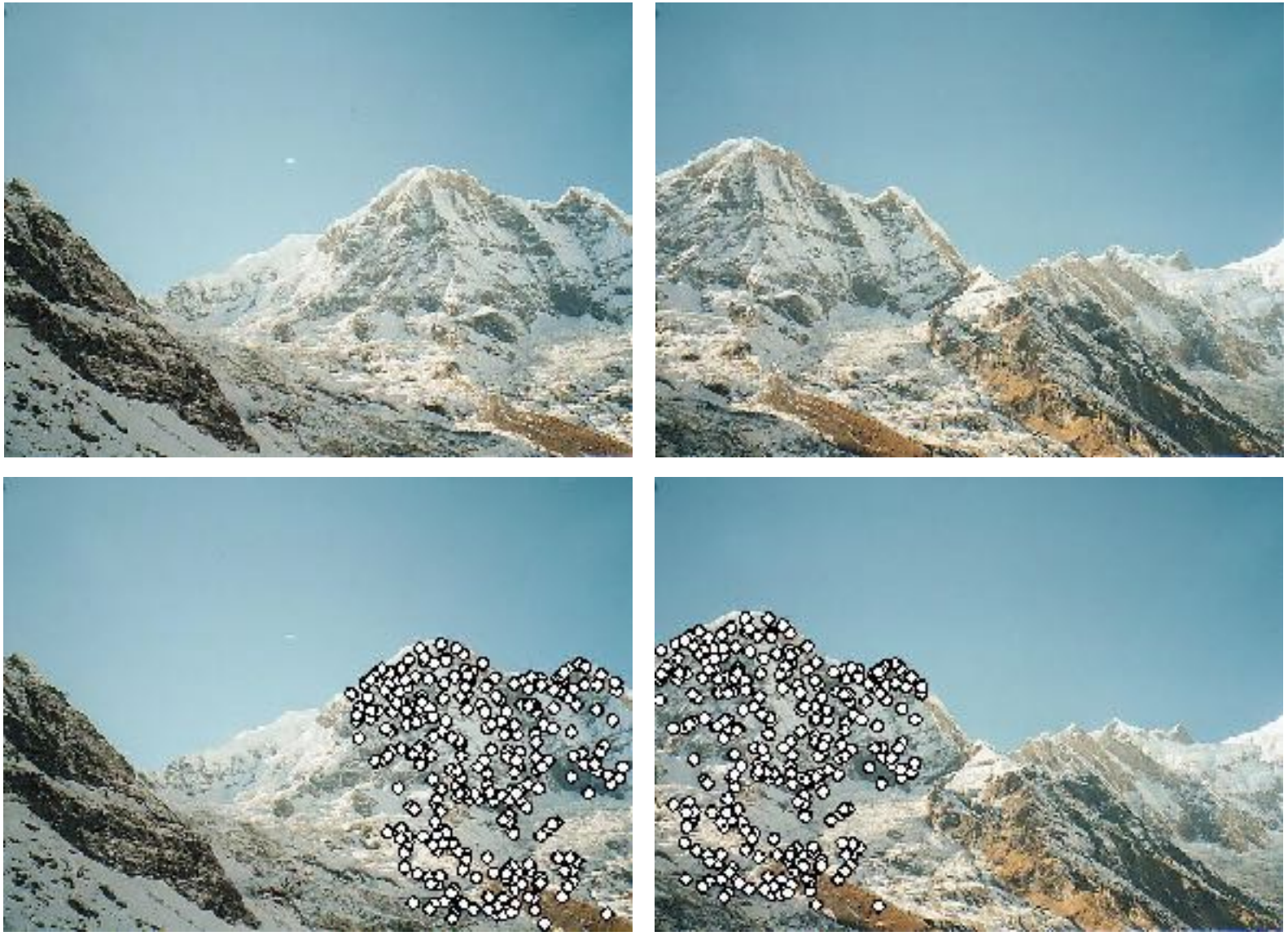
RANSAC recap

- For fitting a model with low number P of parameters (8 for homographies)
- Loop
  - Select P random data points
  - Fit model
  - Count inliers
    (other data points well fit by this model)
- Keep model with largest number of inliers

RANSAC for Homography



RANSAC for Homography



RANSAC for Homography