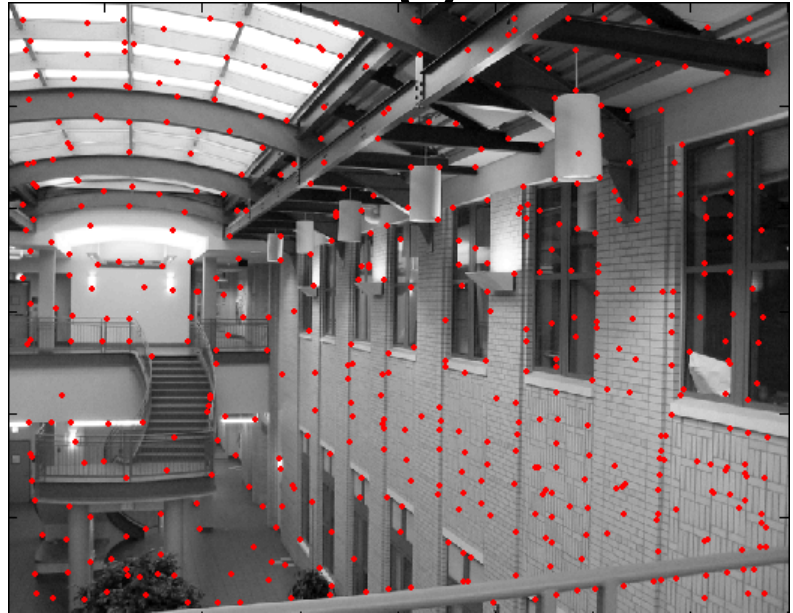
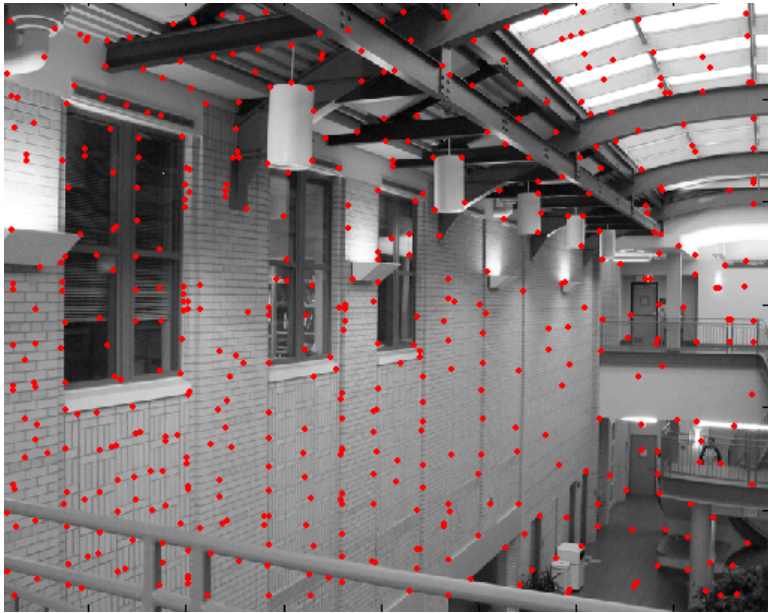
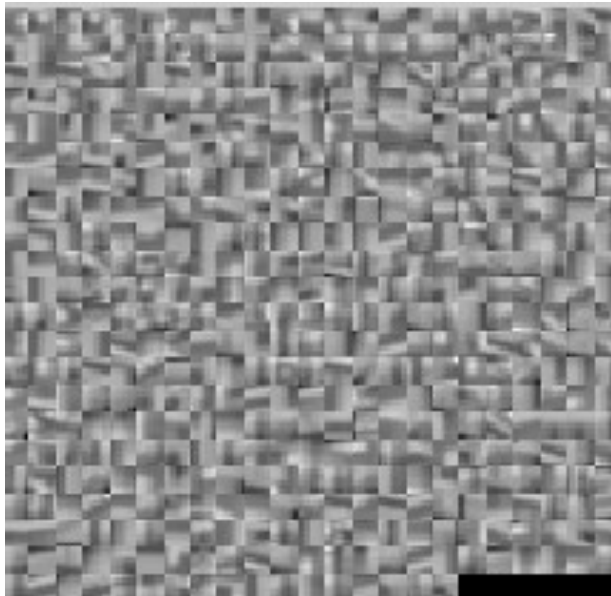


RANSAC

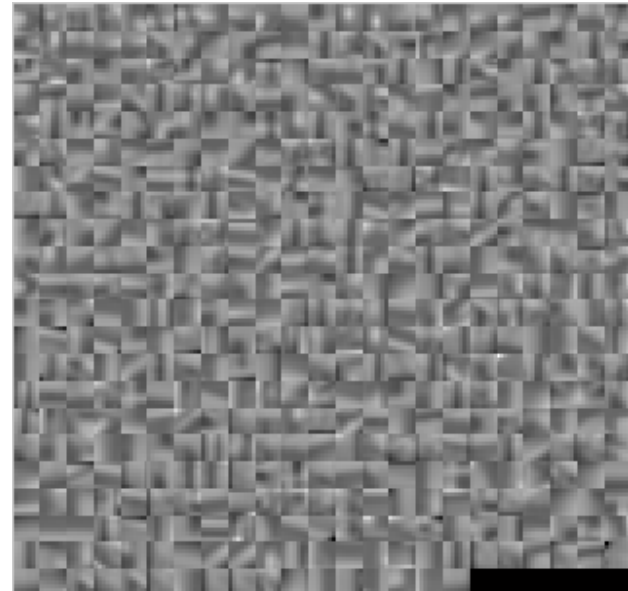
Feature matching



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)

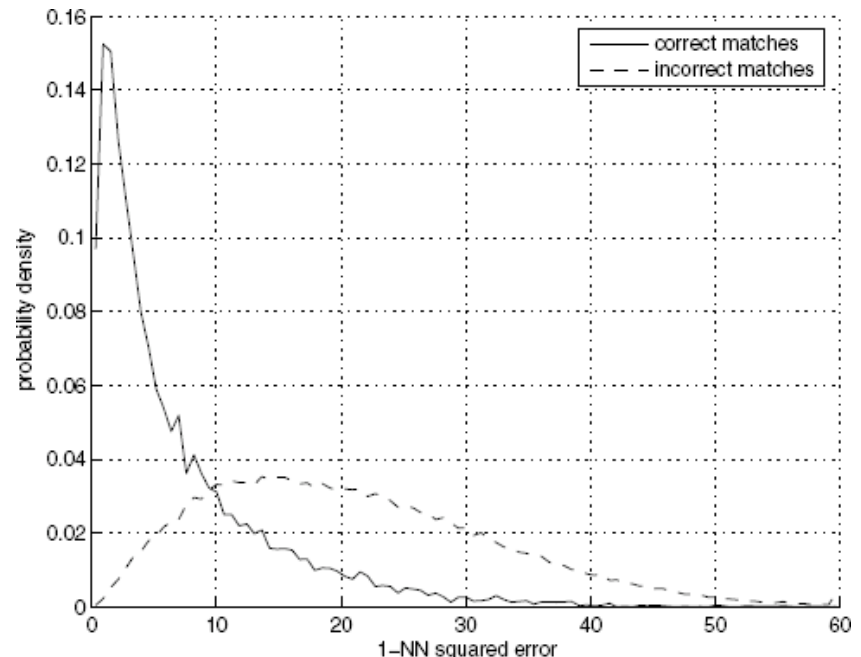
Reject ambiguous matches where there are too many similar points

(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.

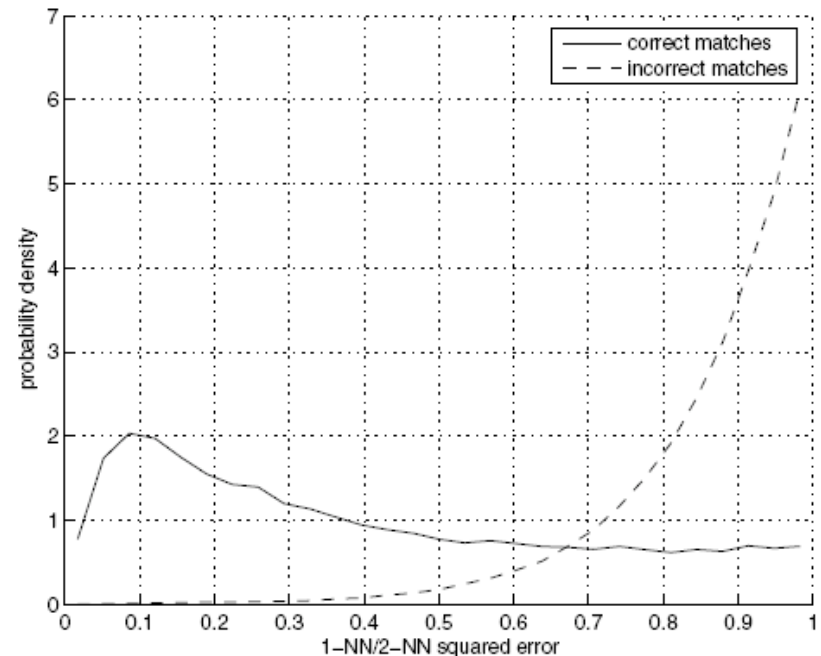
(a) Feature-space outlier rejection

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

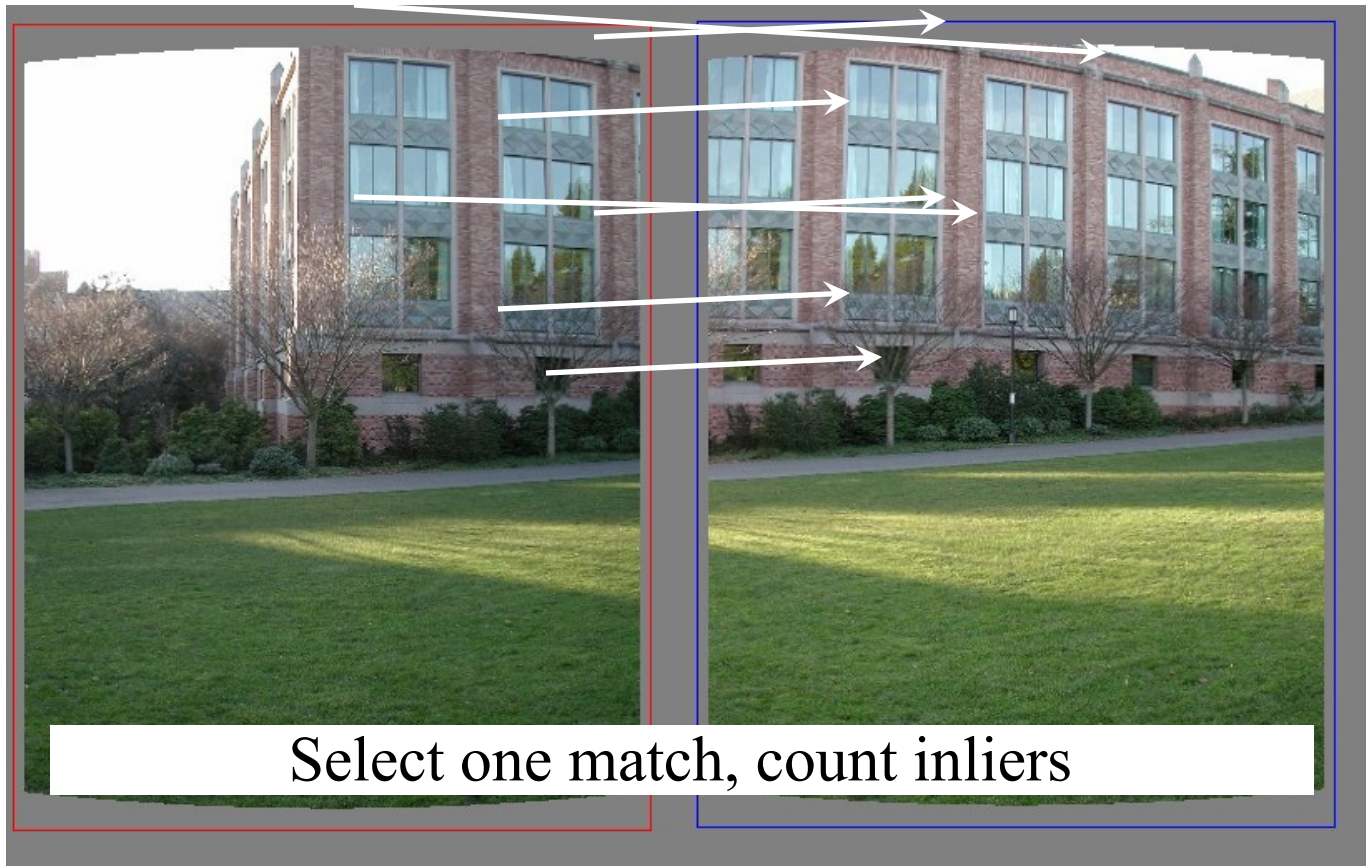


Feature-space outlier rejection

- 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. $1\text{-NN}/2\text{-NN}$
 - That is, is our best match so much better than the rest?

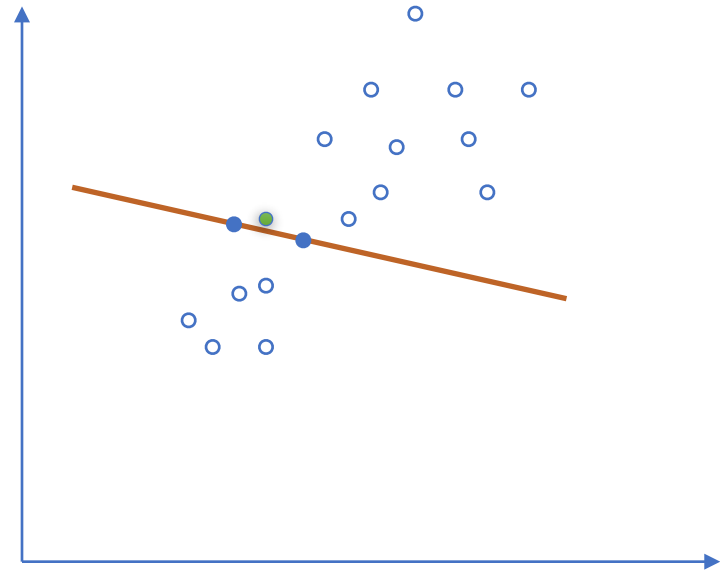


Random Sample Consensus



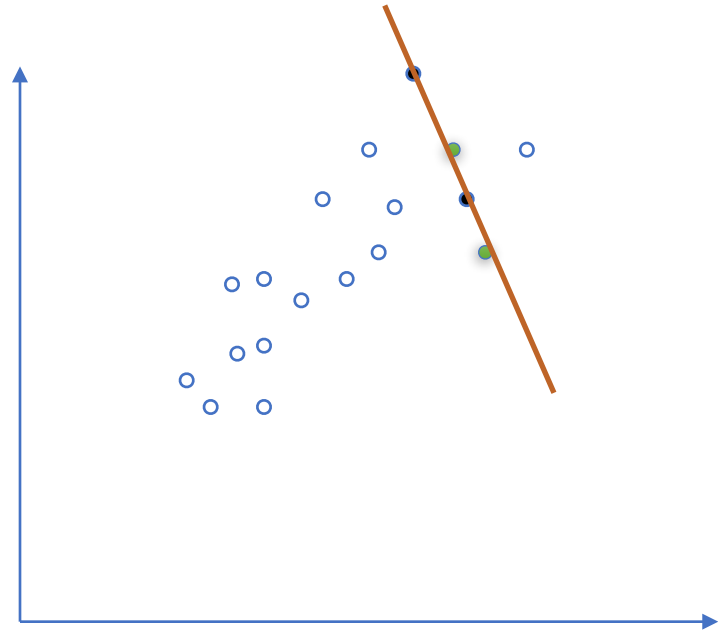
Fit a line

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



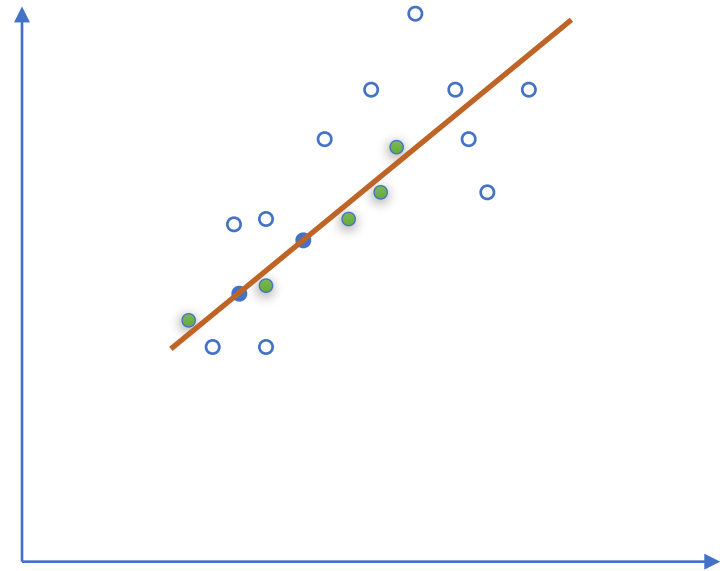
Fit a line

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



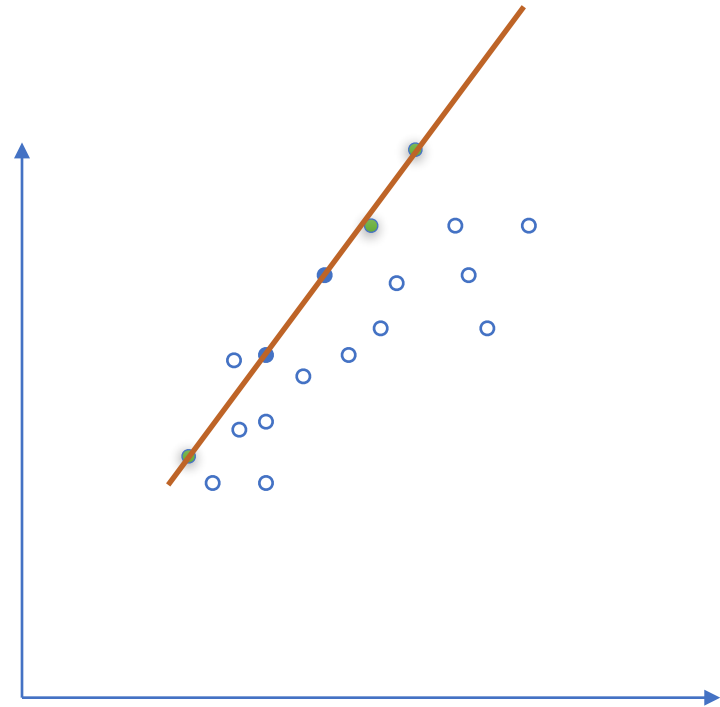
Fit a line

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



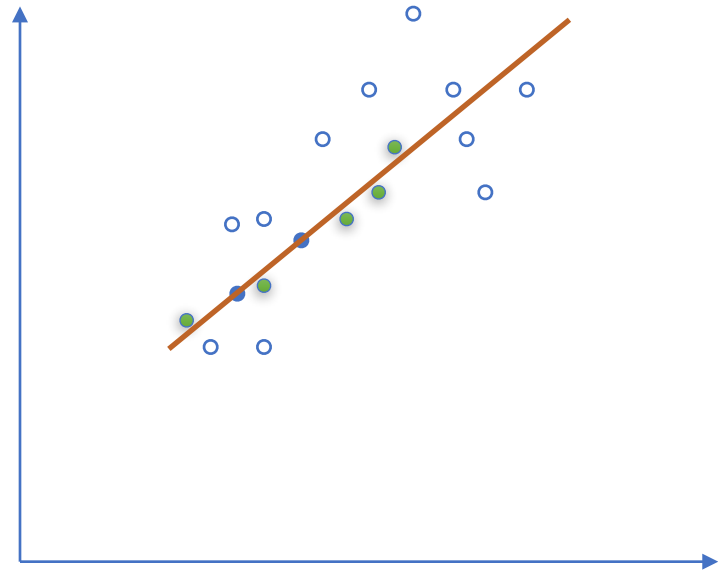
Fit a line

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



Fit a line

- Use biggest set of inliers
- Least squares fit



RANSAC for estimating homography

Determine:

n —the smallest number of points required (e.g., for lines, $n = 2$,
for circles, $n = 3$) **$n=8$ for homography**

k —the number of iterations required

t —the threshold used to identify a point that fits well

d —the number of nearby points required to assert a model fits well **Size of Inlier set**

Until k iterations have occurred

Draw a sample of n points from the data
uniformly and at random

Fit to that set of n points **Compute homography, H**

For each data point outside the sample

Test the distance from the point to the structure **Decision Threshold:** $|p' - Hp| < t$
against t ; if the distance from the point to the structure
is less than t , the point is close

end

If there are d or more points close to the structure
then there is a good fit. Refit the structure using all
these points. Add the result to a collection of good fits.

end

Use the best fit from this collection, using the
fitting error as a criterion

Algorithm 10.4: RANSAC: Fitting Structures Using Random Sample Consensus.

Forsyth & Ponce

RANSAC for estimating homography

RANSAC loop:

Select four feature pairs (at random)

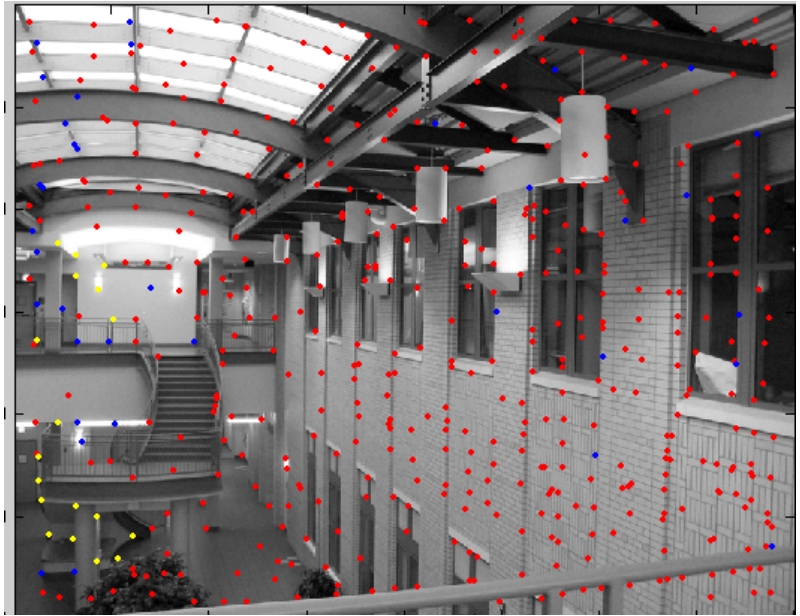
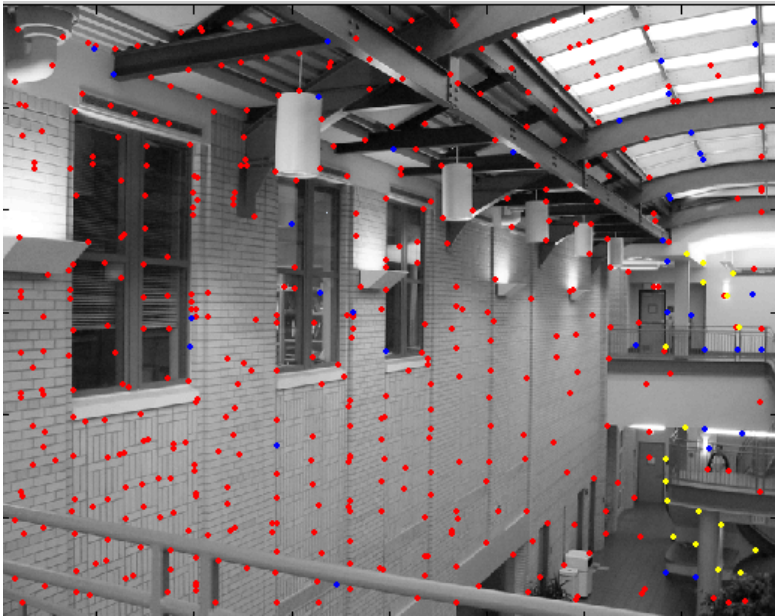
Compute homography H (exact)

Compute inliers where $||p' - Hp|| < t$

Keep largest set of inliers

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

RANSAC



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



Robustness

- Proportion of inliers in our pairs is g (for “g o o d ”)
- Our model needs n pairs
 - $n = 4$ for homography
- Probability that we pick n inliers?
 g^n
- Probability that after k RANSAC iterations we have not picked a set of inliers?
 - $(1 - g^n)^k$

Robustness: example

- Proportion of inliers $g=0.5$
- Probability that we pick $n=4$ inlier pairs?
 - $0.5^4 = 0.0625$ (6 % *chance*)
- Probability that we have not picked a set of inliers?

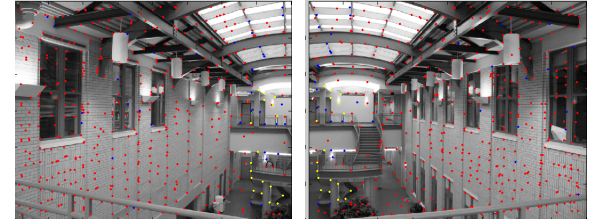
$k = 100$ iterations

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

$k = 1000$ iterations

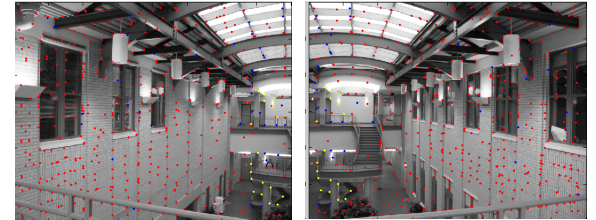
$$(1 - 0.5^4)^{1000} = 0.00157 \text{ (1 chance in } 10^{28}\text{)}$$

Robustness: example



- Proportion of inliers $g=0.3$
- Probability that we pick $n=4$ inlier pairs?
– $0.3^4 = 0.0081$ (0.8% chance)
- Probability that we have not picked a set of inliers?
 - $k=100$ iterations: $(1 - 0.3^4)^{100} = 0.44$ (1 chance in 2)
 - $k=1000$ iterations: $(1 - 0.3^4)^{1000} \rightarrow$ (1 chance in 3400)

Robustness: example



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

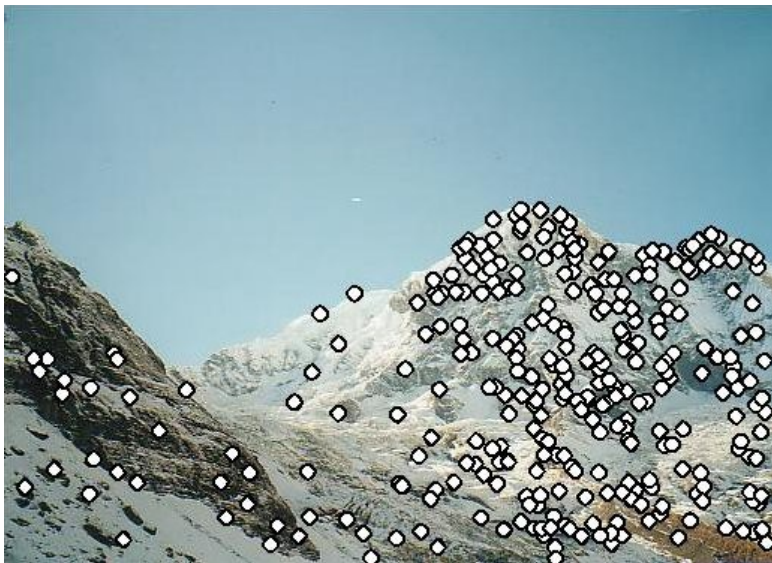
Robustness: conclusions

- 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

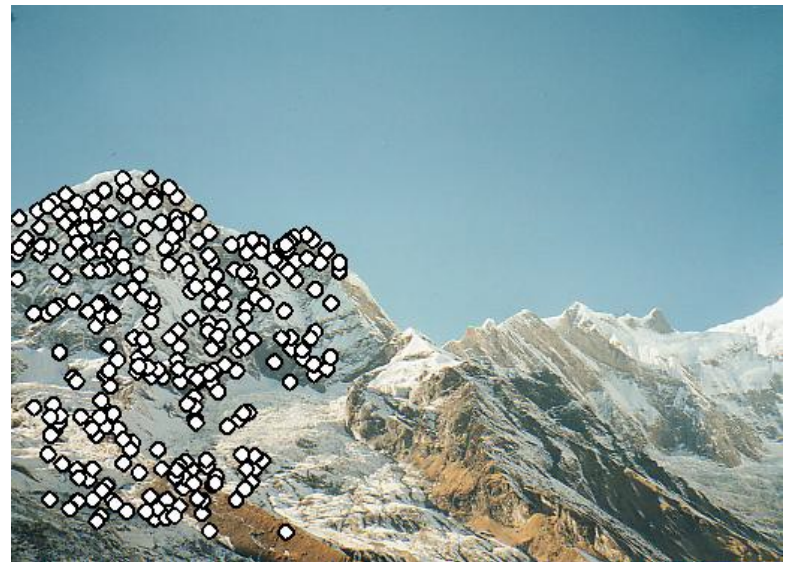
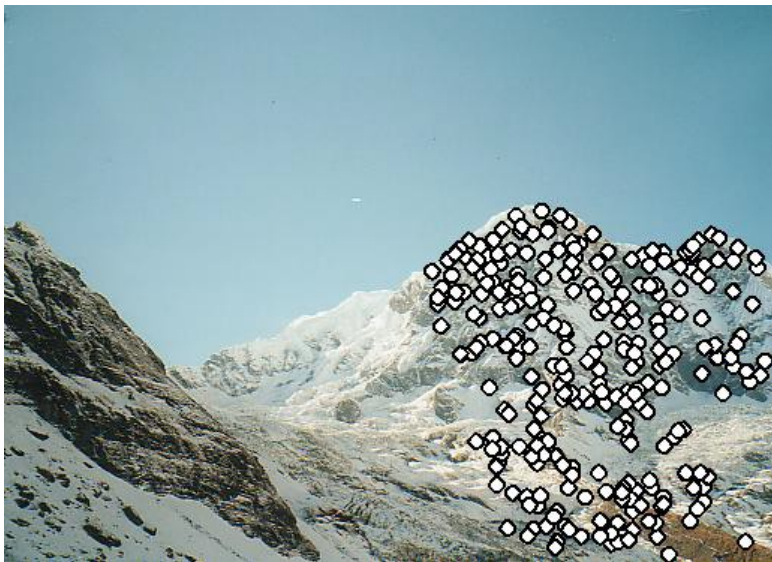
RANSAC recap

- For fitting a model with low number n of parameters (4 pairs for homographies)
- Loop
 - Select n 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

