# **Projection Models and Homogeneous Coordinates**

RANSAC Algorithm

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#### **Topics**

RANSAC - RANdom SAmple Consensus

Further Readings





#### RANSAC – RANdom SAmple Consensus

**Problem:** In calibration we have to deal with inaccuracies in observations and outliers in the data.

There are two types of outliers:

- badly localized points, and
- wrong correspondences.





#### **Outliers in Linear Regression**

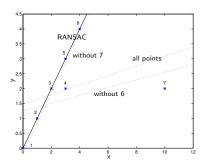


Figure 1: Example of the influence of an outlier in linear regression (least squares method)





### **RANSAC Algorithm**

- 1. Draw samples uniformly and at random from the input data set.
- Cardinality of sample set is the smallest size sufficient to estimate the model parameters.
- 3. Compute the model parameters for each element of the sample data.
- 4. Evaluate the quality of the hypothetical models on the full data set.
  - Cost function for the evaluation of the quality of the model
  - Inliers: data points which agree with the model within an error tolerance
- 5. The hypothesis which gets the most support from the data set is taken as the best estimate.





## How Many Iterations? When Do We Need to Stop?

**Problem:** If not run often enough, we probably still have outliers. **Goal:** Find a model that is determined only from inliers after N iterations.

- Model estimation requires K points.
- p(x): probability that x is an inlier
- p(y): prob. that at least one model that consists only of inliers is picked

Bernoulli trial:  $1 - p(x)^K \rightarrow$  at least 1 out of K points is an outlier After *N* iterations:  $(1 - p(x)^K)^N \rightarrow \text{prob.}$  that all *N* models contain outliers

$$\Rightarrow$$
 1- $p(y) = (1-p(x)^K)^N$ 

We solve the logarithmized equation for N:

$$N = \frac{\log(1 - p(y))}{\log(1 - p(x)^K)}$$





#### **Example**

#### Let us consider that

- the number of model observations is 1000, and
- the number of inliers is only 100 (a worst case scenario, p(x) = 10%).

#### Further assume:

- we have a parabolic model (K = 3), and
- p(y) = 99.99999%.
  - N must be at least 16110.





### **Topics**

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**Further Readings** 





## **Further Readings**

#### For the original work see:

Martin A. Fischler and Robert C. Bolles. "Random Sample Consensus: A Paradigm for Model Fitting with Applications to Image Analysis and Automated Cartography". In: CACM 24.6 (June 1981), pp. 381–395.

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