

# RANSAC

①

for  $i = 0 \dots K$ :

$d_i \in S$  random data points

$M_i \leftarrow \text{fit\_model}(d_i)$

inlier-count  $\leftarrow \sum_{x_i, y_i \in D} \mathbb{1}(|M(x_i) - y_i| < \delta)$  #data points in agreement w/  $M_i$

if inlier-count  $<$  best-count:

best-count  $\leftarrow$  inlier-count

best- $M \leftarrow M_i$

best-data  $\leftarrow \{(x_i, y_i) : |M(x_i) - y_i| < \delta\}$  all data points in agreement w/  $M_i$

$M_{\text{final}} = \text{fit\_model}(\text{best-data})$

## Parameters

$K$  - # iterations (hypotheses)

$S$  - # data points needed to fit a model

$\delta$  - inlier threshold

## Choosing parameter values:

$\delta$ : based on expected inlier noise. Common case:

Assume Gaussian w/ variance  $\sigma^2$

Let  $\delta \approx 1.96 \cdot \sigma$

$S$ : based on specific problem:

- Linear regression - 2  $x, y$  points

- Translation fitting - one match  $(x, y) \leftrightarrow (x', y')$

- Affine - 3 matches

- Homography - 4 matches

- Ellipse (why not?) - 3  $(x, y)$  points

# Choosing Parameter Values (cont):

$K$  (# iterations) - suppose we want to find a set of  $s$  inliers with probability  $\geq P$

Assume we can estimate the inlier ratio

$$r = \frac{\# \text{ inliers}}{\# \text{ data points}}$$

In one hypothesis,

$$P(\text{choose all inliers}) = r^s$$

$$P(\text{at least one outlier}) = 1 - r^s$$

*Bad Thing*

Over  $K$  trials,

$$P(\text{at least one outlier all } K \text{ trials}) = (1 - r^s)^K$$

*Bad Thing happens K times in a row*

$$P = P(\text{no outliers in at least one trial}) = 1 - (1 - r^s)^K$$

*Bad Thing doesn't happen K times in a row*  
*Success*

What  $K$  do I need to make  $P(\text{success}) \geq P$ ?

$$P \geq 1 - (1 - r^s)^K$$

$$1 - P \geq (1 - r^s)^K$$

$$\log(1 - P) \geq K \log(1 - r^s)$$

$$\frac{\log(1 - P)}{\log(1 - r^s)} \geq K$$

Sanity checks:

Lower Prob. of success  $\rightarrow$  fewer iterations

More points to fit a model ( $s$ )  $\rightarrow$  more iterations