

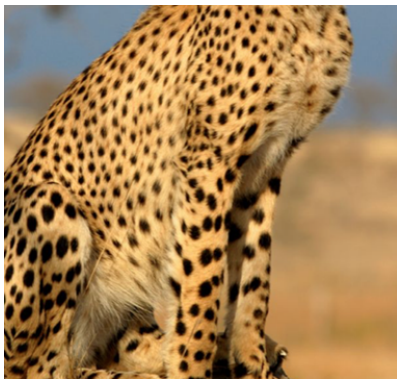
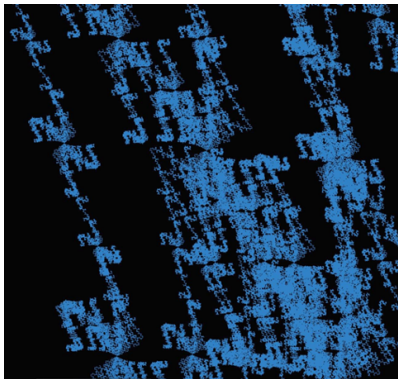


# Pattern recognition

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# What is a pattern?

Repetition, similarity



# What is a pattern?

Repetition, similarity



# Block matching

- Patch shift that minimizes difference
- Straightforward
- Non-invariant, unstable

$$R(x, y) = \sum_{x', y'} (T(x', y') - I(x + x', y + y'))^2$$

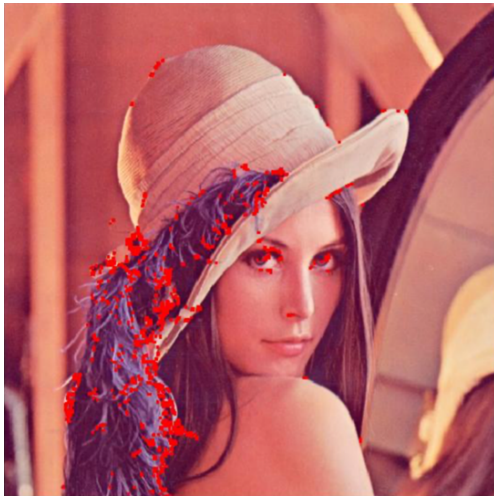
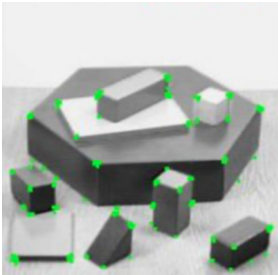
# Block matching

- Patch shift that minimizes difference
- Straightforward
- Non-invariant, unstable



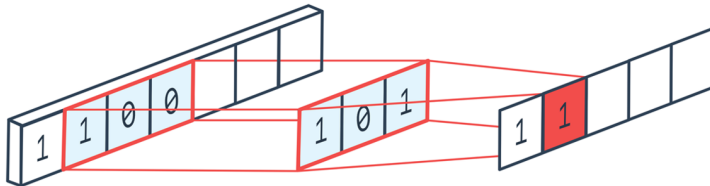
# Key points

- Corners, distinct elements
- Much less in number than pixels
- Easily trackable



# Convolution

The sum of the element products for all kernel positions



# Smoothing filter circuit

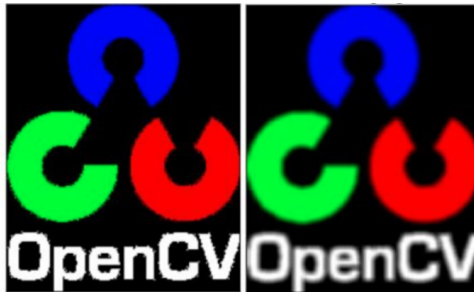
- Kernel of unit elements
- Automatic normalization
- The element is equal to the average of neighbors

$$K = \frac{1}{25} \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$



# Smoothing filter circuit

- Kernel of unit elements
- Automatic normalization
- The element is equal to the average of neighbors

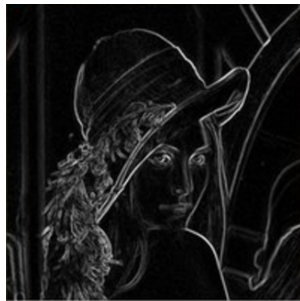


# Border highlighting filters

- Discrete approximation of the derivative
- Equal to zero on a constant signal

$$G_x = \begin{bmatrix} -3 & 0 & +3 \\ -10 & 0 & +10 \\ -3 & 0 & +3 \end{bmatrix}$$

$$G_y = \begin{bmatrix} -3 & -10 & -3 \\ 0 & 0 & 0 \\ +3 & +10 & +3 \end{bmatrix}$$



# Least squares method

$$y = ax + b$$

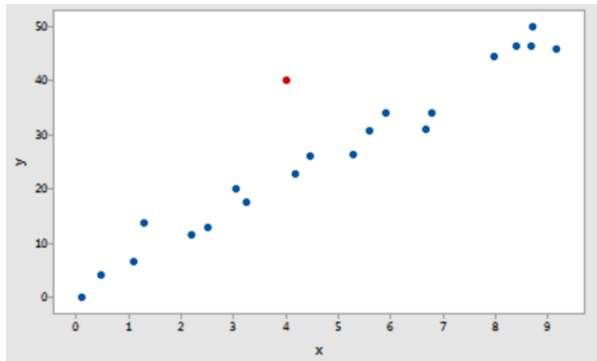
$$e_i = y_i - ax_i - b$$

$$E = \sum_{i=1}^n (y_i - ax_i - b)^2$$

$$\begin{cases} \frac{\partial E}{\partial a} = 2 \sum_{i=1}^n (ax_i + b - y_i)x_i = 0 \\ \frac{\partial E}{\partial b} = 2 \sum_{i=1}^n (ax_i + b - y_i) = 0 \end{cases}$$

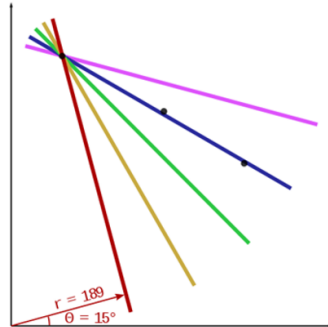
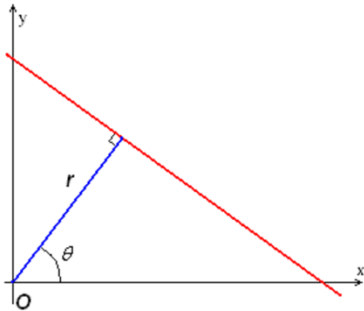
$$\hat{b} = \frac{\sum_{i=1}^n y_i \sum_{i=1}^n x_i^2 - \sum_{i=1}^n y_i x_i \sum_{i=1}^n x_i}{n \sum_{i=1}^n x_i^2 - \left( \sum_{i=1}^n x_i \right)^2}$$

$$\hat{a} = \frac{\sum_{i=1}^n y_i x_i - \hat{b} \sum_{i=1}^n x_i}{\sum_{i=1}^n x_i^2}$$



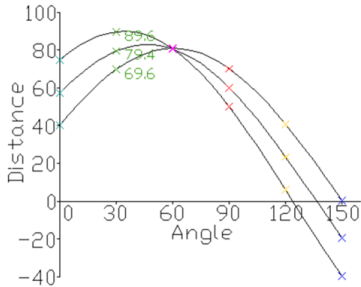
# Hough transform

- Parameterization via angle and distance
- Finite (discrete) step

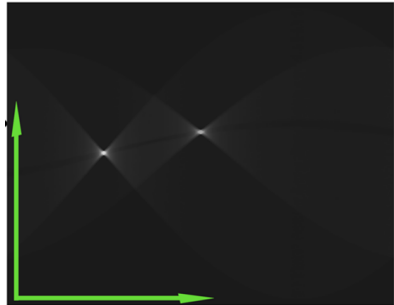


# Hough transform

- The data point is mapped to a curve
- Result - local/global maxima

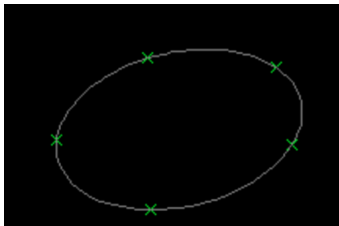


Rendering of Transform Results



# RANSAC

- Random subsample
- Model obtainment
- Quality assessment



# RANSAC

- Searching for a model with a maximum of inliers
- Works correctly with desired probability

$P$  - expected success probability

$\alpha$  - share of the inliers

$n$  - number of data points

$k$  - number of points that determine the model

$\alpha$

$\alpha^k$

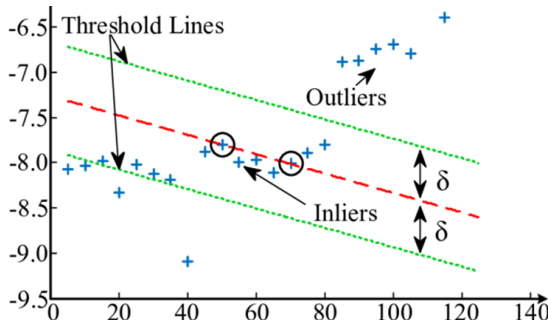
$1 - \alpha^k$

$$(1 - \alpha^k)^m = 1 - P$$

$$\ln(1 - \alpha^k)^m = \ln(1 - P)$$

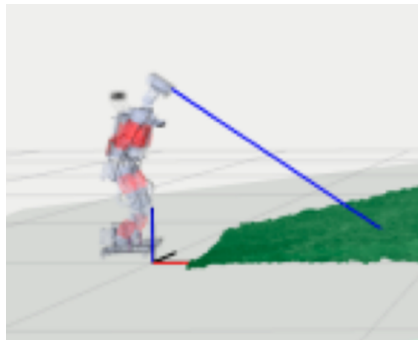
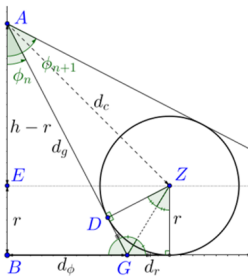
$$m \ln(1 - \alpha^k) = \ln(1 - P)$$

$$m = \frac{\ln 1 - P}{\ln(1 - \alpha^k)}$$



# TinyYOLOv3 object detection

- Neural networks work well...
- ... but don't get the depth on mono





# TinyYOLOv3 object detection

- 1.3 times longer
- Considerably better (mAP@0.9 0.503 → 0.599)





# Thank you for your time

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