

Проективная геометрия в компьютерном зрении

Ерухимов Виктор Львович (*Itseez*), Илья Лысенков (*Itseez*)



Kurt Wenner, Disaster

http://kurtwenner.com/galleries/pavement/pavement_3/pages/StreetPaintingGallery3.018.htm

Copyrighted Material

SECOND EDITION

Multiple View Geometry in computer vision



Richard Hartley and Andrew Zisserman

Copyrighted Material

CAMBRIDGE

<http://www.robots.ox.ac.uk/~vgg/hzbook/>

Зачем нужна геометрия в компьютерном зрении?



©2001 How Stuff Works

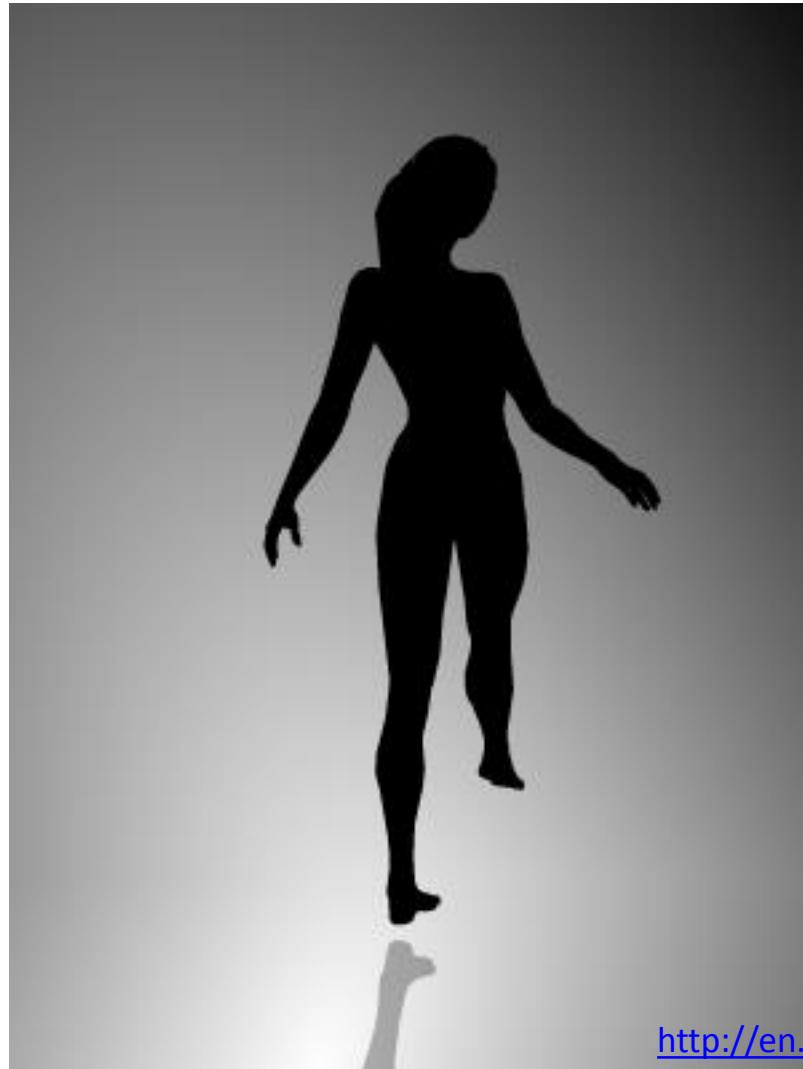


Willow Garage, <http://www.youtube.com/watch?v=gYqfa-YtvW4>
Garfield, <http://www.imdb.com/media/rm1129879808/tt0356634>

Volvo, <http://bit.ly/162KtRZ>

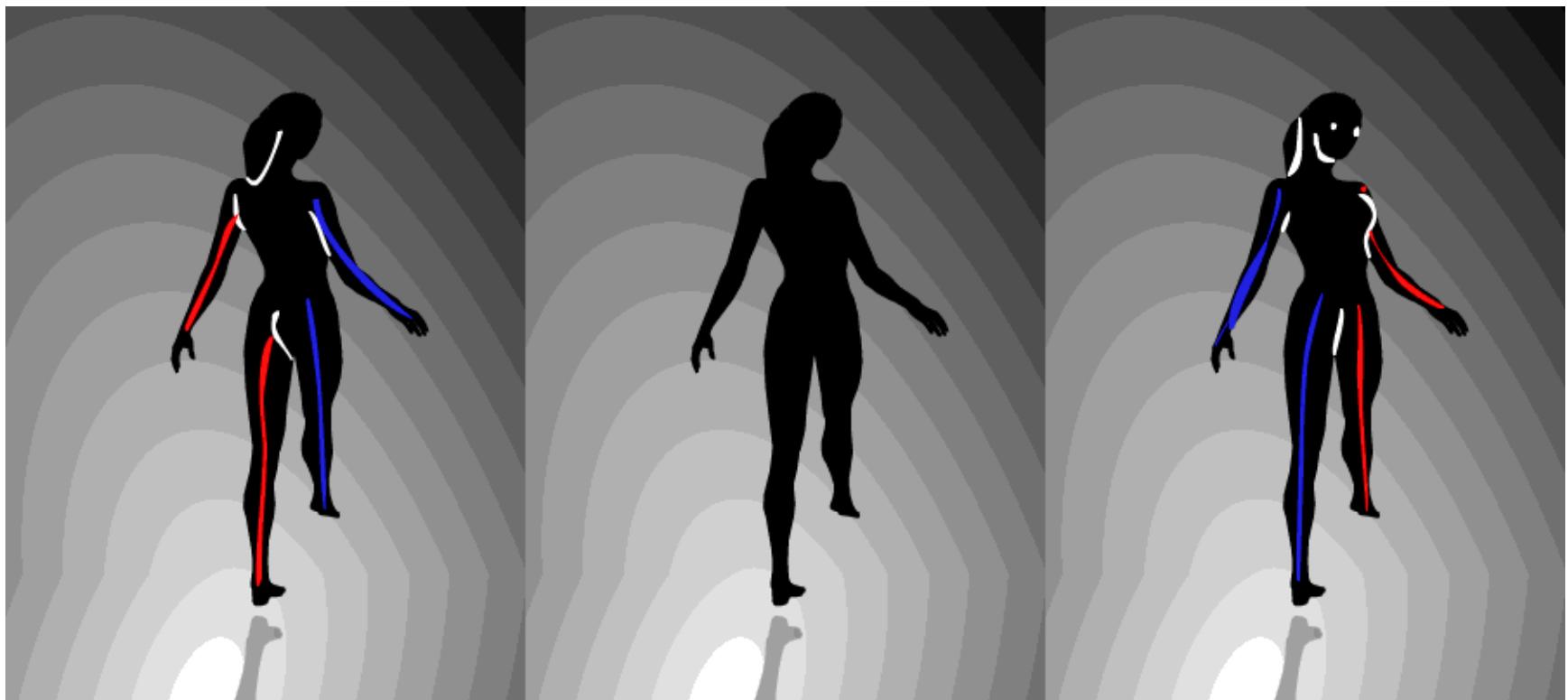
How stuff works, <http://computer.howstuffworks.com>

В какую сторону вращается танцовщица?



http://en.wikipedia.org/wiki/Spinning_Dancer

В какую сторону вращается танцовщица?





<http://mathworld.wolfram.com/YoungGirl-OldWomanIllusion.html>

Pinhole camera model

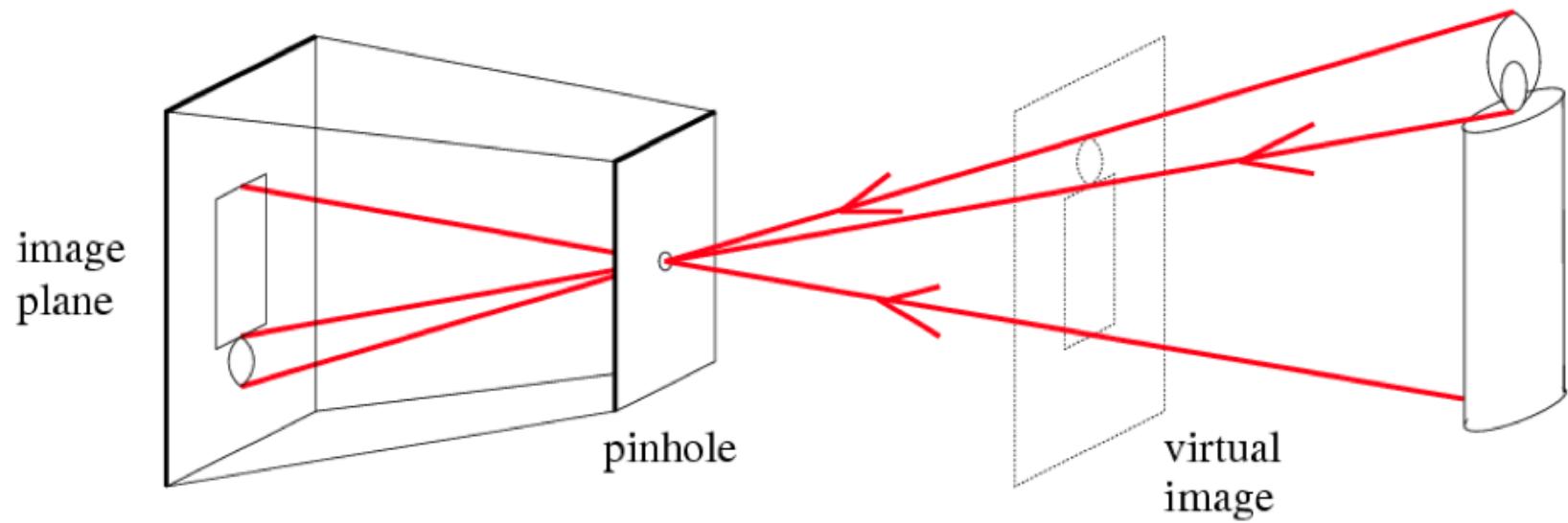
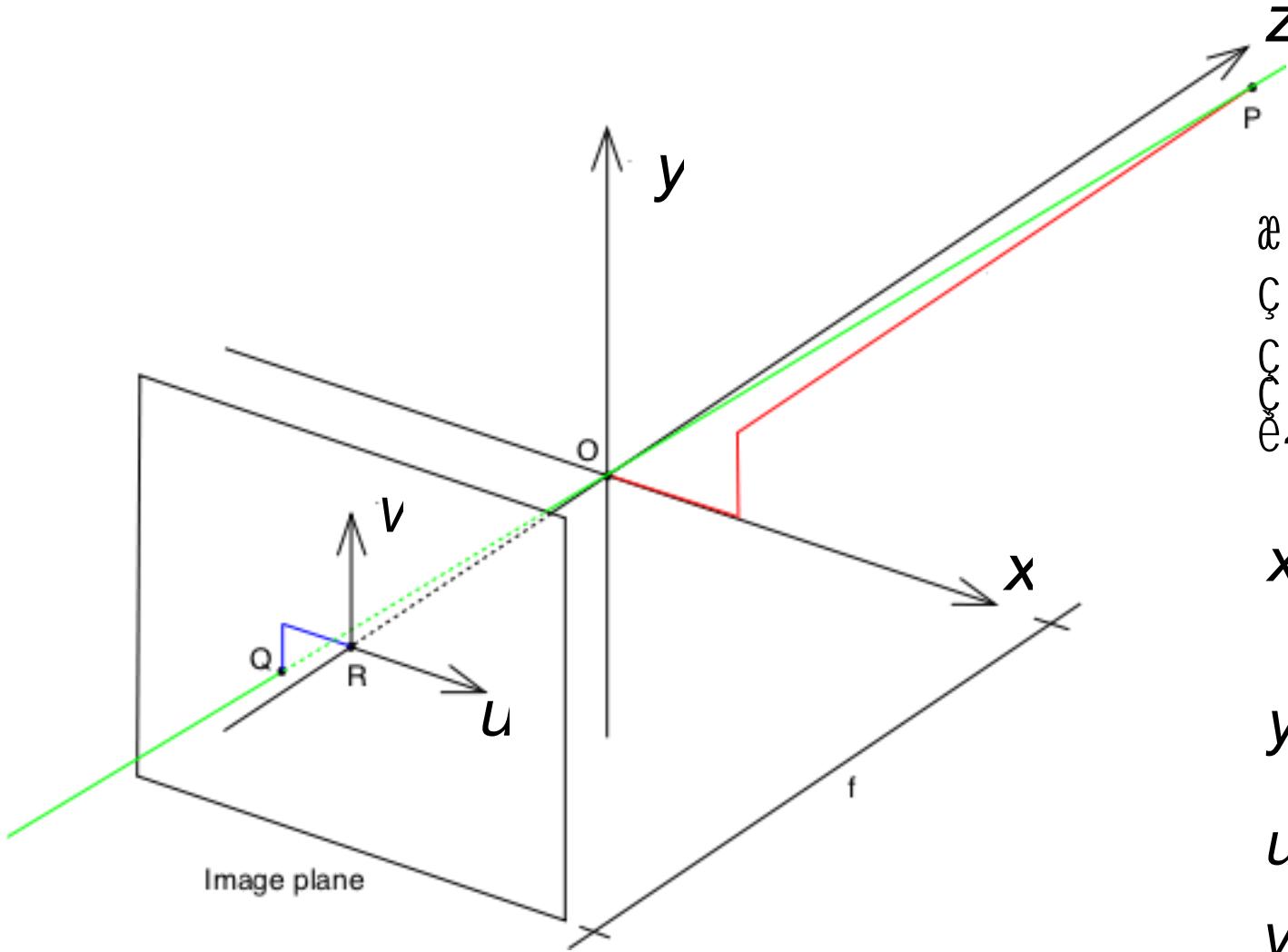


Image from J. Sivic's presentation,

http://www.ens-lyon.fr/LIP/Arenaire/ERVision/camera_geometry_alignment_final.pdf

Pinhole camera model



$$\begin{matrix} \alpha X^0 \\ C Y^0 \\ C Z^0 \end{matrix} = \begin{matrix} \alpha X_0^0 \\ C Y_0^0 \\ C Z_0^0 \end{matrix} + t$$

$$x' = \frac{x}{z}$$

$$y' = \frac{y}{z}$$

$$u = f_x x' + c_x$$

$$v = f_y y' + c_y$$

Camera projection matrix

$$\begin{matrix} \mathbf{x} & \mathbf{y} & \mathbf{z} \\ \mathbf{w} & \mathbf{v} & \mathbf{u} \\ \mathbf{c} & \mathbf{d} & \mathbf{e} \\ 1 & 1 & 1 \end{matrix} = P \begin{matrix} \mathbf{X}^0 \\ \mathbf{Y}^0 \\ \mathbf{Z}^0 \\ 0 \end{matrix}$$

$$P = K[R|T]$$

$$K = \begin{matrix} f_x & 0 & c_x^0 \\ 0 & f_y & c_y^0 \\ 0 & 0 & 1 \end{matrix}$$

Camera distortion



http://commons.wikimedia.org/wiki/File:Fisheye_lens_room_1.jpg

Distortion model

$$x'' = x'(1 + k_1 r^2 + k_2 r^4 + k_3 r^6) + 2p_1 x' y' + p_2(r^2 + 2x'^2)$$

$$y'' = y'(1 + k_1 r^2 + k_2 r^4 + k_3 r^6) + p_1(r^2 + 2y'^2) + 2p_2 x' y'$$

$$\text{where } r^2 = x'^2 + y'^2$$

$$u = f_x * x'' + c_x$$

$$v = f_y * y'' + c_y$$

Pose estimation

Detected image
points:

$$\begin{matrix} \text{í æ } U_i \\ \text{í ç } v_i \\ \text{í e } p_i \end{matrix}_{i=1..n}$$

Train object
points:

$$\begin{matrix} \text{í æ } X_i \\ \text{í ç } Y_i \\ \text{í ç } Z_i \end{matrix}_{i=1..n}$$

A class of object
transformations:

$$f \begin{matrix} \text{é æ } X^0 \\ \text{é ç } Y^0 \\ \text{é ç } Z^0 \end{matrix}, R, T = R \begin{matrix} \text{é ç } Y^0 \\ \text{é ç } Z^0 \end{matrix} + T$$

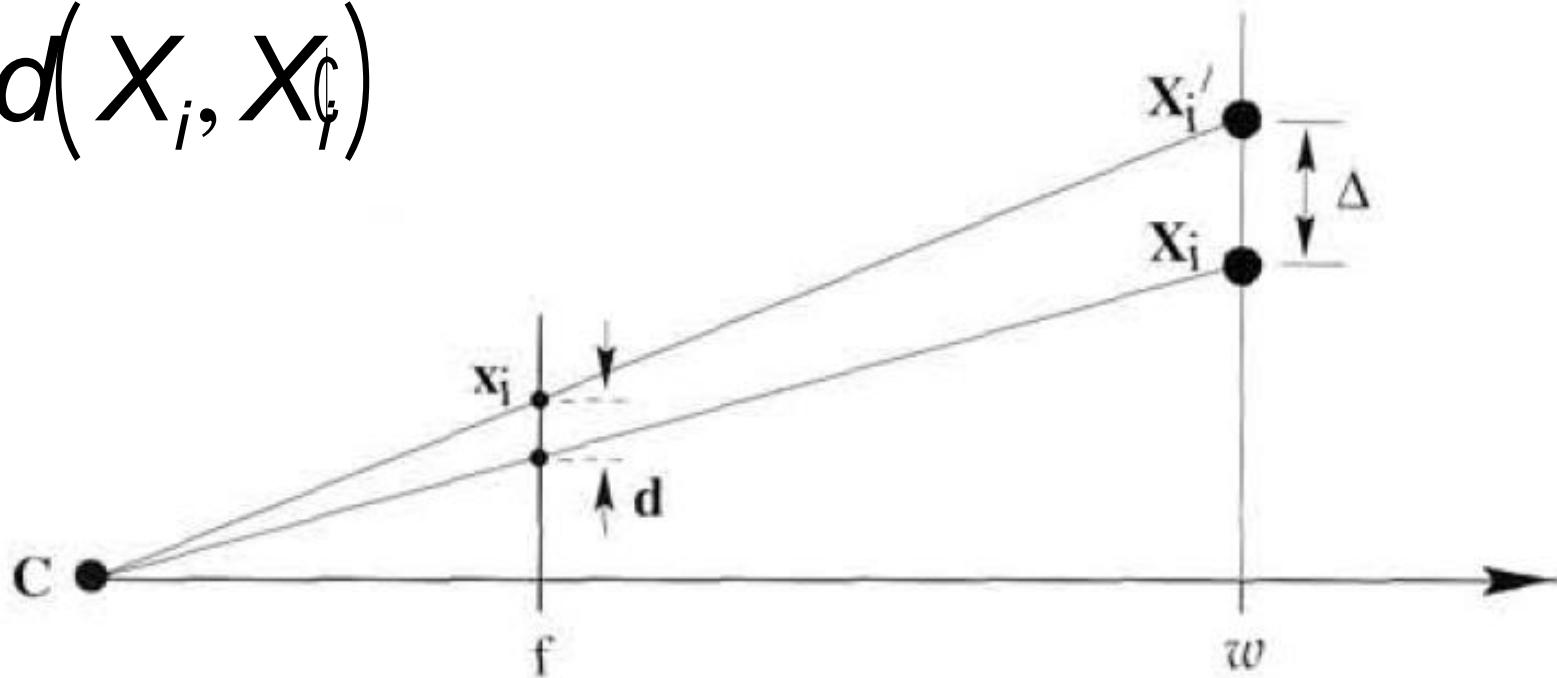
Reprojection error

$$\begin{array}{c} \hat{u}_i^p \\ \hat{v}_i^p \\ \hat{w}_i \end{array} = P \begin{array}{c} \hat{X}_i \\ \hat{Y}_i \\ \hat{Z}_i \end{array}$$

$$error(P) = \sum_i \frac{\|u_i^p - \hat{u}_i^p\|^2}{\|v_i^p\|}$$

3D geometric error

$$d(X_i, X_{\text{c}})$$

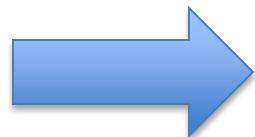


Perspective-n-Points problem

$$\min_{R, T} \text{error}(K, R, T)$$

Direct Linear Transformation

$$\begin{array}{c} \mathfrak{X} u_i^p w_i \\ \zeta v_i^p w_i \\ \zeta e w_i \end{array} \div \begin{array}{c} \mathfrak{X}_i \\ \zeta Y_i \\ \zeta Z_i \\ \zeta 1 \end{array} \div \emptyset = P \begin{array}{c} \mathfrak{X}_i \\ \zeta Y_i \\ \zeta Z_i \\ \zeta 1 \end{array} \div \begin{array}{c} \mathfrak{X}_i \\ \zeta Y_i \\ \zeta Z_i \\ \zeta 1 \end{array} \div \emptyset$$



$$\begin{array}{c} \mathfrak{X} u_i^p \\ \zeta v_i^p \\ \zeta 1 \end{array} \div \begin{array}{c} \mathfrak{X}_i \\ \zeta Y_i \\ \zeta Z_i \\ \zeta 1 \end{array} \div \emptyset - P \begin{array}{c} \mathfrak{X}_i \\ \zeta Y_i \\ \zeta Z_i \\ \zeta 1 \end{array} \div \begin{array}{c} \mathfrak{X}_i \\ \zeta Y_i \\ \zeta Z_i \\ \zeta 1 \end{array} \div \emptyset = 0$$

Other methods

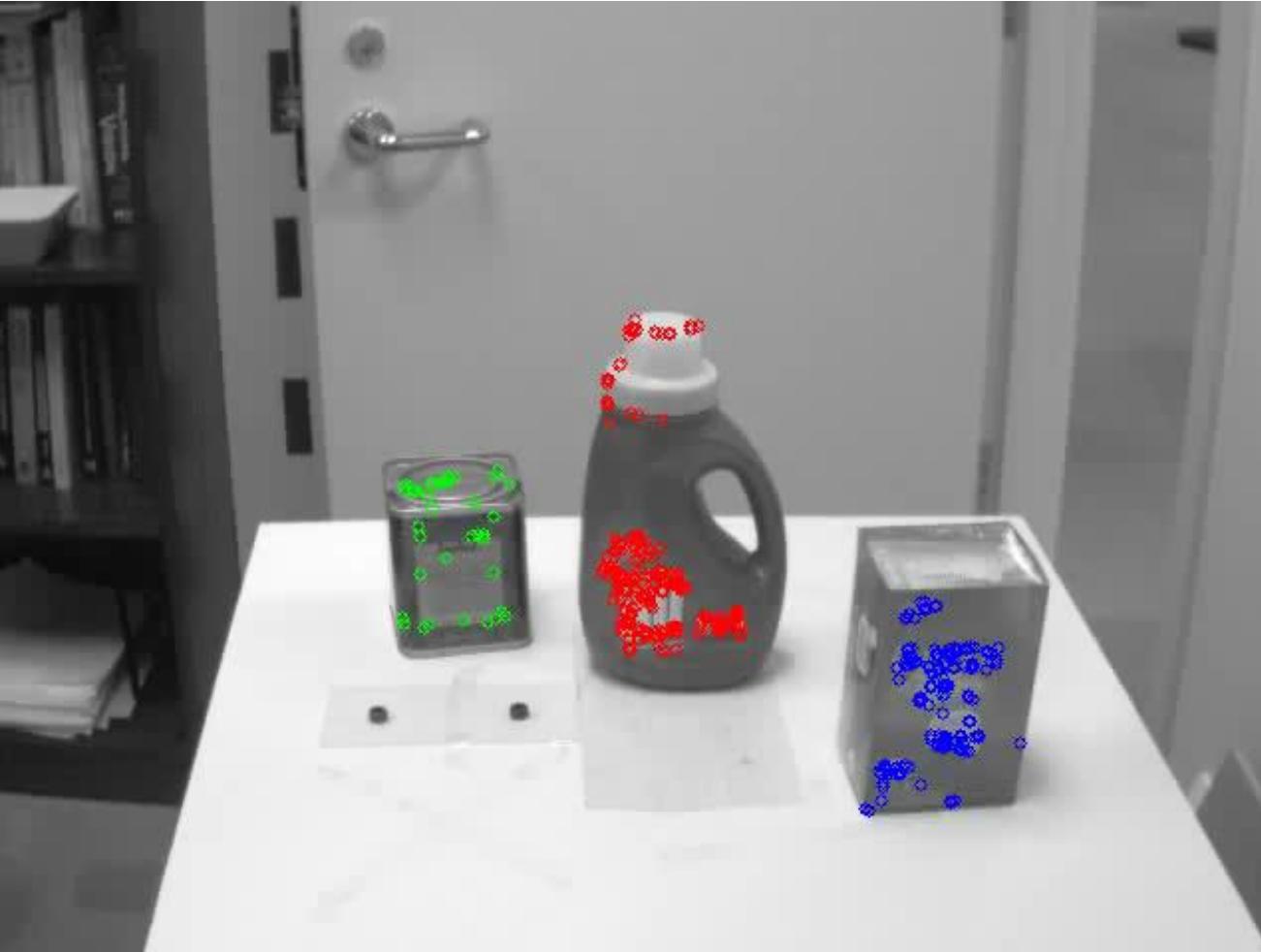
- $O(N)$ closed-form methods
- Levenberg-Marquardt
- P3P, P4P
- RANSAC

Random Sample Consensus

- Do n iterations until #inliers > inlierThreshold
 - Draw k matches randomly
 - Find the transformation
 - Calculate inliers count
 - Remember the best solution

The number of iterations required $\sim \frac{\# \text{ matches}^k}{\# \text{ inliers}^\theta}$

Object detection example

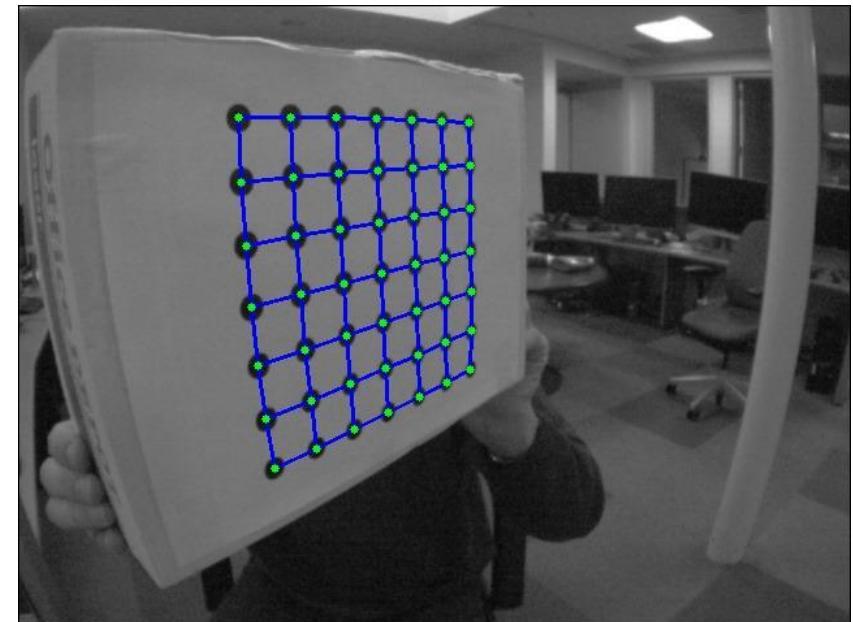


Iryna Gordon and David G. Lowe,
"What and where: 3D object
recognition with accurate pose," in
*Toward Category-Level Object
Recognition*, eds. J. Ponce, M.
Hebert, C. Schmid, and A.
Zisserman, (Springer-Verlag, 2006),
pp. 67-82.

Manuel Martinez Torres, Alvaro
Collet Romea, and Siddhartha
Srinivasa, MOPED: A Scalable and
Low Latency Object Recognition
and Pose Estimation
System, Proceedings of ICRA 2010,
May, 2010.

Camera calibration

- Estimate camera parameters given images of a known template



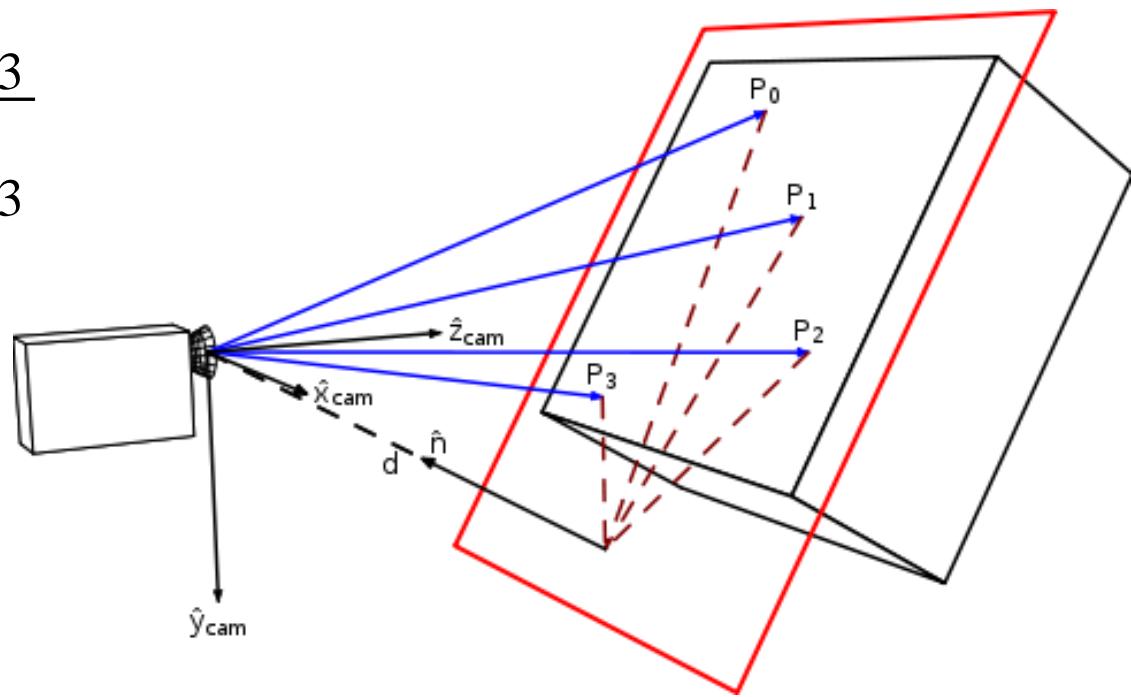
OpenCV test data

Homography

$$\tilde{u} = \frac{h_{11}u + h_{12}v + h_{13}}{h_{31}u + h_{32}v + h_{33}}$$

$$\tilde{v} = \frac{h_{21}u + h_{22}v + h_{23}}{h_{31}u + h_{32}v + h_{33}}$$

$$\begin{matrix} \text{æ} & \tilde{w} \\ \zeta & \tilde{w} \\ \zeta & w \end{matrix} \begin{matrix} \text{æ} & \tilde{w} \\ \zeta & \tilde{w} \\ \zeta & w \end{matrix} \begin{matrix} \text{æ} & \tilde{w} \\ \zeta & \tilde{w} \\ \zeta & w \end{matrix}$$



Нахождение гомографии

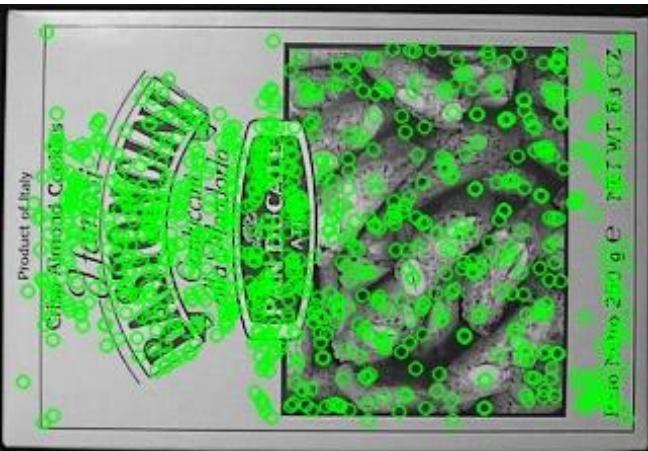
- Direct Linear Transformation

$$\begin{array}{c} \tilde{U}_i \\ \vdots \\ W_i \\ \tilde{V}_i \\ \vdots \\ \tilde{e}_1 \\ \emptyset \end{array} = H \begin{array}{c} U_i \\ \vdots \\ C \\ V_i \\ \vdots \\ e_1 \\ \emptyset \end{array}$$



$$\begin{array}{c} \tilde{U}_i \\ \vdots \\ \tilde{V}_i \\ \vdots \\ \tilde{e}_1 \\ \emptyset \end{array} - H \begin{array}{c} U_i \\ \vdots \\ C \\ V_i \\ \vdots \\ e_1 \\ \emptyset \end{array} = 0$$

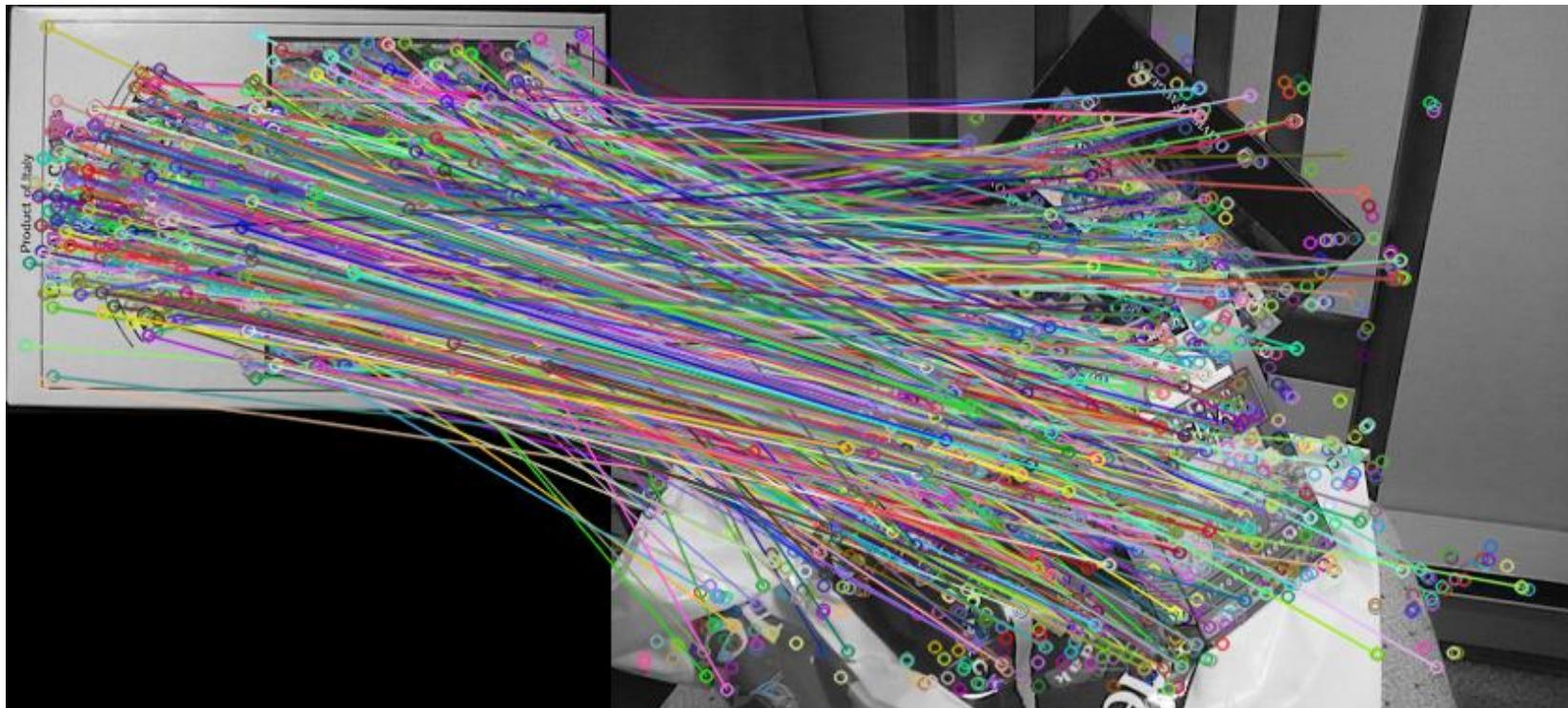
Keypoints example



OpenCV features2d tutorial,

http://docs.opencv.org/doc/tutorials/features2d/table_of_content_features2d/table_of_content_features2d.html

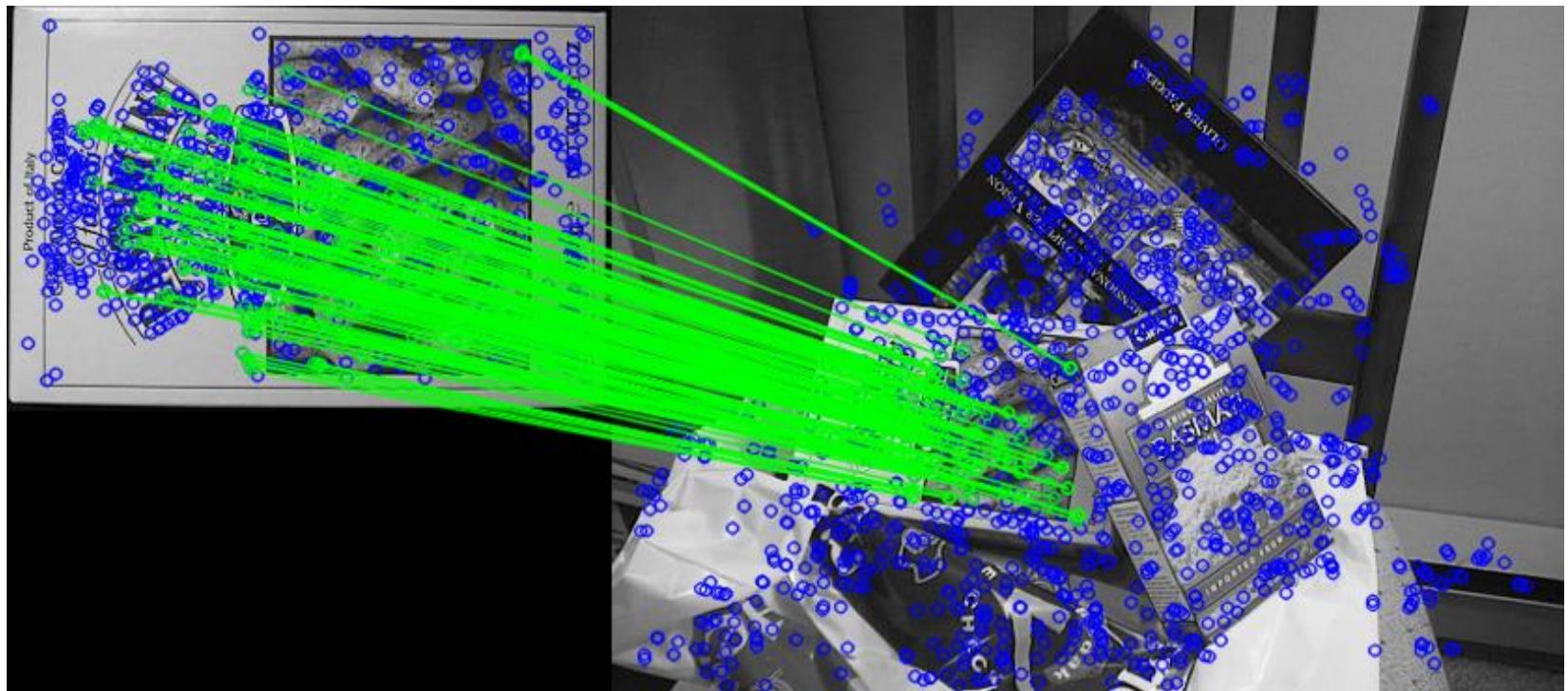
Matching descriptors example



OpenCV features2d tutorial,

http://docs.opencv.org/doc/tutorials/features2d/table_of_content_features2d/table_of_content_features2d.html

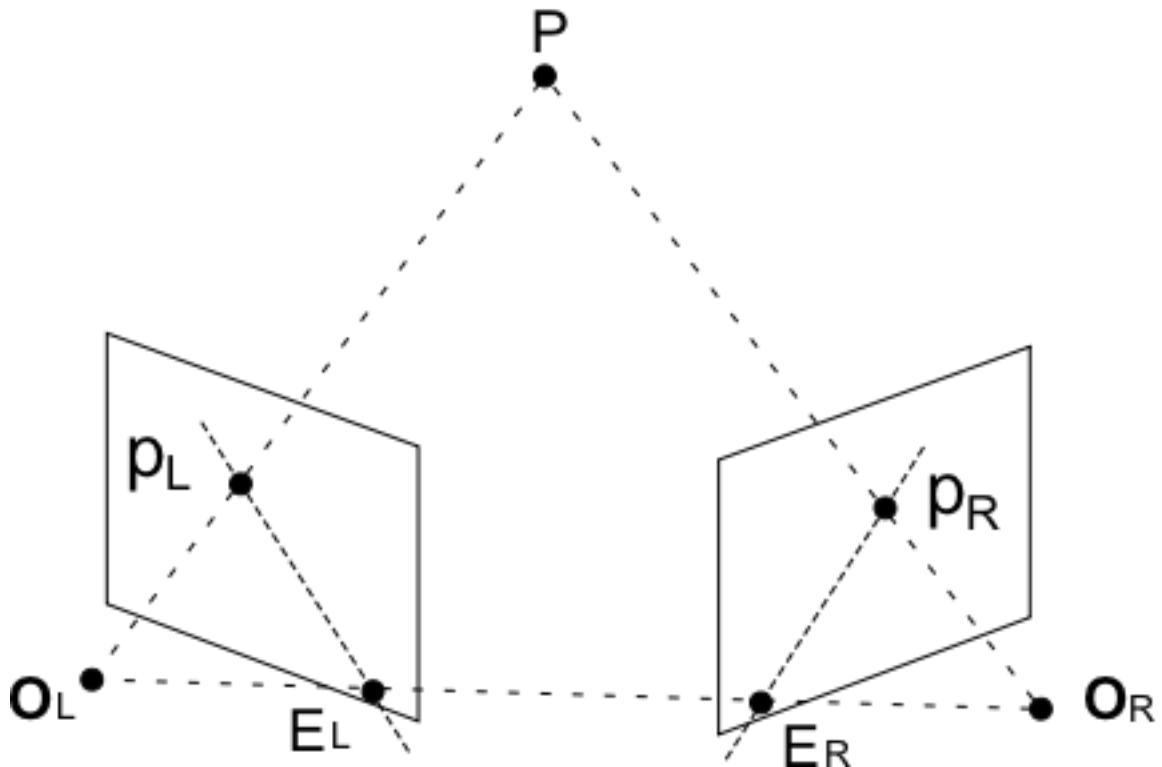
Geometry validation



OpenCV features2d tutorial,

http://docs.opencv.org/doc/tutorials/features2d/table_of_content_features2d/table_of_content_features2d.html

Stereo: epipolar geometry



Fundamental/essential matrix constraint

$$(u_1, v_1, 1) \times F \times \begin{matrix} u_2 \\ v_2 \\ 1 \end{matrix} = 0$$

$$(x_1, y_1, 1) \times E \times \begin{matrix} x_2 \\ y_2 \\ 1 \end{matrix} = 0$$

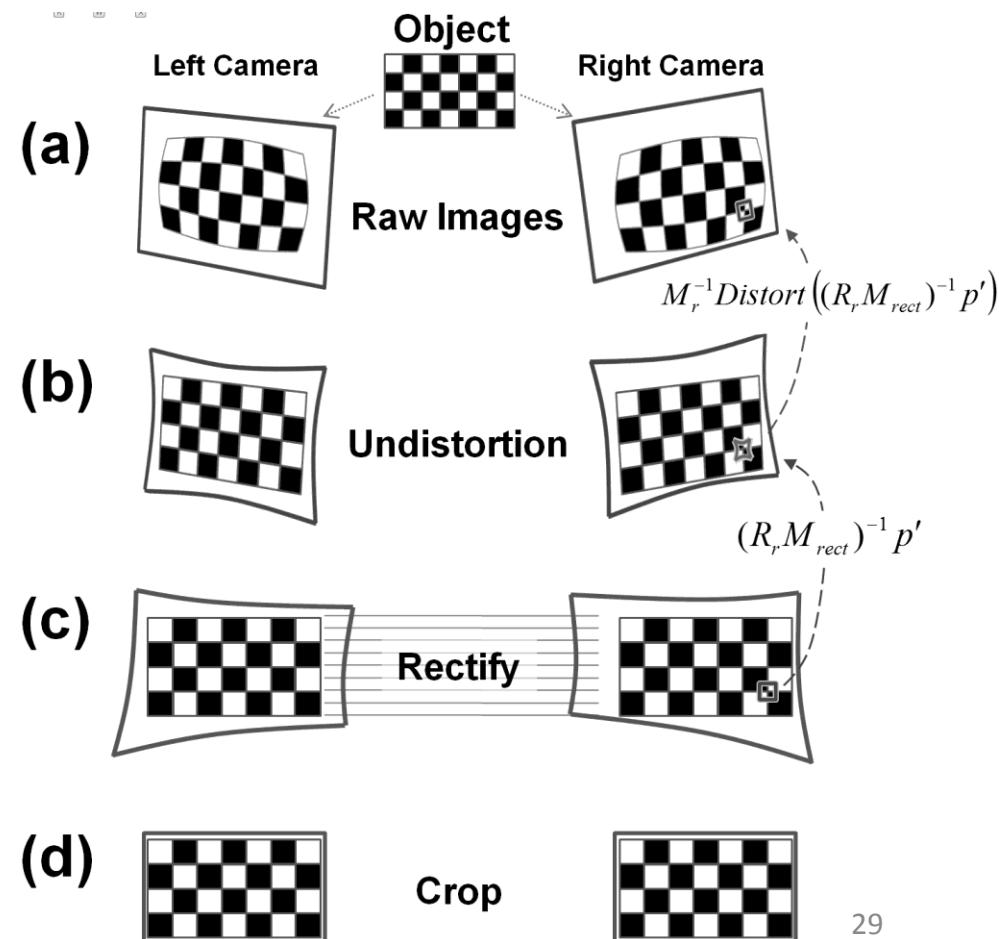
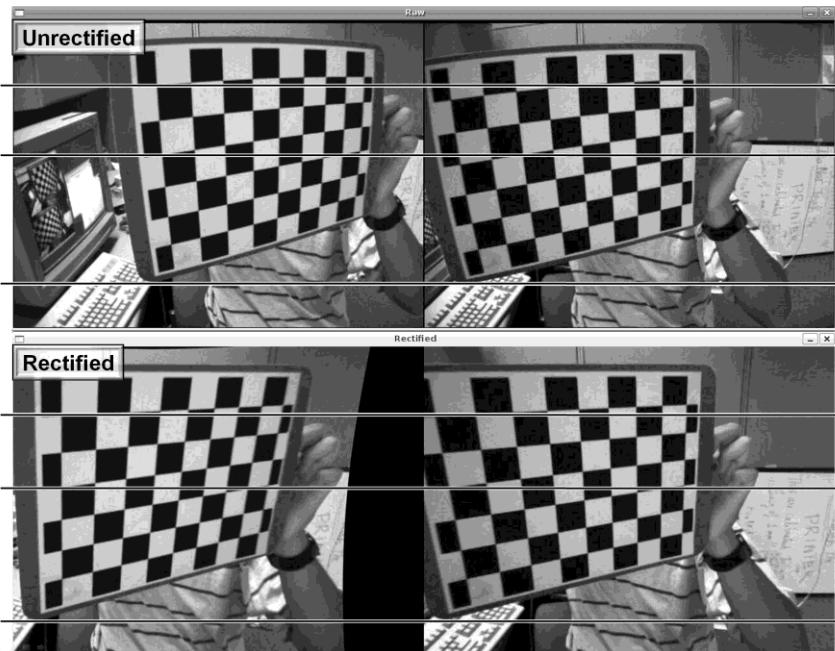
Нахождение фундаментальной матрицы

- По 8 соответствиям
 - Опциональная нормализация для устойчивости
 - Решение системы уравнений $q_i^T F q_i = 0$
 - Замена F на ближайшую сингулярную
- По 7 соответствиям
 - Общее решение имеет вид $F = aF_1 + (1-a)F_2$
 - Используем условие $\det(F) = 0$ и получаем уравнение 3-й степени относительно F

The Fundamental Matrix Song

Stereo Rectification

- Algorithm steps are shown at right:
- Goal:
 - Each row of the image contains the same world points
 - “Epipolar constraint”

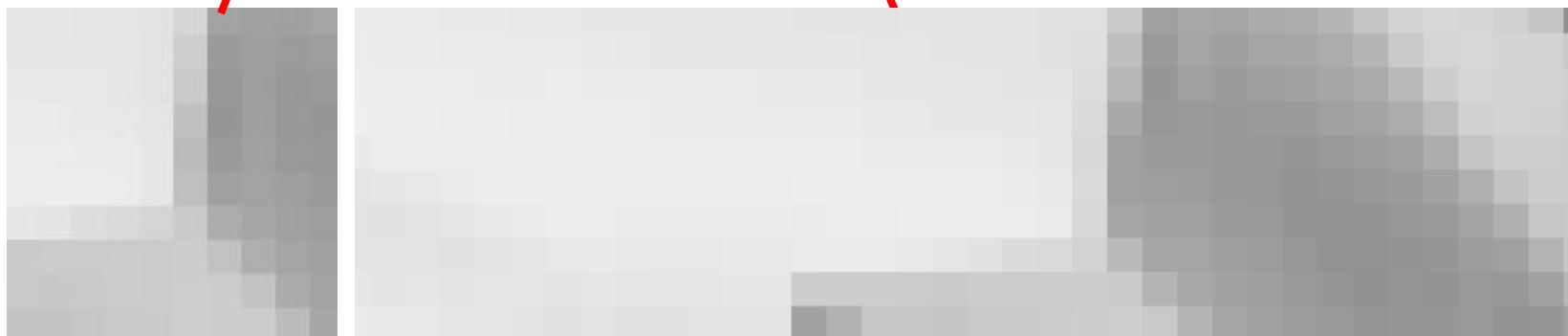


Stereo correspondence block matching

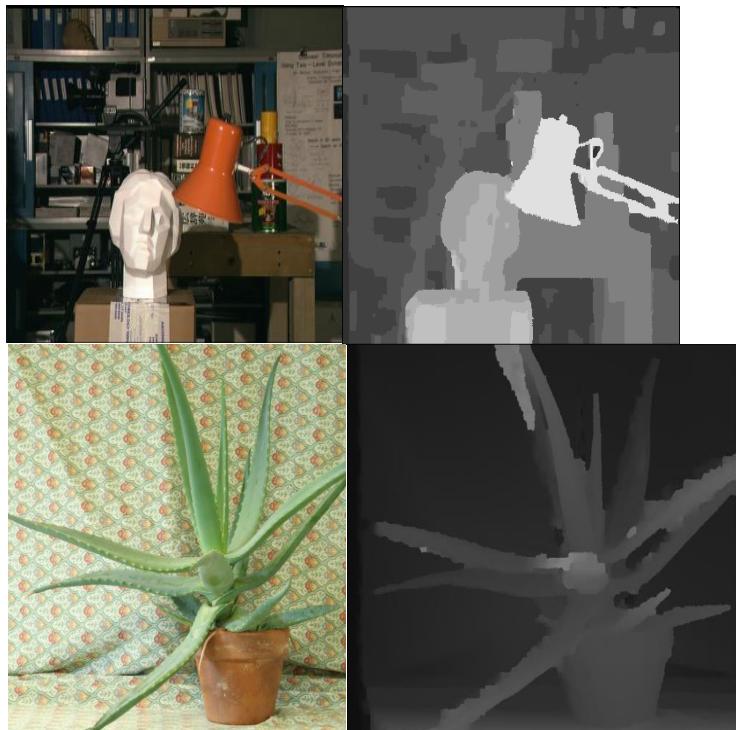
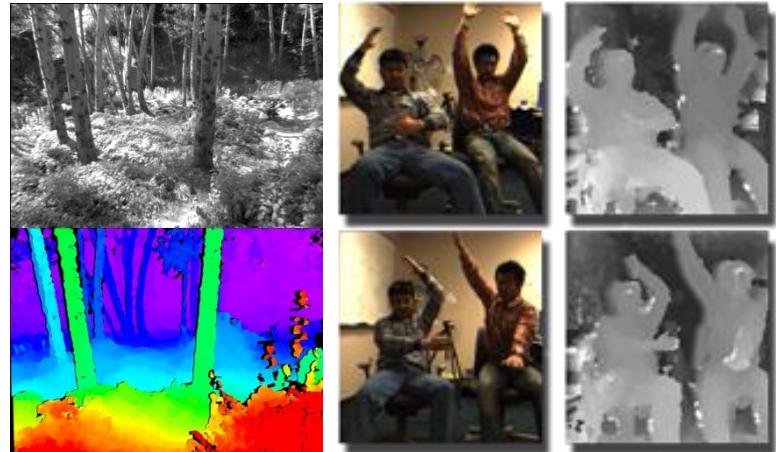


For each block in left image:

Search for the corresponding block in the right image such that SSD or SAD between pixel intensities is minimum



Stereo Matching



$$Z = f_x \frac{T}{D}$$

<http://vision.middlebury.edu/stereo/data/>

OpenCV functions: single view

- SVD
- projectPoints
- undistort
- solvePnP
- calibrateCamera
- findChessboardCorners/findCirclesGrid

OpenCV functions: two views

- findHomography
- findFundamentalMat
- stereoCalibrate
- stereoRectify/reprojectImageTo3D/initUndistortRectifyMap