# **StereoVision**

**Paulo Dias** 





## **Sumary**



- Introduction Stereopsis
- Camera models and Camera calibration
- Stereo Vision
  - Frontal parallel arrangement
  - Epipolar geometry
  - Essential and Fundamental Matrix
  - Image rectification
  - Template Matching
  - Stereo Matching

## Introduction



What is stereo?

Where is it coming from?

Where can you see it or use it?

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## **Human Visual System**



- Many of the perceptual cues we use to visualize 3D structure are available in 2D projections
- Such cues include:
  - occlusion (one object partially covering another)
  - perspective (point of view)
  - familiar size (we know the real-world sizes of many objects)
  - atmospheric haze (objects further away look more washed out)
- Four cues are missing from single 2D views:
  - stereo parallax seeing a different image with each eye
  - movement parallax -seeing different images when we move the head
  - Accommodation the eyes' lenses focus on the object of interest
  - Convergence both eyes converge on the object of interest

## **Stereopsis**

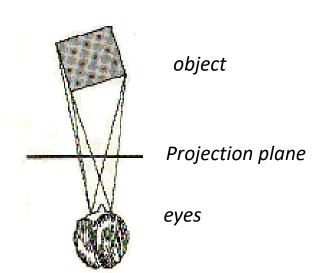


Stereo ="solid" or "three-dimensional" opsis = appearance or sight

Also: "binocular vision",

"binocular depth perception"

"stereoscopic depth perception"



- Stereopsis is the impression of depth that is perceived when a scene is viewed with both eyes by someone with normal binocular vision
- Binocular disparity is due to the different position of our two eyes

## **Stereopsis**



- Depth perception in stereo is based on stereopsis:
  - when the brain registers and fuses two images
  - Image parallax means that the two eyes register different images (horizontal shift)
  - The amount of shift depends on the "interpupillary distance" (IPD) (varies for each person in the range of 53-73 mm)
  - Works in the near field (to a few meters from the eye)

## **Stereopsis**







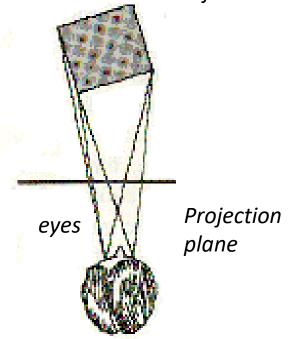
(Hearn and Baker, 2002)

object





Right eye image Left eye image



## **Stereopsis: implications for Graphic devices**



- Need to present two images of the same scene
- The two images can be presented:
  - at the same time on two displays (HMD)
  - time-sequenced on one display (active glasses)
  - spatially-sequenced on one display (auto-stereoscopic displays)



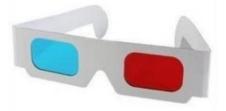
Left eye, right eye images (Burdea and Coiffet., 2003)



## Common ways to produce 3D sensation



Anaglyphs: two colored images and color coded glasses (red/cyan(green))



- Two images with different light polarization and polarizing glasses
  - Linear and circular
- Double frame-rate displays combined with LCD shutter glasses
- Autostereoscopic displays
  - Parallax barrier and lenticular lens
- Head Mounted Displays (HMDs)
- and "exotic displays"



# **Sumary**

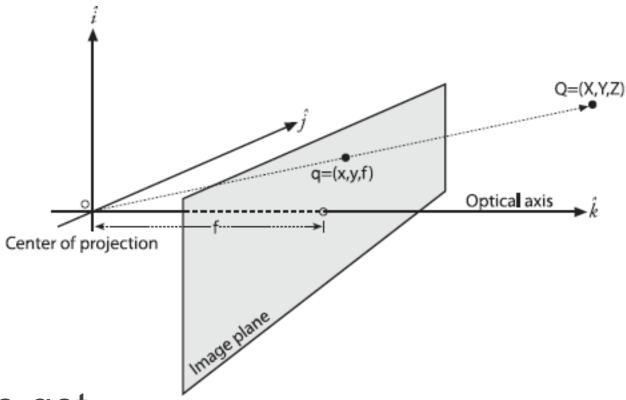


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### Camera model



pinhole camera model



We get

$$x = f \frac{X}{Z}$$

$$y = f \frac{Y}{Z}$$

# **OpenCV Camera model**



- OpenCV Camera model:
  - 4 intrinsic parameters:
    - Focal distance:  $f_x$ ,  $f_y$
    - Optical centre:  $c_x$ ,  $c_y$
  - 5 distortion parameters
    - Lens distortion:  $k_1$ ,  $k_2$ ,  $k_3$ ,  $p_1$ ,  $p_2$
  - 6 extrinsic parameters:
    - Rotation:  $r_x$ ,  $r_y$ ,  $r_z$
    - Translation:  $t_x$ ,  $t_y$ ,  $t_z$

Total: 15 parameters

Other models: Tsai, Heikkila, Zhang

# **Sumary**

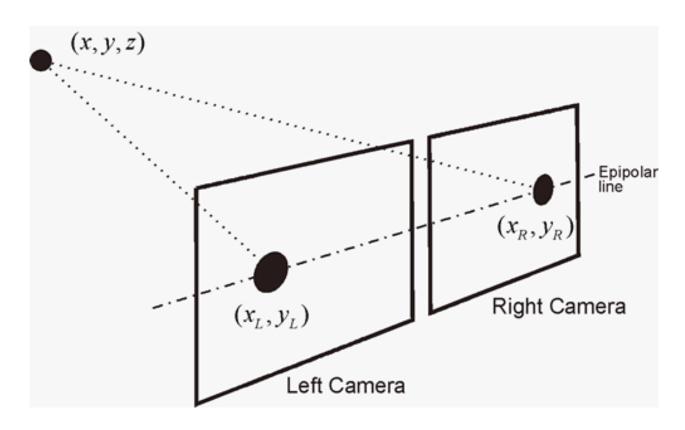


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### **StereoVision**

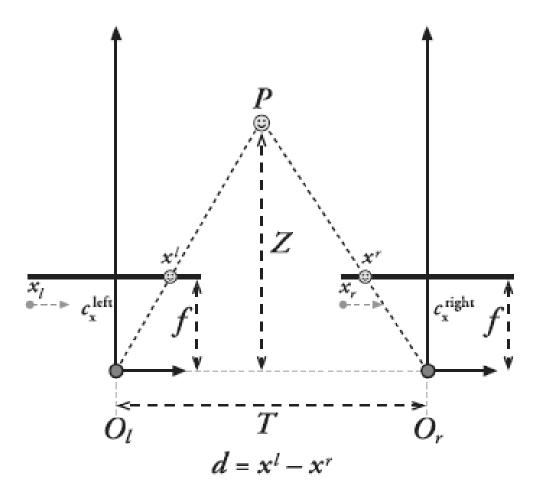


- Capability to define depth from 2 images
- Possible by computing correspondences between two images

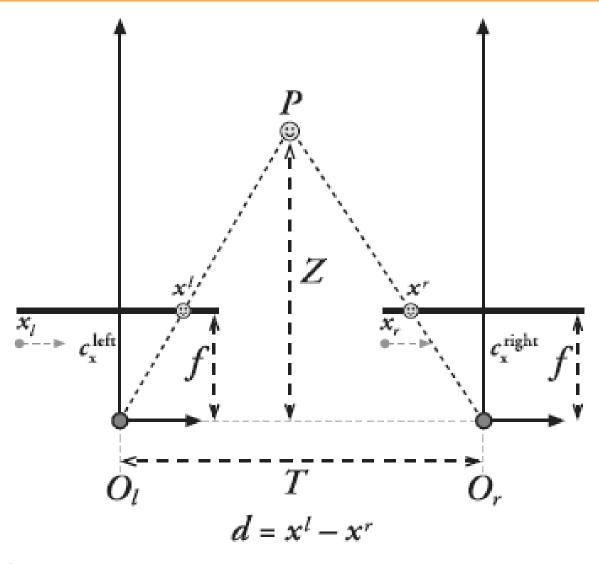




 Two perfectly aligned, coplanar cameras with same focal distance:









• 
$$\frac{T}{Z} = \frac{T - (x^l - x^r)}{(Z - f)}$$
 then  $Z = \frac{T(Z - f)}{T - (x^l - x^r)}$ ,

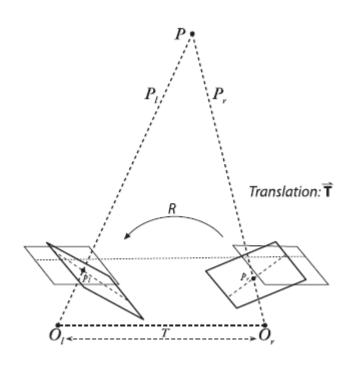
• 
$$TZ - Z(x^l - x^r) = TZ - fT$$

• So: 
$$Z = \frac{fT}{d}$$

=>Stereo system have good depth resolution for close objects since depth is inversely proportional to disparity.



- Easy to relate correspondence to depth in frontal parallel arrangement
- Problem: how to map real configuration to frontal parallel arrangement.



## **Sumary**

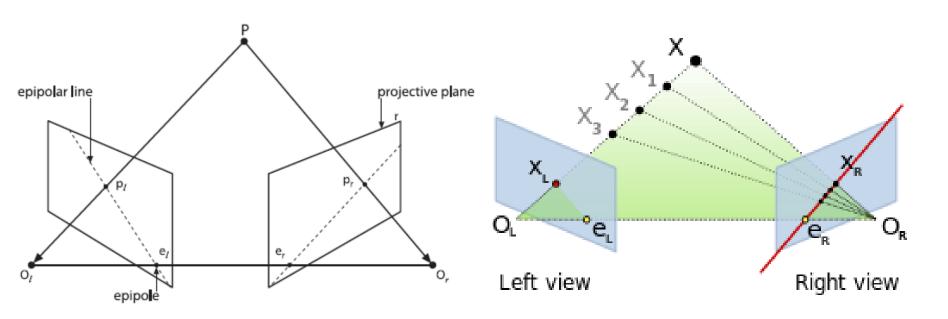


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## **Epipolar geometry**



 An epipole is a projection of the optical centre of a camera on the other image plane



http://www.ai.sri.com/~luong/research/Meta3DViewer/EpipolarGeo.html

# **Epipolar geometry**



- What is it useful for?:
  - Given a point in an image, its corresponding point in the other image lies on the corresponding epipolar line
  - Order is preserved (given 2 points A e B in a given order in one images, order will be the same in the other image)

=>Epipolar geometry transform a 2D search (in image) into a 1D search (along epipolar lines) saving resources and avoiding errors.

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### **Essential and Fundamental Matrices**



- Epipolar Geometry is defined by:
  - Information about relative position between the cameras (rotation and translation)
     [extrinsic] – Essential Matrix(E)
  - Intrinsic parameters of the cameras (focal length, lens distortion, optical centre, etc...) – Fundamental Matrix(F)

### **Essential matrix**



- Matrix that maps a 3D point in one image with its corresponding 3D point on the other image considering translation and rotation between cameras:
- $p_l^T E p_r = 0$
- $p_l$  and  $p_r$  are in camera 3D coordinate system

## **Fundamental matrices**



- Matrix that maps a pixel in one image with its corresponding pixel on the other image considering rotation, translation and intrinsic parameters of the cameras:
- $u_l^T F u_r = 0$
- $u_l$  e  $u_r$  are in image 2D coordinate system

## **Fundamental and essential matrices**



- Fundamental and Essential matrices represent the transformation between the stereo pair images. Fundamental matrix operates in image coordinates (pixels) and Essential matrix operates in physical coordinates.
- Possible to evaluate with 8 point correspondences (eight point algorithm: <a href="http://www.cs.unc.edu/~marc/tutorial/node">http://www.cs.unc.edu/~marc/tutorial/node</a>
   54.html)

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## **Image rectification**



 Given the fundamental matrix, it is possible to rectify an image by aligning epipolar lines in rows on the two rectified images getting a frontal parallel arrangement.

## **Image rectification**



Original images







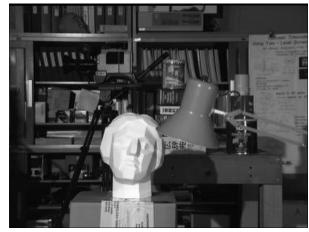
Rectified images

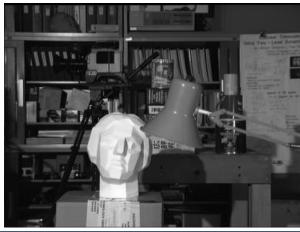
From Visual 3D Modeling from Images (<a href="http://www.cs.unc.edu/~marc/tutorial/">http://www.cs.unc.edu/~marc/tutorial/</a>)

## **Disparity map**



 In rectified images, disparity is simply difference between pixel coordinates xl and xr.







Tsukuba head and lamp stereo dataset

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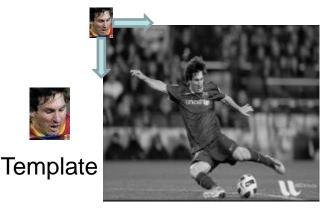
## **Template Matching**



• Template matching is moved to all positions (u, v) in image and computes p(a, b) to evaluate how well template matches image in that position



Original Image



**Template Matching** 



p(a,b) Matching function



Detected Point

• Object is on image at position where occurs  $\max(p(a,b))$  if  $\max(p(a,b)) > \text{threshold}$ 

# **Template Matching**



Several possible functions to compare template and

image p(a, b):

Square difference matching



Correlation coefficient matching

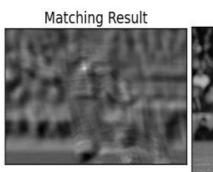
Matching Result

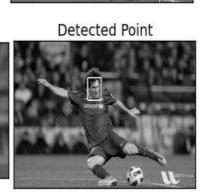
Matching Result

Detected Point

Detected Point

Detected Point





- ...

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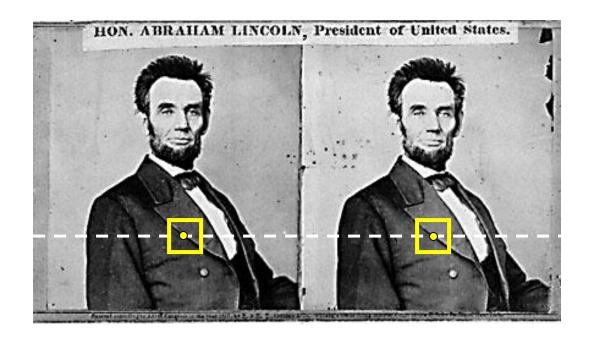
# **Stereo Matching algorithm**



- Match Pixels in Conjugate Epipolar Lines
  - This is a tough problem
  - Numerous approaches
    - A good survey and evaluation: <a href="http://vision.middlebury.edu/stereo/">http://vision.middlebury.edu/stereo/</a>

## **Basic stereo algorithm**





For each epipolar line

For each pixel in the left image

- compare with every pixel on same epipolar line in right image
- pick pixel with minimum match cost

# **Block Matching algorithm**

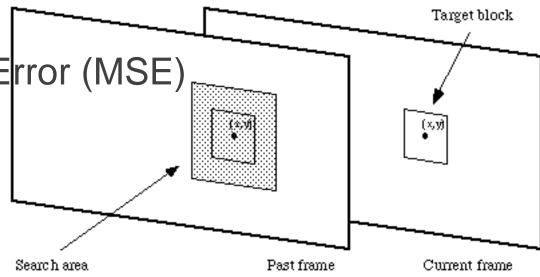


- Block Matching:
  - Divides an image into macroblocks and compare each with a corresponding block and its neighbours in a another image
- Several Metrics

Mean difference or Mean Absolute Difference
 (MAD)

Mean Squared Error (MSE)

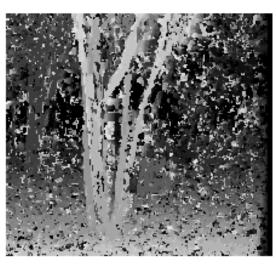
**–** . . .



#### Window size









W = 3

W = 20

#### Effect of window size

Smaller window

Larger window

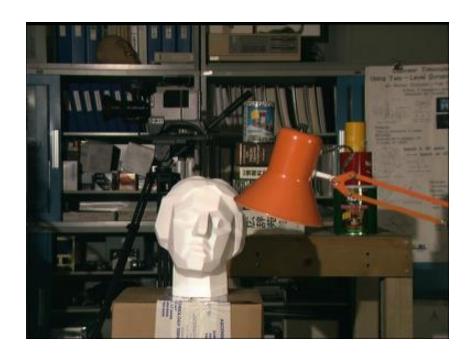
## Better results with adaptive window

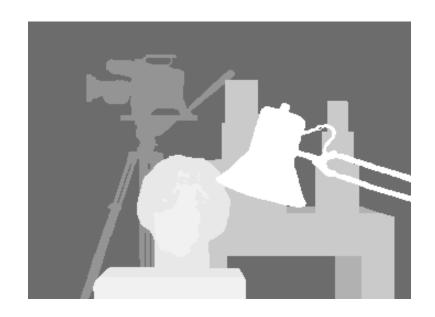
- T. Kanade and M. Okutomi, <u>A Stereo Matching</u>
   <u>Algorithm with an Adaptive Window: Theory and</u>
   <u>Experiment</u>, Proc. International Conference on
   Robotics and Automation, 1991.
- D. Scharstein and R. Szeliski. <u>Stereo matching with nonlinear diffusion</u>. International Journal of Computer Vision, 28(2):155-174, July 1998

## Stereo results



- Data from University of Tsukuba
- Similar results on other images with ground truth





Scene

Ground truth

# **Stereo: Real application**



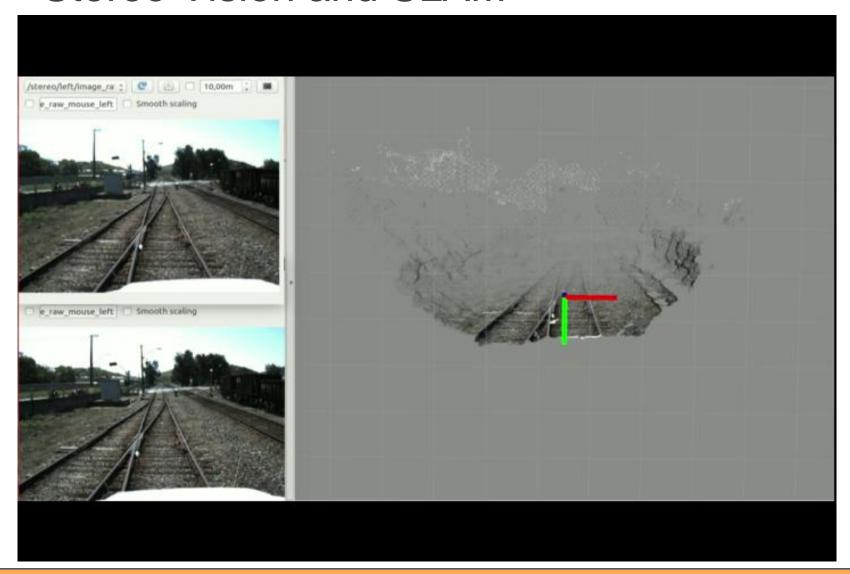
- Sistemas Embarcados de Vistoria (SEV)
  - Stereo vision system with two AVT Mako cameras
  - baseline is around 0.1 meters
  - horizontally aligned.



# **Stereo: Real application**



Stereo Vision and SLAM



## Stereo Vision - Errors



- Camera calibration errors
- Poor image resolution
- Occlusions
- Violations of brightness constancy (specular reflections)
- Large motions
- Low-contrast image regions

## **Stereo Vision - Steps**



- Calibrate cameras
- Rectify images
- Compute disparity
- Estimate depth

# **Stereo Vision in OpenCV**



- cvFindChessboardCorners: detect chessboard corner in stereo images
- cvStereoCalibrate: calibrates stereo rig
- cvStereoRectify: computes rotations that make both camera planes the same.
- cvInitUndistortRectifyMap and cvRemap: use to compute undistortion map and rectified images
- Stereo correspondence (ex: cvFindStereoCorrespondenceBM): computes the disparity map.
- cvReprojectImageTo3D: disparity map to 3D with calibrated cameras



# OPENCV STEREO VISION DEMO

#### Some references



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- Olivier Faugeras Three-dimensional computer vision: a geometric viewpoint. MIT Press Cambridge, MA, USA ©1993
- Szeliski, R. (2010).. Computer Vision: Algorithms and Applications, Springer
- Quang-Tuan Luong. "Learning Epipolar Geometry".
   Retrieved 2007-03-04.