



# ***Reconstruction of three dimensional scenes***

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# Problem formulation

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***Three dimensional automatic scene sensing means capturing shape, appearance and spatial coordinates of real objects.***

***Shape*** - geometry of 3D object

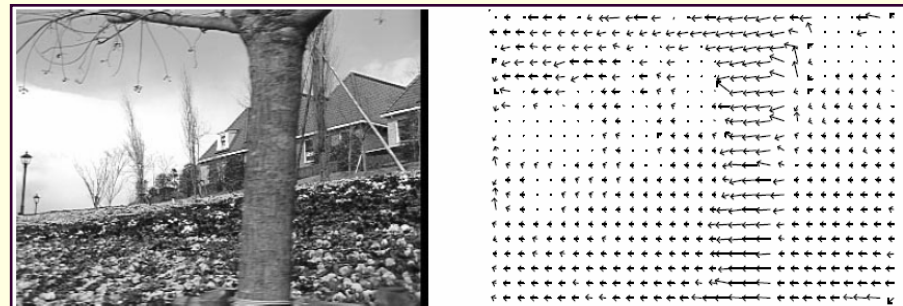
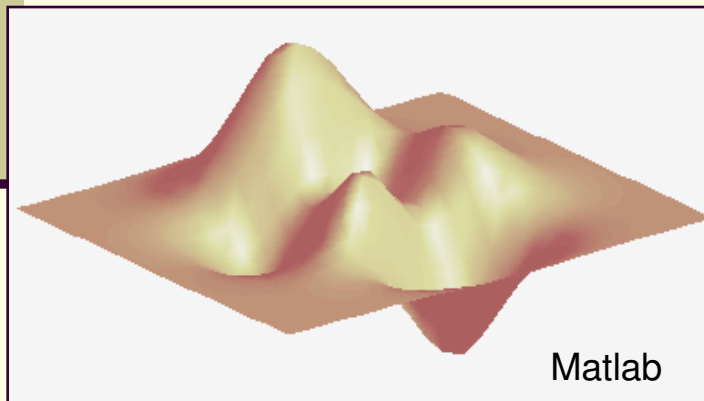
***Appearance*** - surface attributes: colour, texture, reflectance

***Spatial coordinates*** – (X, Y, Z) coordinates in 3D

*There are methods of 3D scene reconstruction from a set of 2D images*

# 3D scene reconstruction methods

- *Stereo image pair matching*
- *Structure from motion camera image sequence*
- *Shape reconstruction from shading*
- *Projection of structured light*
- *Laser scanning*



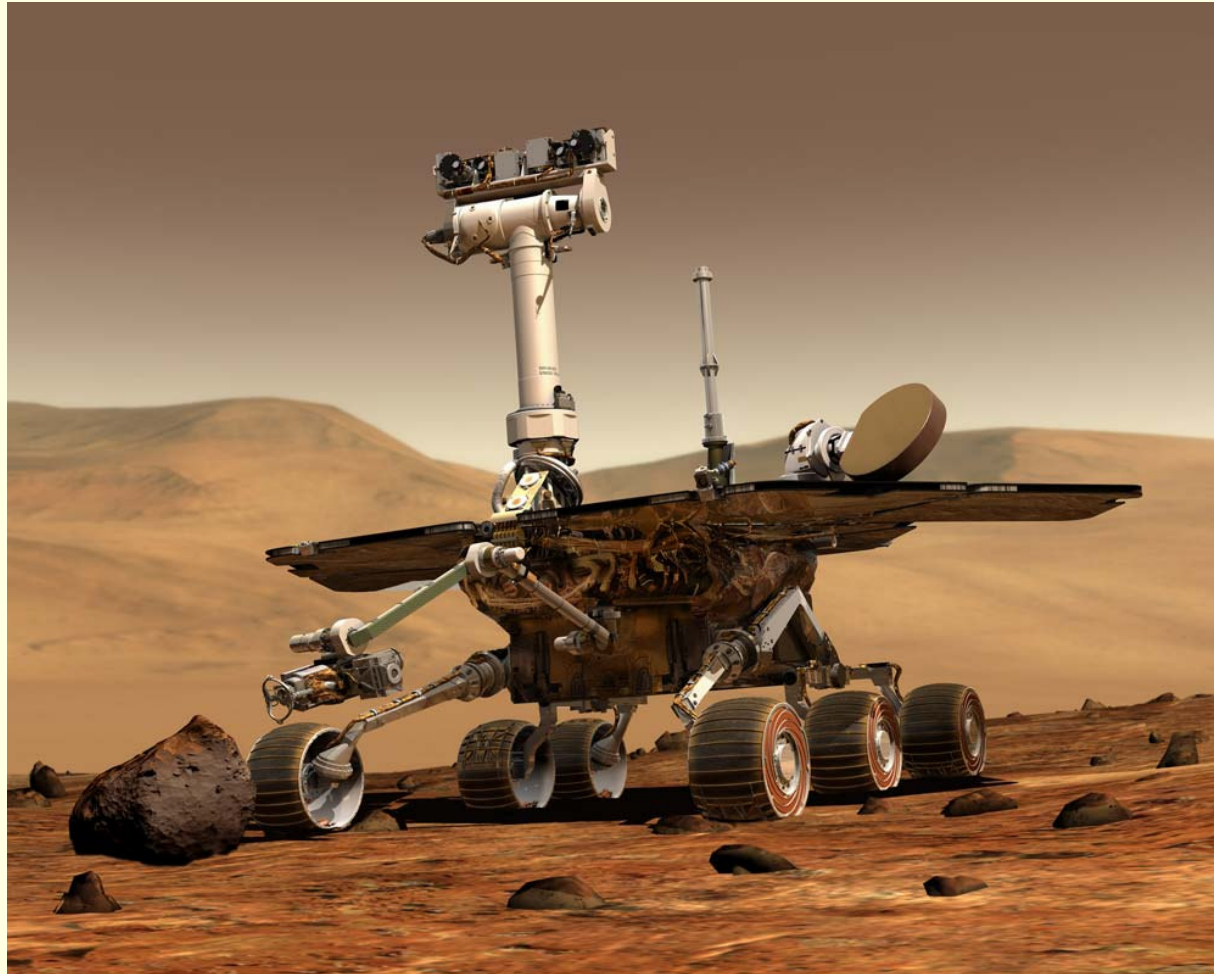
# Applications

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- *Robotics*
- *Medicine*
- *Electronic travel aid for the blind*
- *Product design*
- *Virtual reality*
- *Computer games*

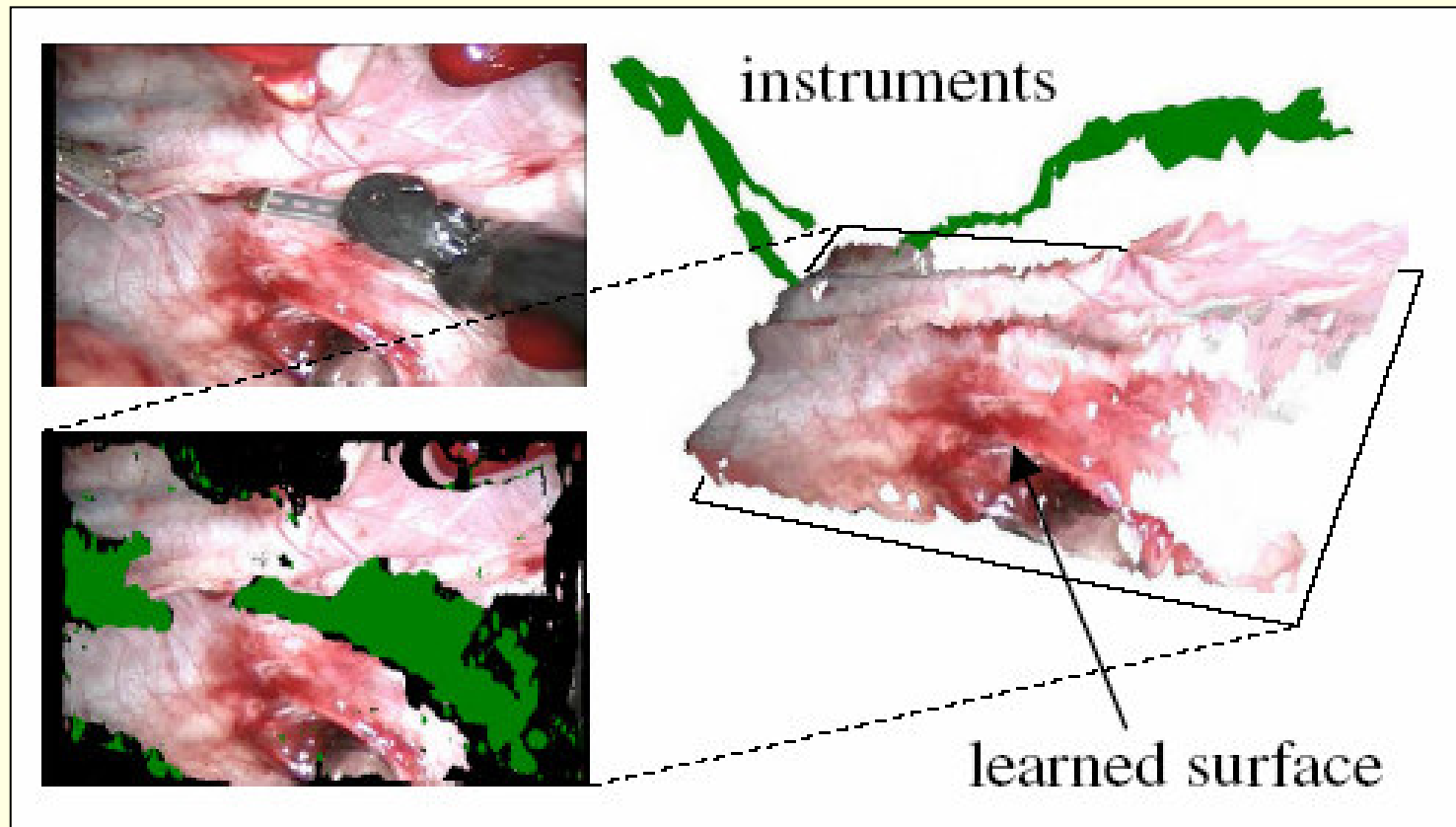
# Robotics - robot vision

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*Image copied from* : <http://marsrovers.jpl.nasa.gov>

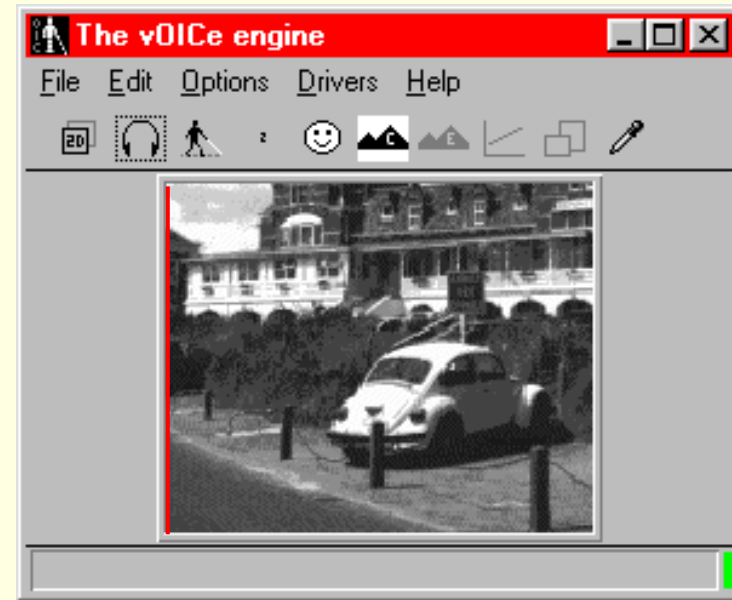
# Medical imaging: 3D endoscopic surgery



From: “3D Reconstruction of the operating field for image overlay in 3D-endoscopic surgery” by F. Mourgues, F. Devernay and E. Coste-Maniere

# The vOICe travel aid system for blind people

The vOICe in IEEE Spectrum, February 2004



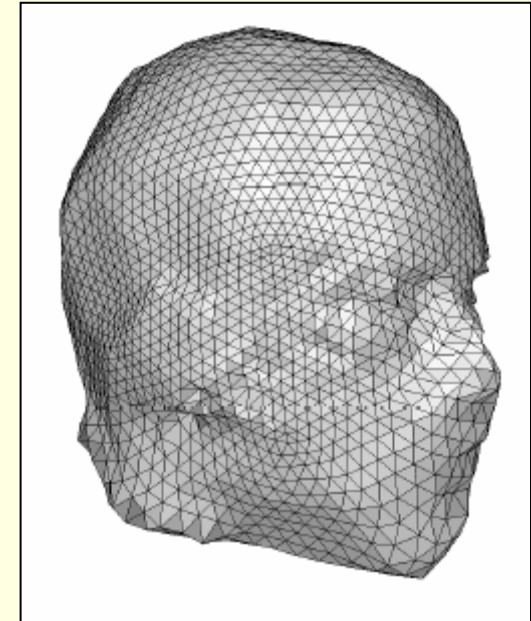
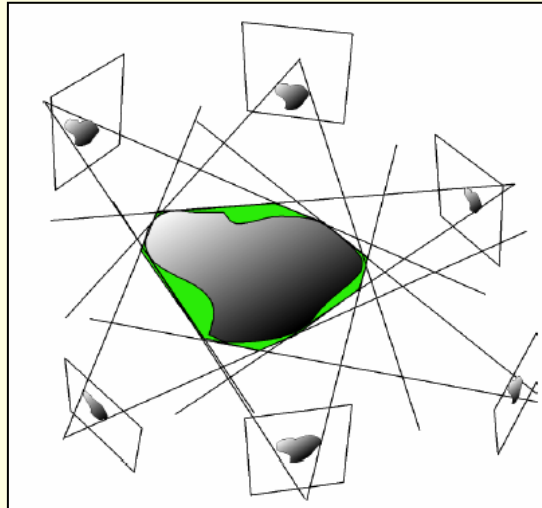
## Acoustic code:

- *pitch* – *object location*
- *loudness* – *object brightness*



# 3D object modelling

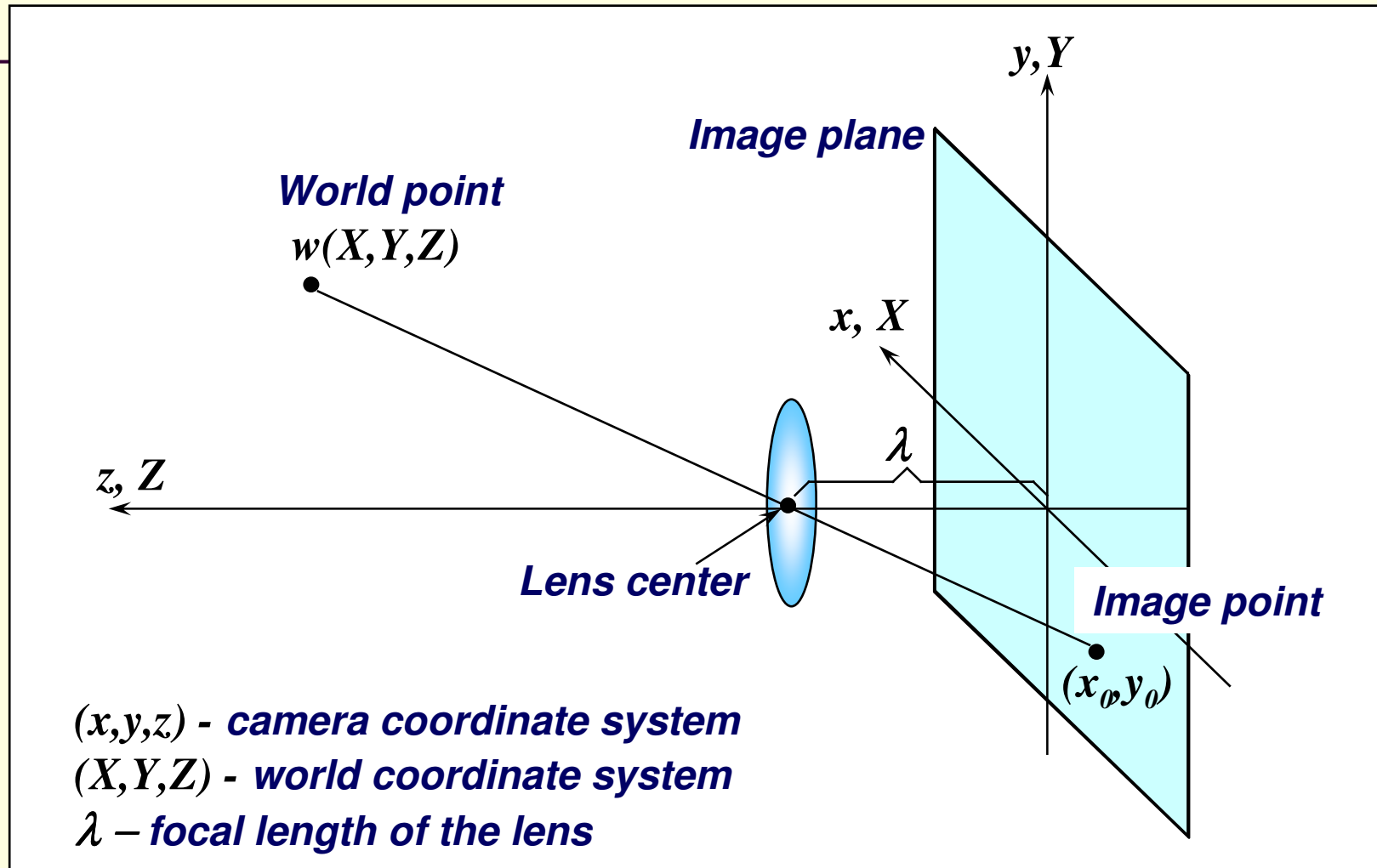
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Copied from the article: *“Spatio-Temporal Stereo using Multi-Resolution Subdivision Surfaces”* by J. Neumann and Y. Aloimonos



# Perspective Transformations



**Basic model of the image formation process  
assumed "pin-hole" camera model**

# Perspective Transformation

*Projection of the world point onto the image plane*

**Assumptions:**

- *world and camera coordinate systems are identically aligned*
- $Z > \lambda$

$$\frac{x_0}{\lambda} = -\frac{X}{Z - \lambda} = \frac{X}{\lambda - Z}$$

$$\frac{y_0}{\lambda} = -\frac{Y}{Z - \lambda} = \frac{Y}{\lambda - Z}$$



$$x_0 = \frac{\lambda X}{\lambda - Z}$$

$$y_0 = \frac{\lambda Y}{\lambda - Z}$$



*Nonlinear versus Z*

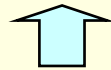
# Inverse Perspective Transformation

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*Projection of the image point onto the world point*

$$X = \frac{x_0}{\lambda} (\lambda - Z)$$

$$Y = \frac{y_0}{\lambda} (\lambda - Z)$$



*Two equations three unknowns*

# Conclusions

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*Mapping of 3D scene onto the image plane is a **many-to-one transformation**: image point corresponds to a set of collinear 3D points*

*The inverse transformation cannot be performed on the basis of a single image*

# Other cues about distance from a monocular view

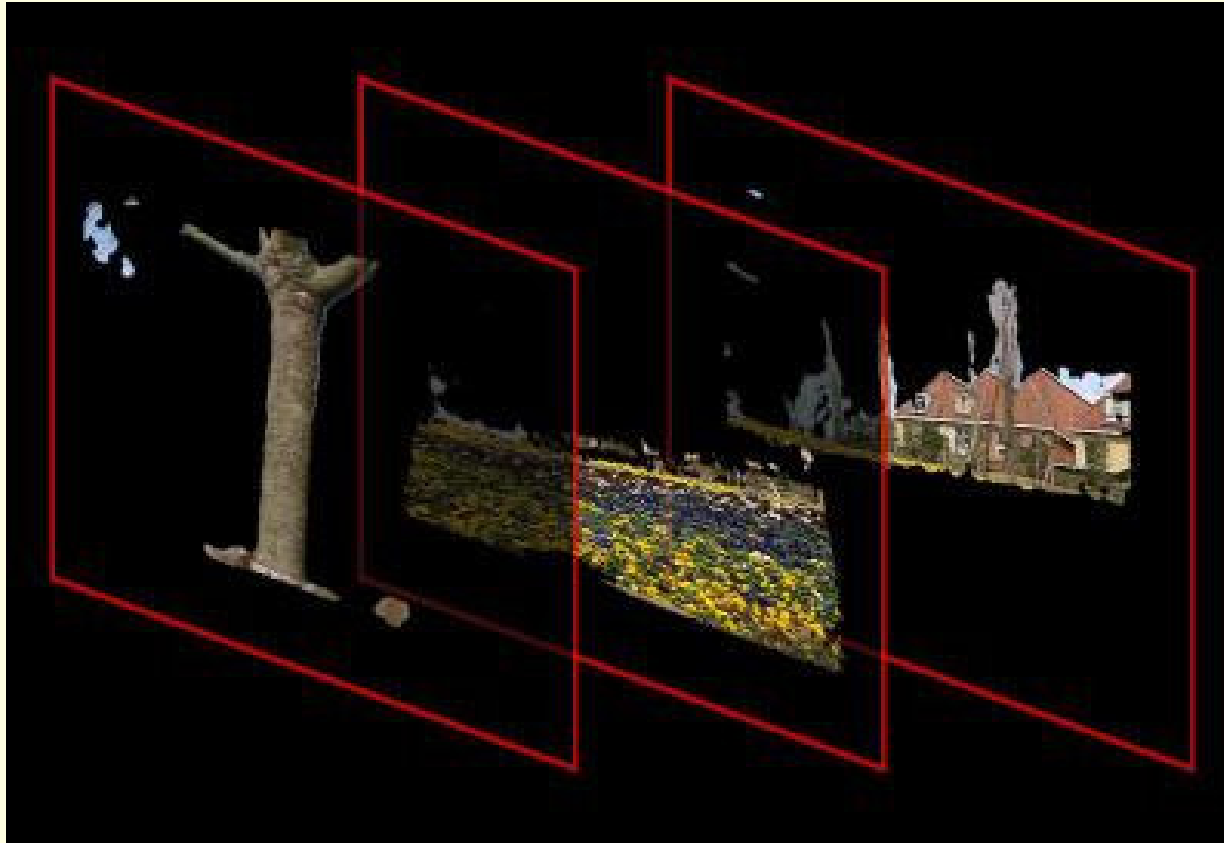
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***Motion***

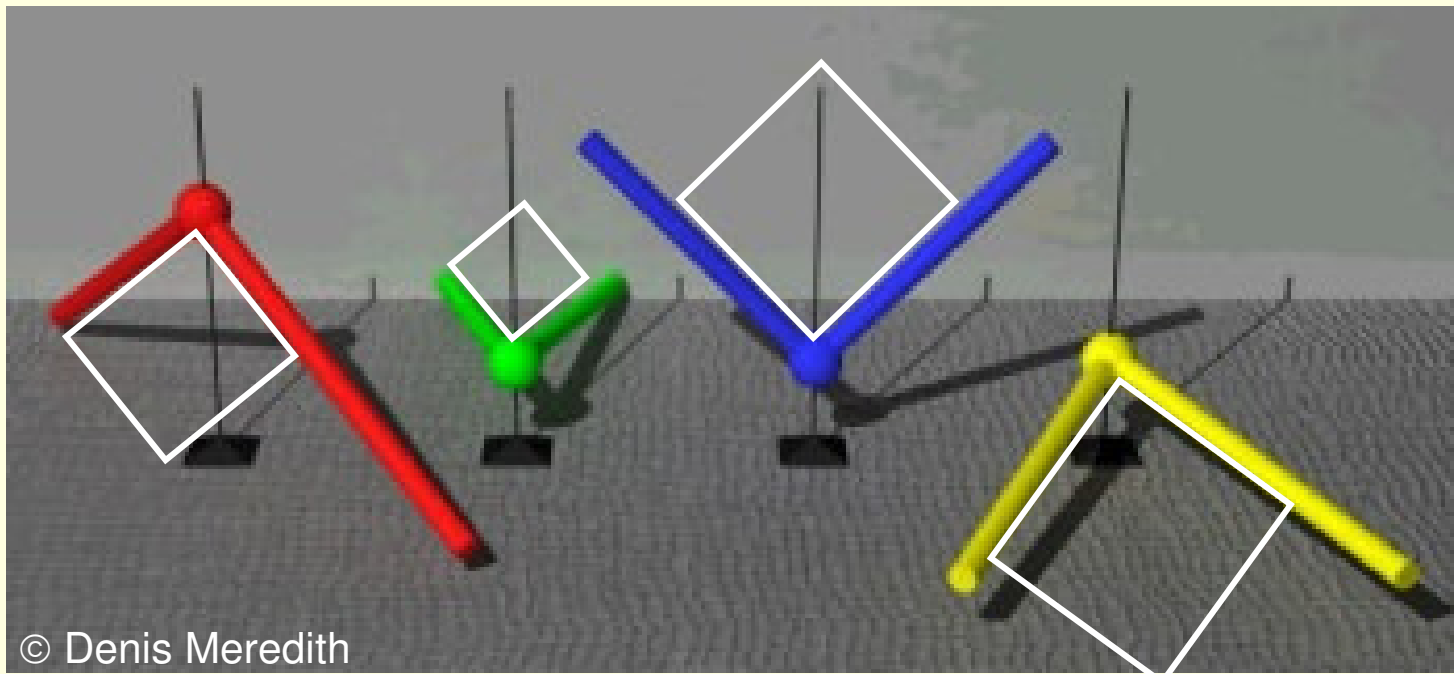
# Other cues about distance from a monocular view

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# Other cues about distance from a monocular view

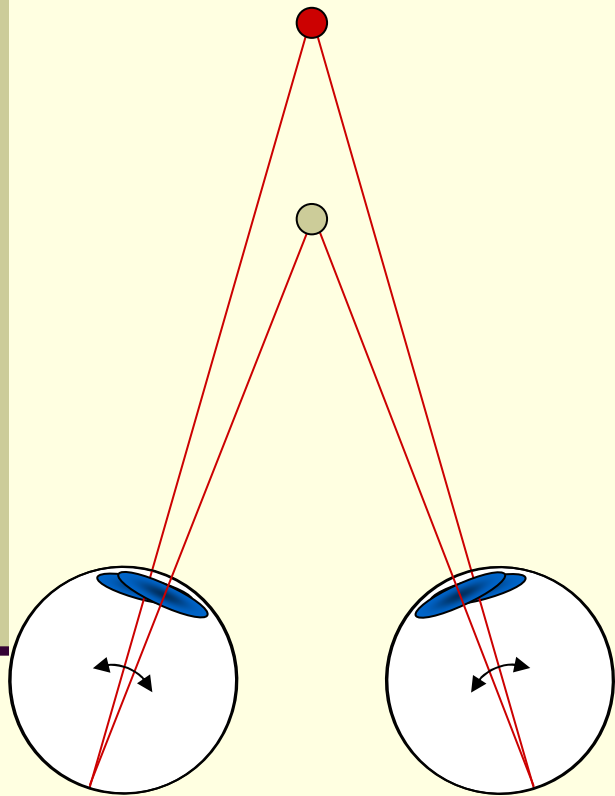
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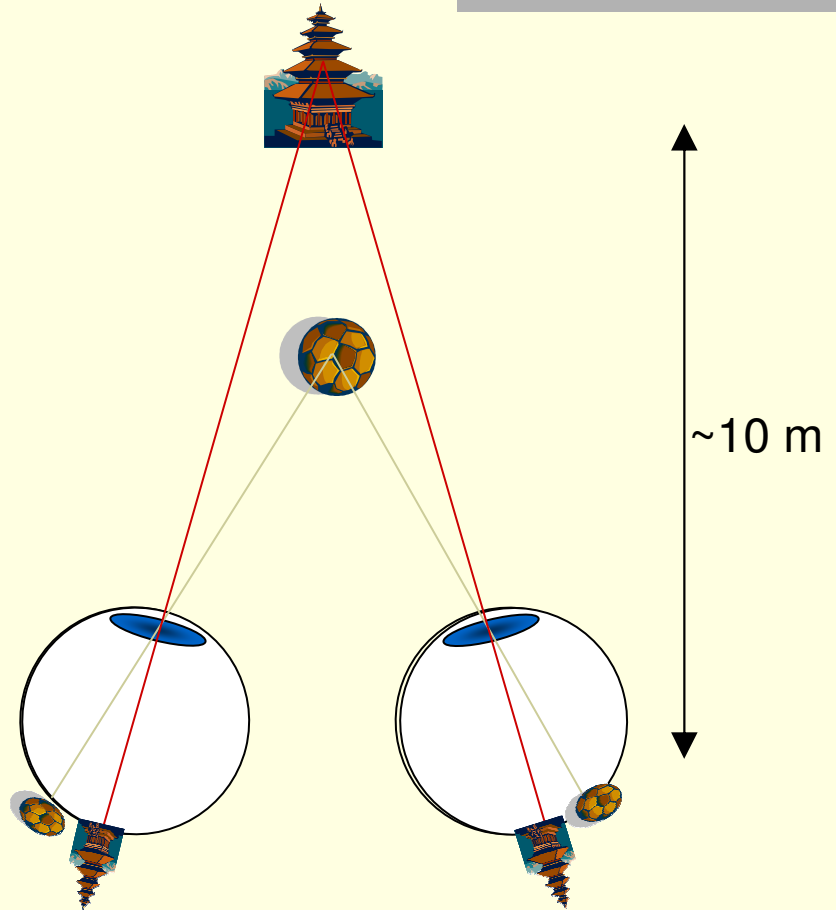
*„Precoded” perception of perspective*



# Binocular vision

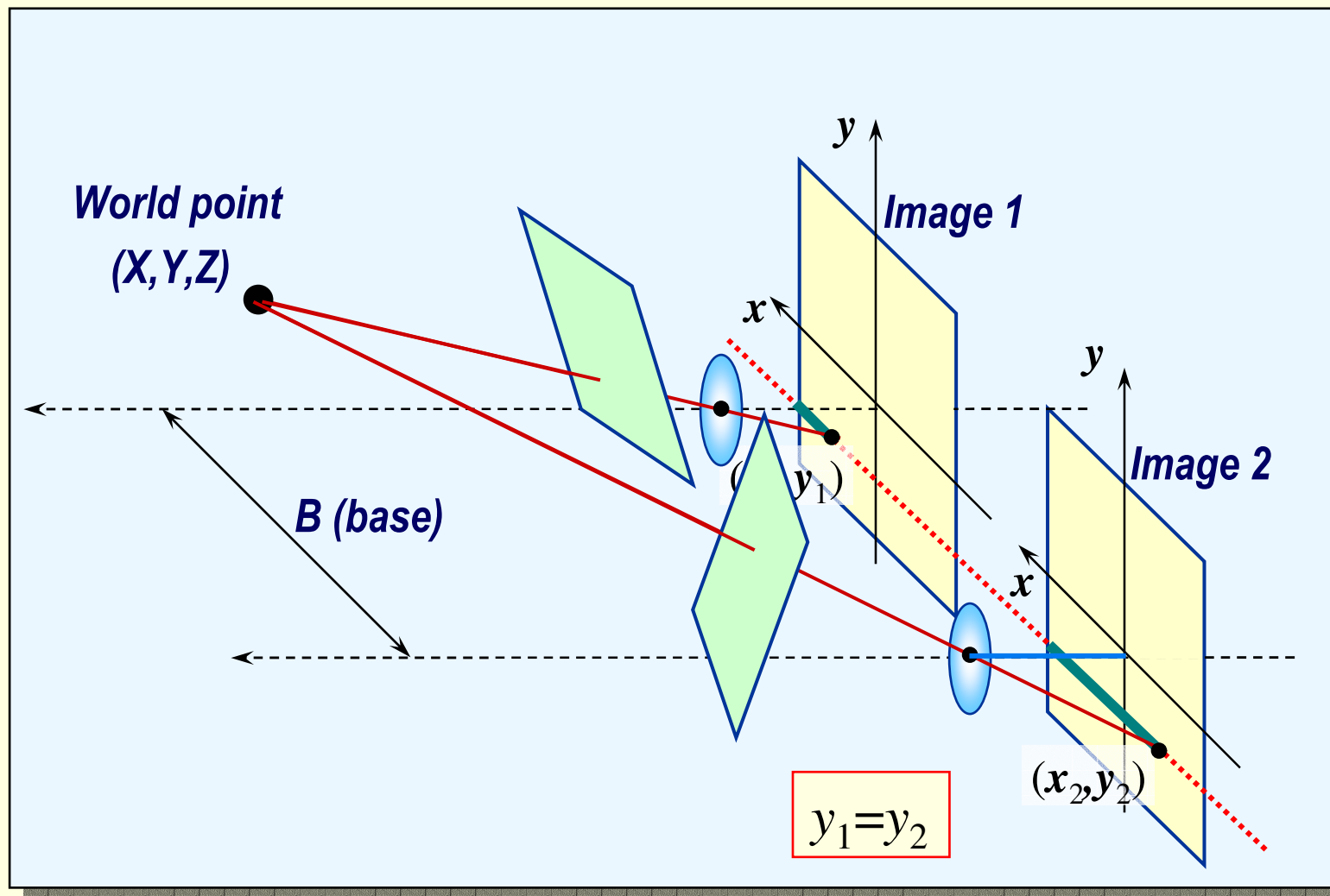


*Convergence of eyes  
optical axis*

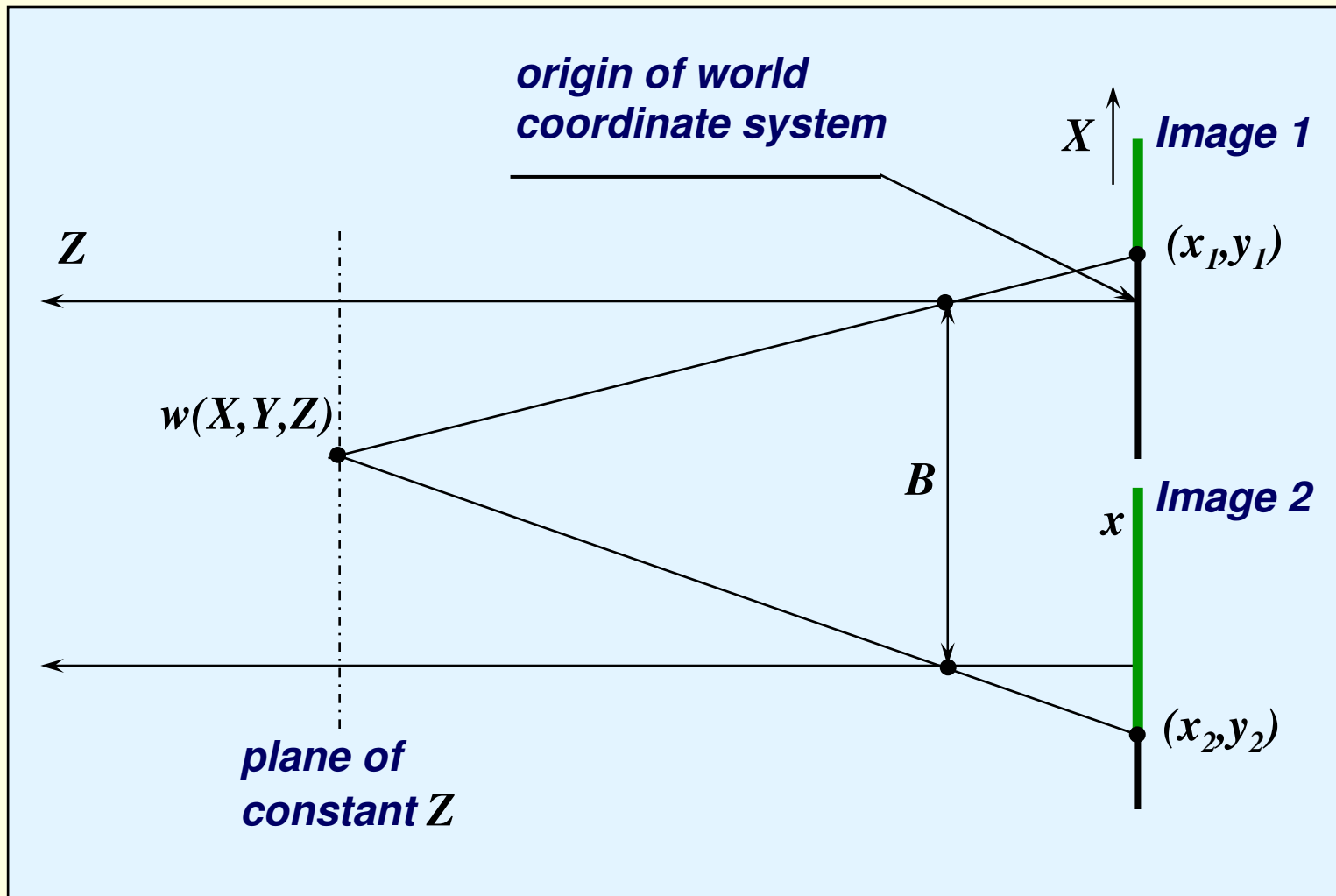


*Binocular disparity*

# Stereo image acquisition: (epipolar case)



# Stereo image acquisition - top view



# 3D point reconstruction in stereoscopy: Inverse Perspective Transformation

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$$X = \frac{x_1}{\lambda} (\lambda - Z)$$

$$Y = \frac{y_1}{\lambda} (\lambda - Z)$$

$$Z = \lambda - \frac{\lambda B}{x_2 - x_1}$$

*disparity*

*Three equations - three unknowns*

# Digital image matching: correspondence problem

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**Digital image matching** *automatically establishes the correspondence between primitives extracted from two or more digital images depicting at least part of the same scene*

## **Image matching problems:**

- *selection of primitives for matching*
- *choice of models for mapping of primitives*
- *measure of similarity of corresponding primitives*
- *matching algorithm*
- *matching strategy*

# Digital image matching approaches

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## Local correspondence methods:

- *Block matching*
- *Gradient-based optimisation*
- *Feature matching*

## Global correspondence methods:

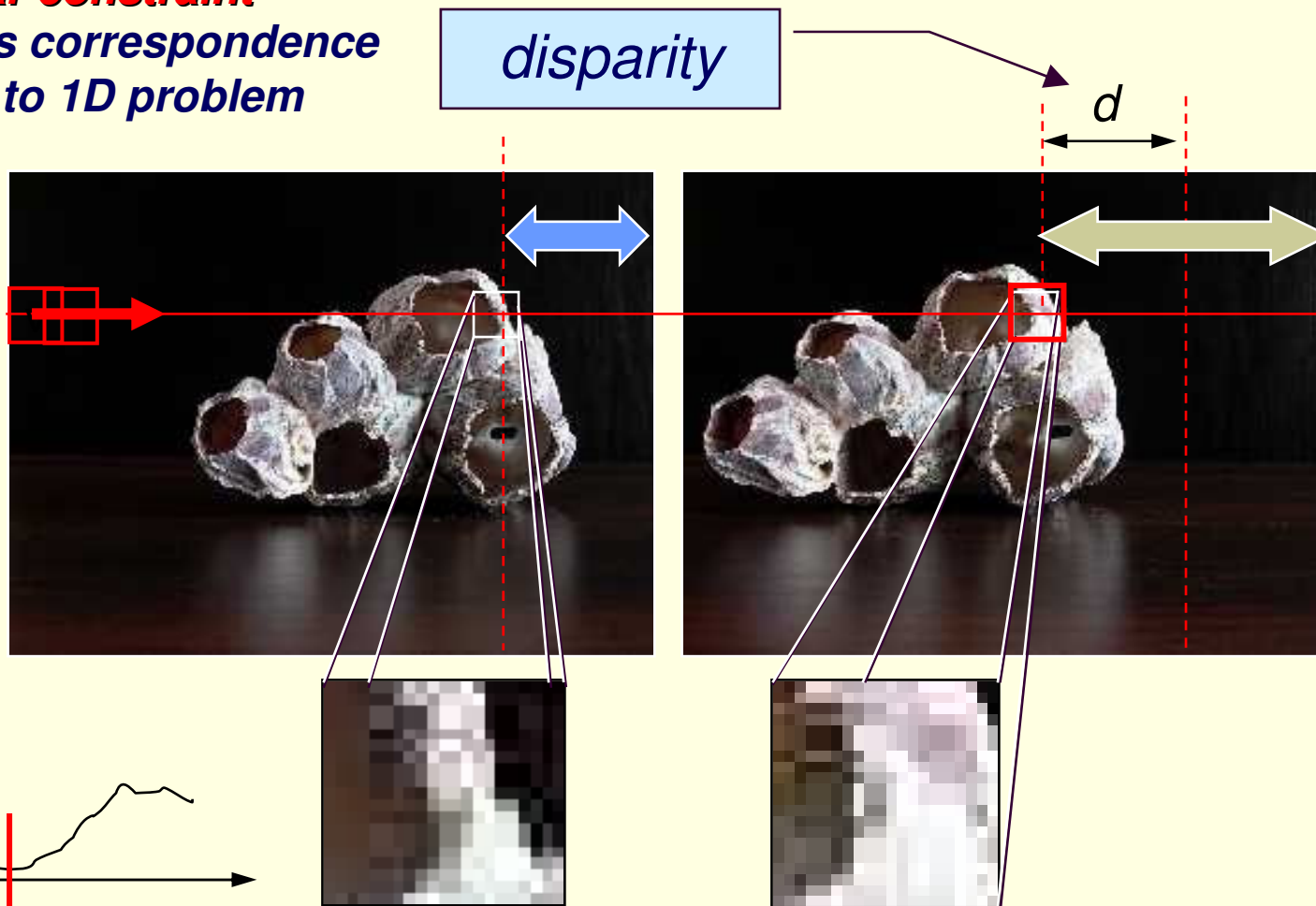
- *Dynamic programming*
- *Intrinsic curves*
- *Graph cuts*
- *Other methods*

# Digital image matching - concept of disparity

**Epipolar constraint**  
reduces correspondence  
search to 1D problem

here:  
 $y_2 = y_1$

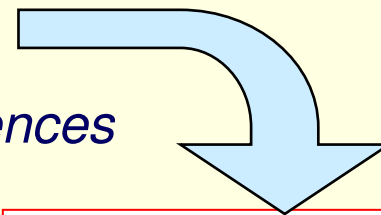
disparity





# Metrics in block image matching

- *Normalised cross-correlation*
- *Sum of squared differences SSD*
- *Normalised sum of squared differences*
- *Sum of absolute differences SAD*
- *Rank*
- *Census*



$$\sum_{x,y} (I_1(x,y) - I_2(x+d,y))^2 \rightarrow \min$$

89	63	72
67	55	64
58	51	49

→ 2

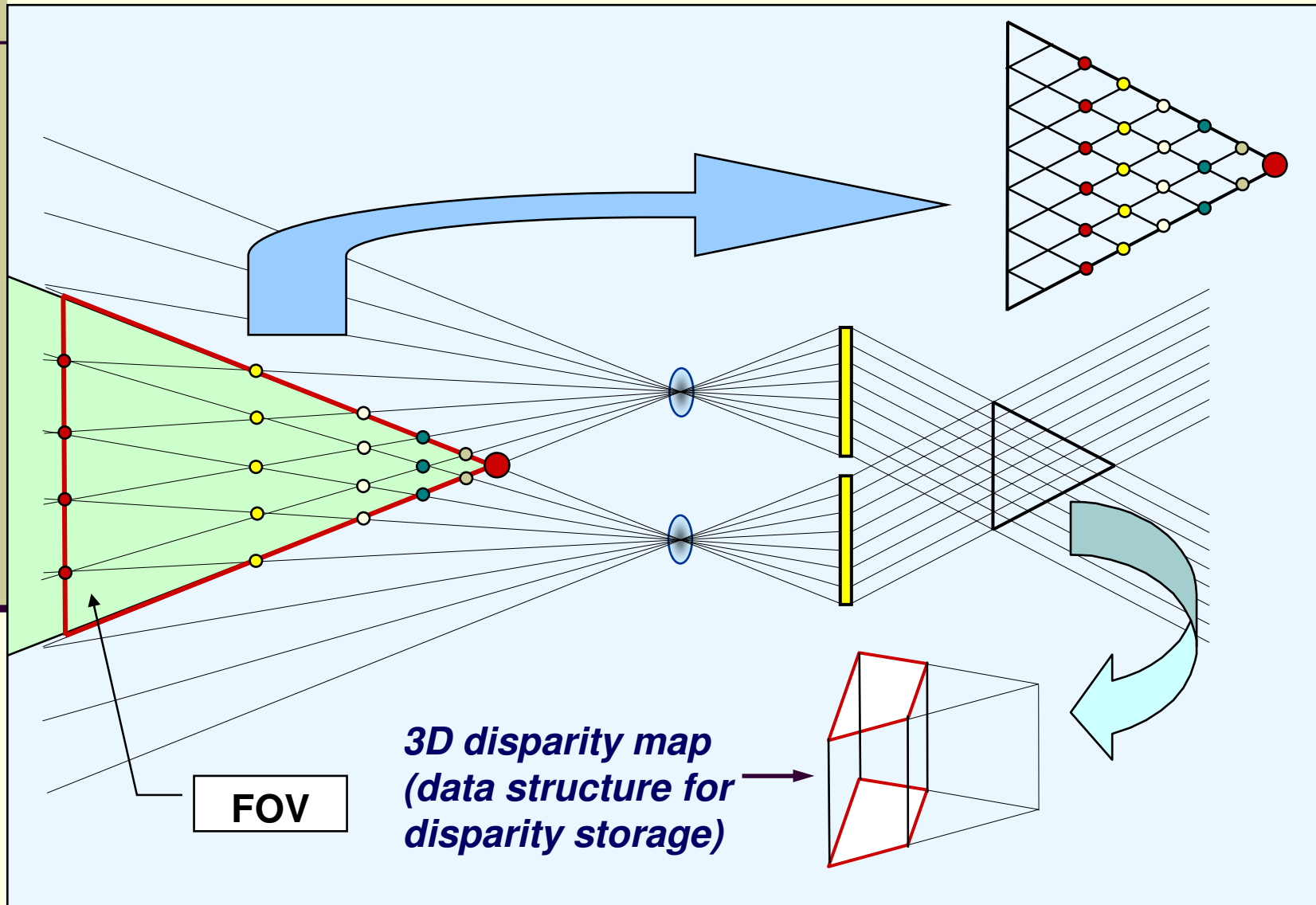
*Rank*

89	63	72
67	55	64
58	51	49

→ 00000110

*Census*

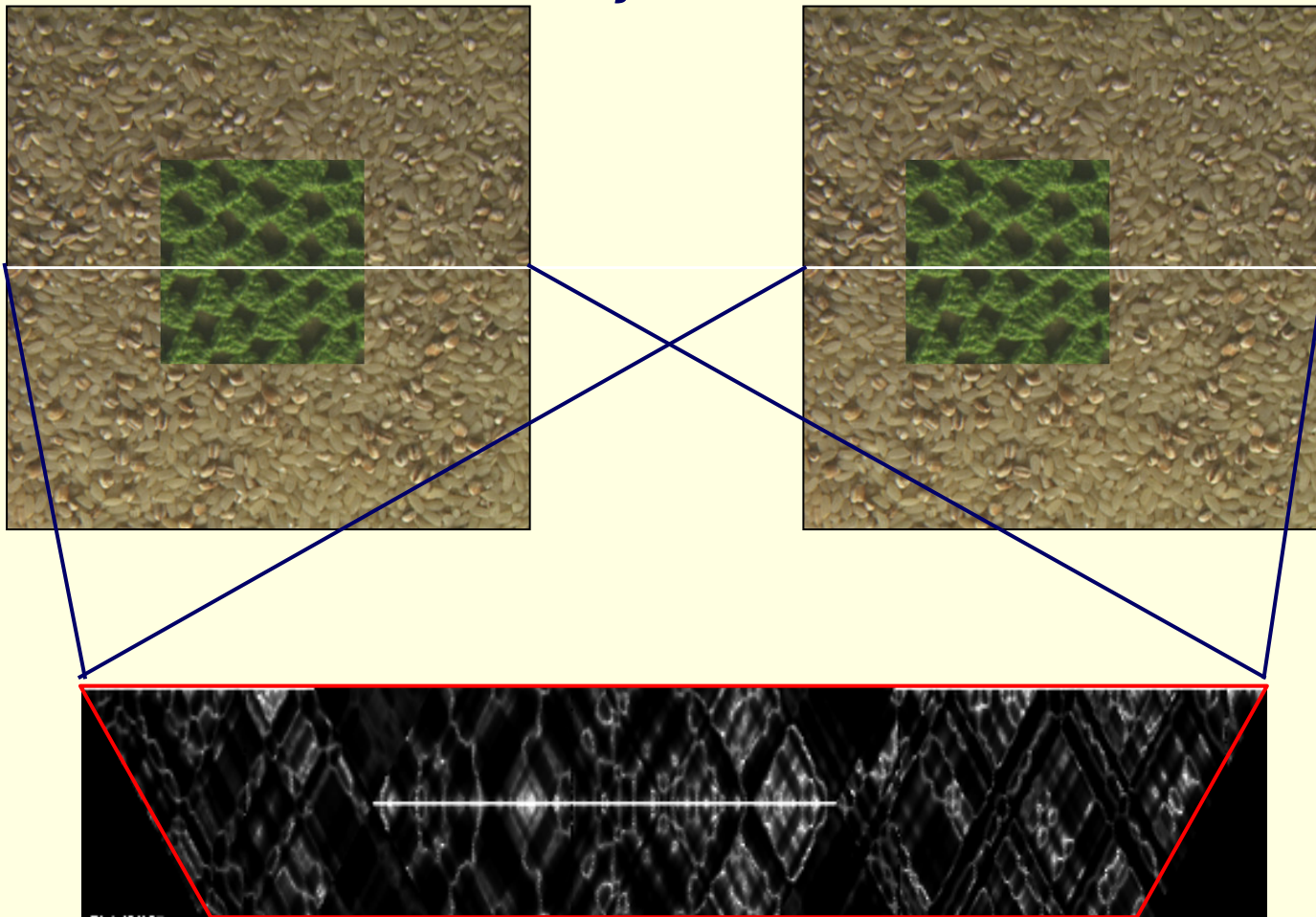
# Digital image matching - 3D disparity map building



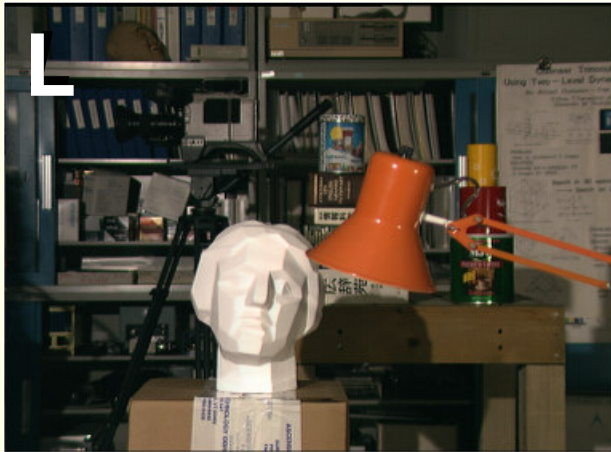
# Digital image matching

## - intersection of disparity map

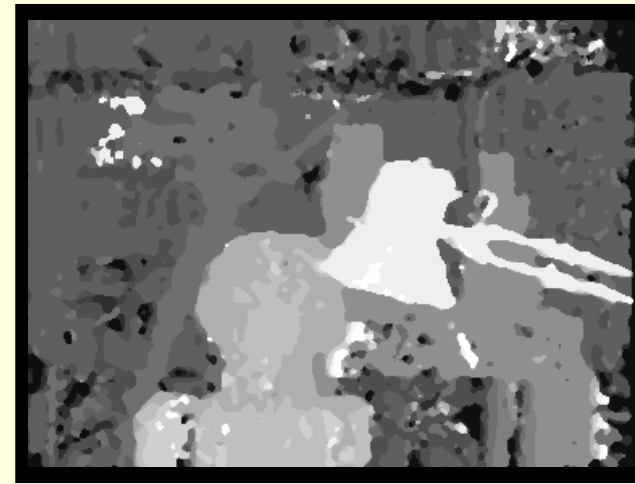
*Outex – University of Oulu Texture Database*



# Image data for verification of matching algorithms

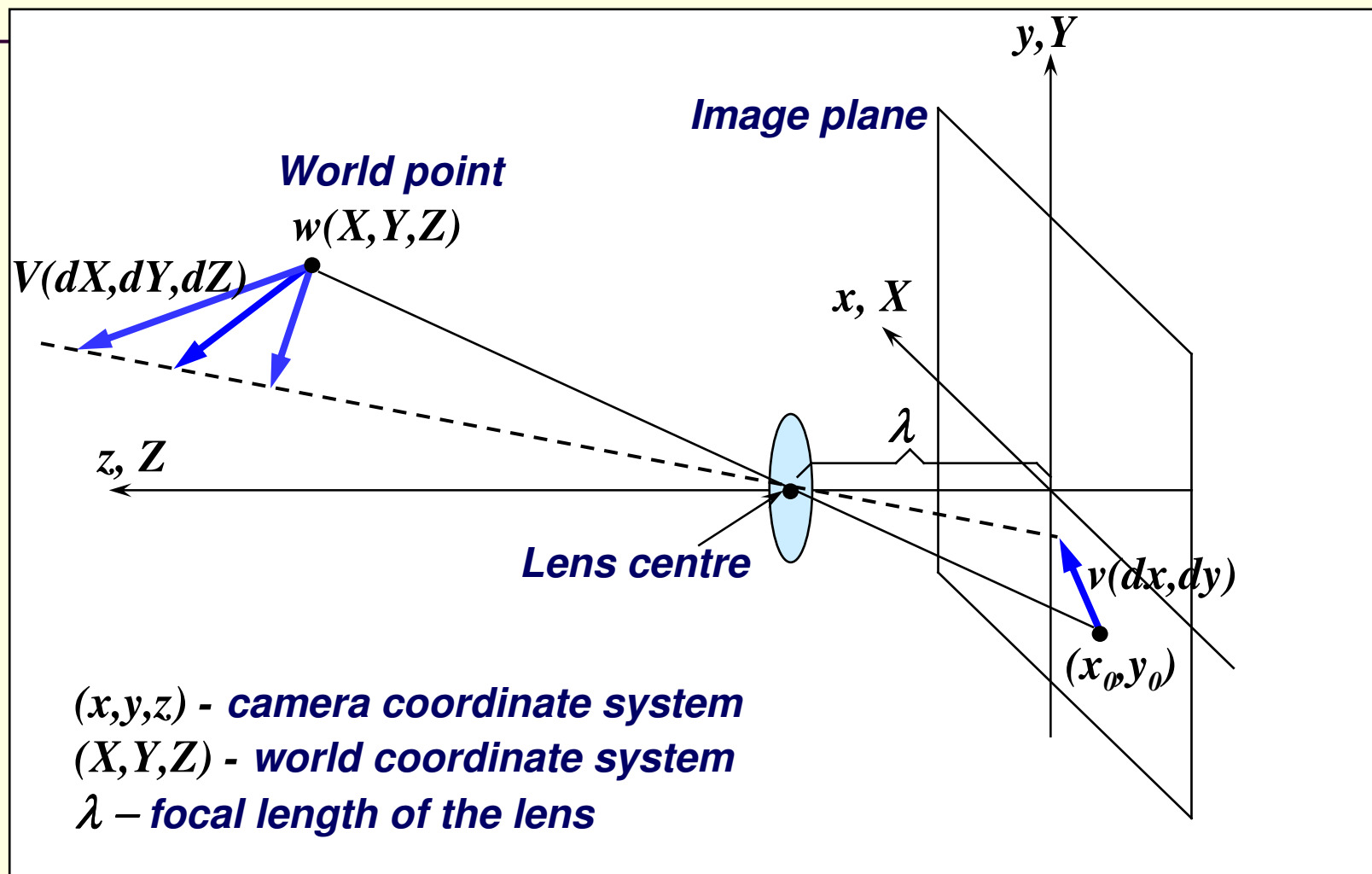


*ground truth disparity image*



*computed disparity image*

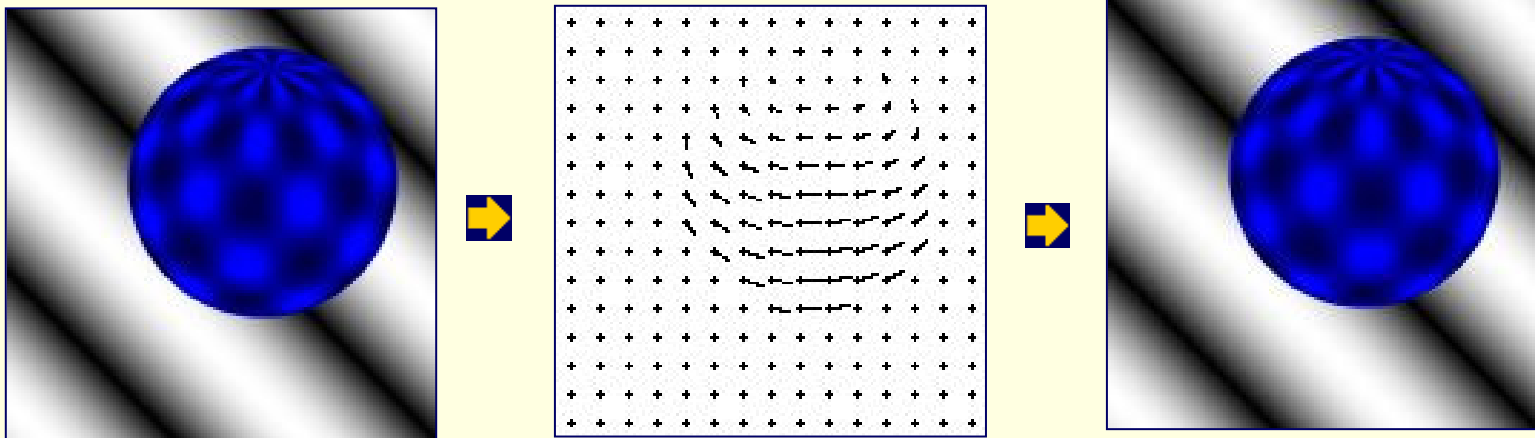
# Optical flow – projection of 3D movement vector



**Perspective transformations apply**

# The concept of optical flow

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Images from:

<http://www.cs.otago.ac.nz/research/vision/Research/index.html>

B.K.P. Horn, B.G. Schunk „**Determining Optical Flow**”, Artificial Intelligence, vol. 17, 1981, pp. 185-203



