



Div. Ingeniería de Sistemas y Automática

Universidad Miguel Hernández

INTRODUCTION TO 3D VISION



- Introduction
- Binocular settings
- Single camera equations
- A simple stereo system
- Correlation based methods

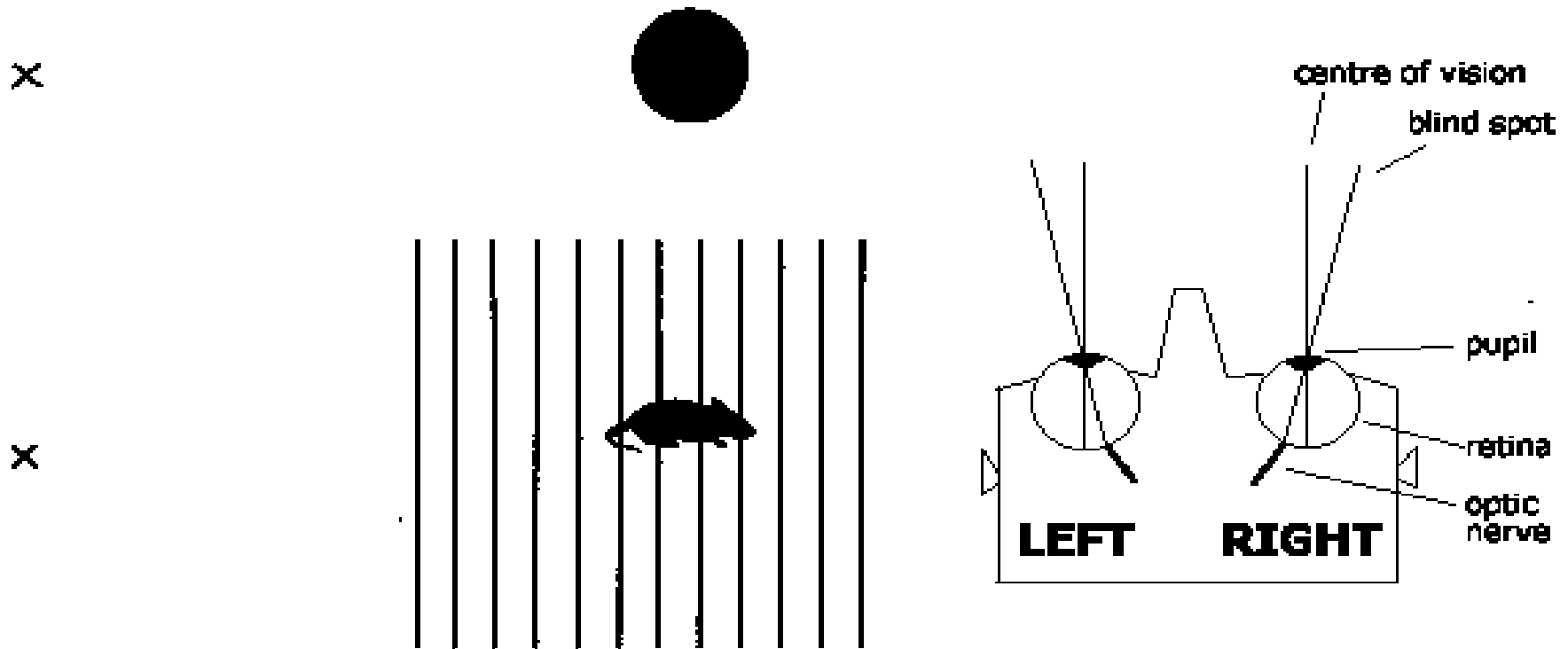
- 3D Vision aims at estimating the spatial information of the world by means of different images taken from different viewpoints.
- We will concentrate on stereo vision

- Most animals rely on stereo vision to extract 3D information from its environment.

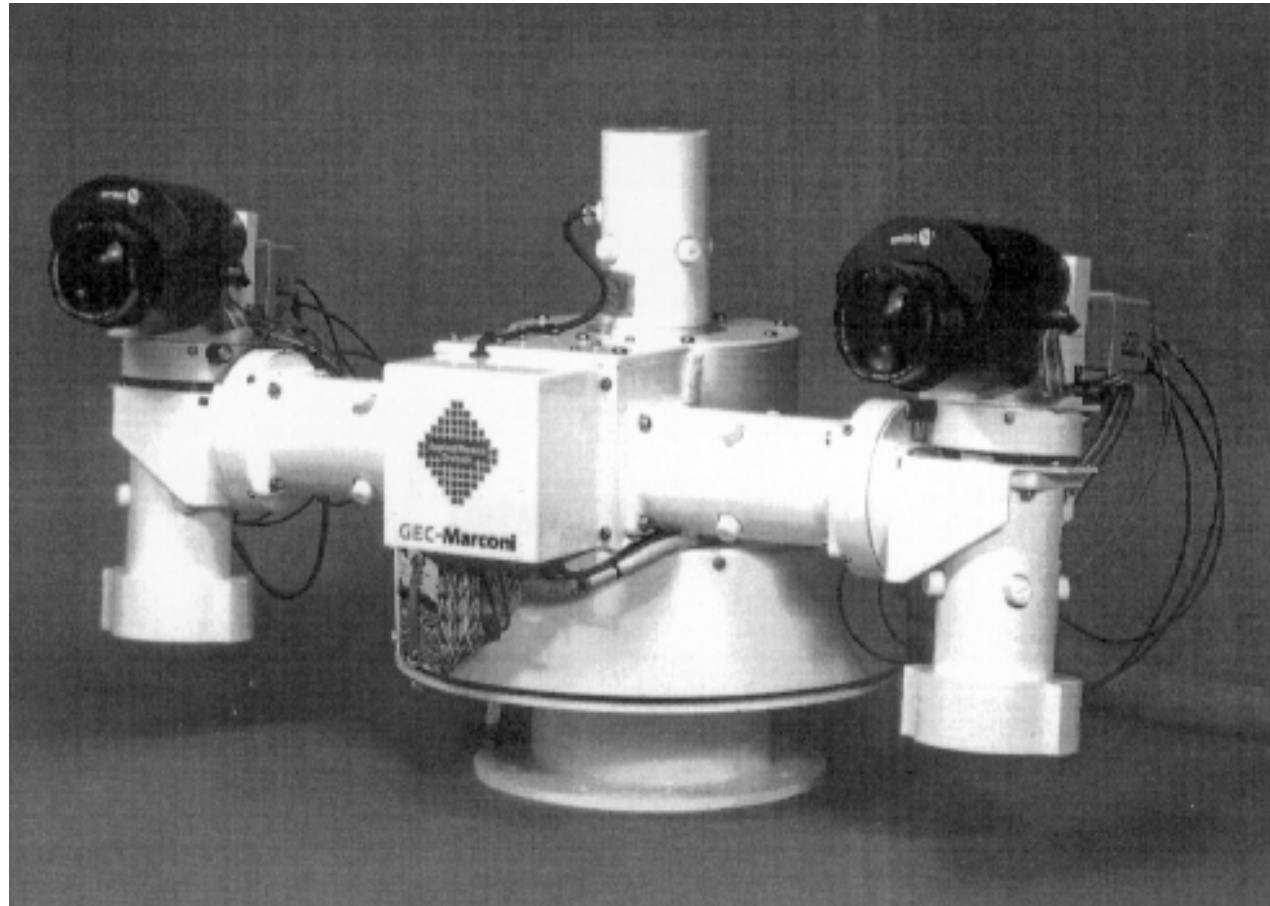


- Experiment:
 - Hold one thumb at arm's length and close the right and left eye alternatively.
 - With (little) surprise you find that the relative position between thumb and background changes. It is this difference from viewpoint that is used to compute 3D information.
 - But... try to explain why, with both eyes open you only see one thumb.
 - And.... What about the blind spot?... It is filled with information coming from the other eye.

- Experiment 2: Blind spot.
 - Close your left eye and focus on the cross with your right eye; hold the paper at about 25cm apart from your face. Approach or separate the paper slowly until the black point “disappears”.
 - a) Why does this fact occur?
 - b) Repeat the exercise with the image below. What happens with the barrels?



- Two problems in stereo:
 - The problem of correspondence:
 - Which item in the left eye corresponds to which item in the right eye?
 - 3D reconstruction:
 - Our brain computes the difference in retinal position between corresponding items (disparity).
 - If the geometry of the stereo system is known, then we can build up a 3D map of the viewed scene.

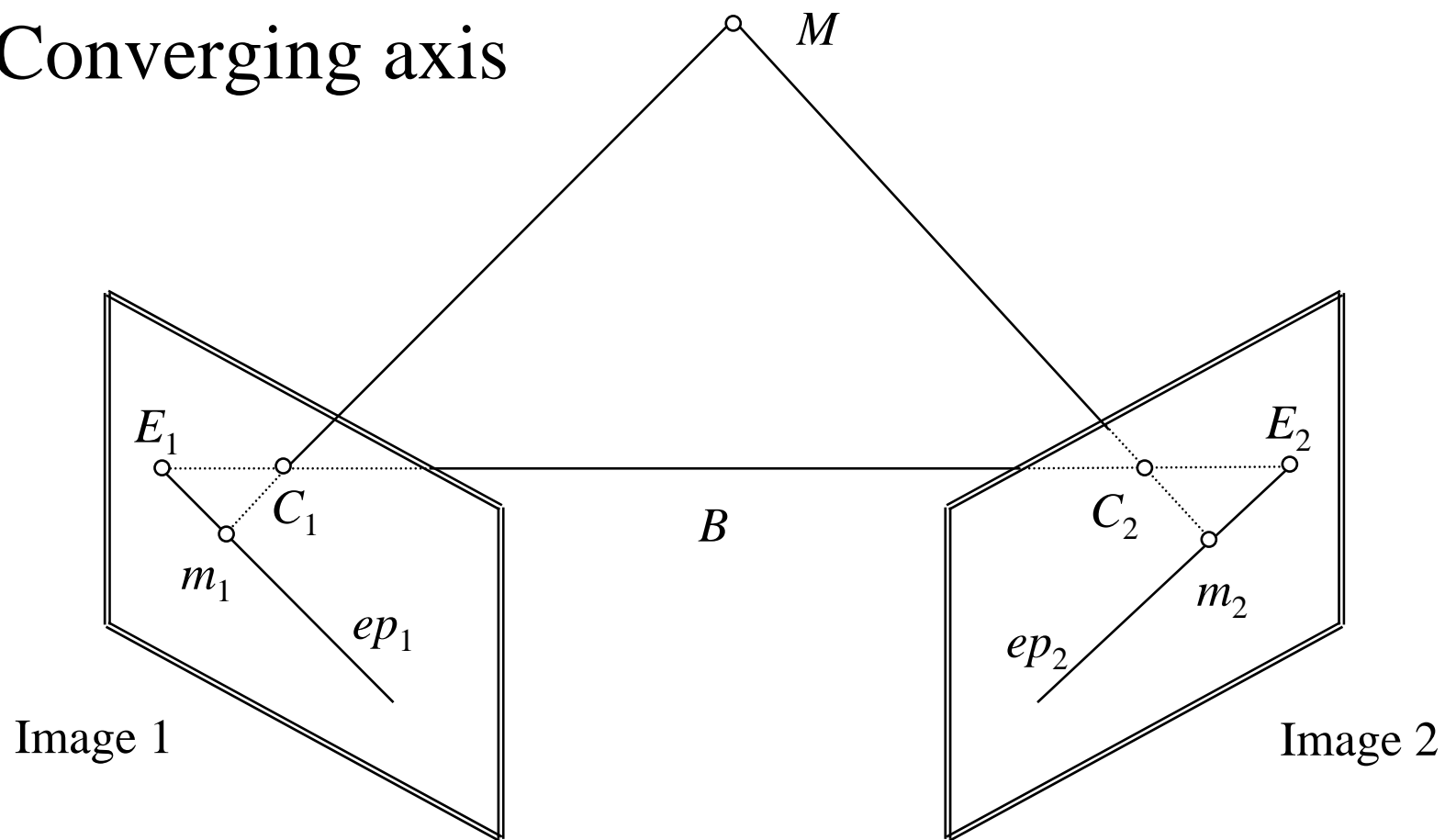


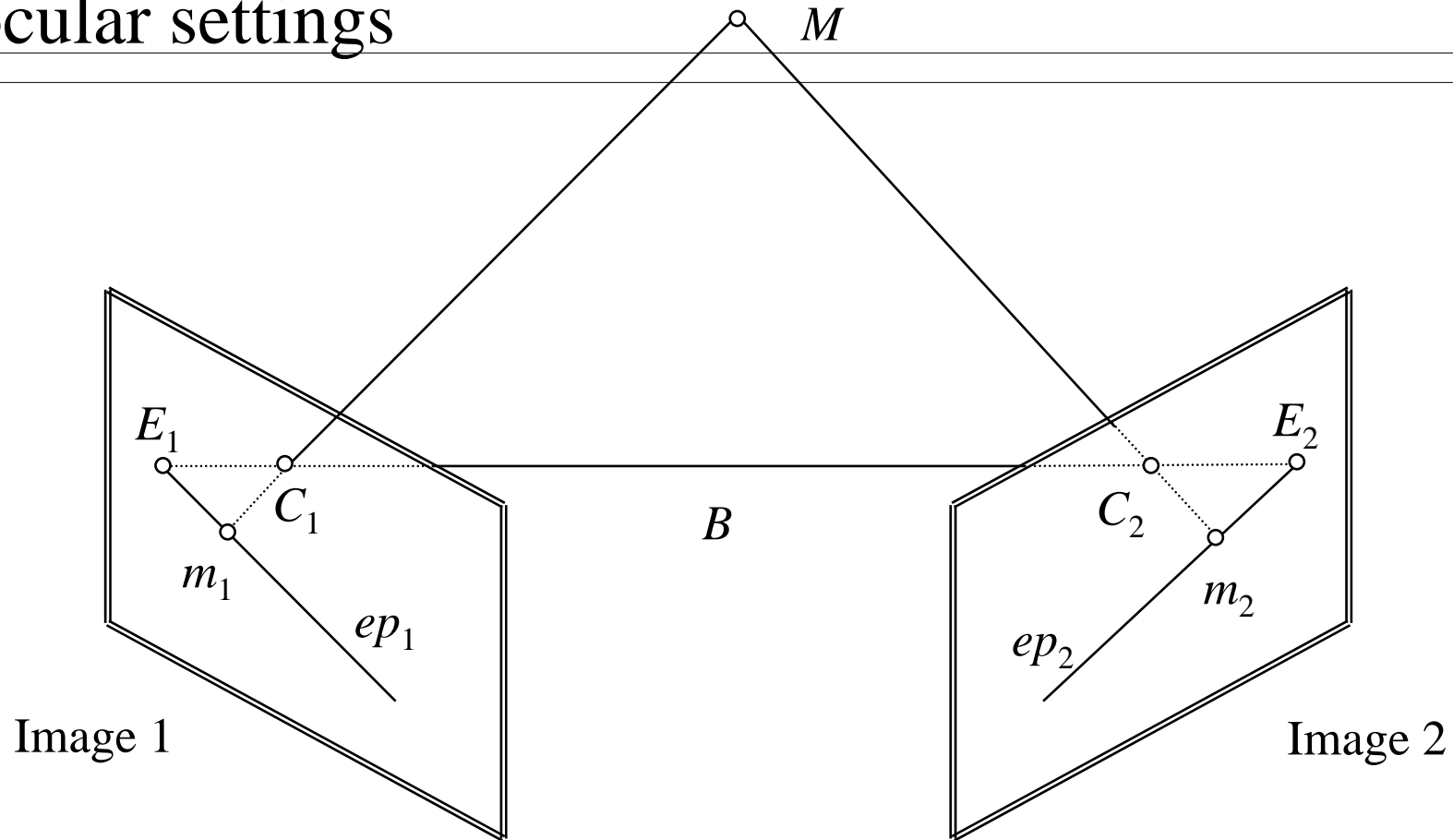
Stereo rig with $(3+3+3)$ DOF

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- *Optical axis can have different configurations:*
 - *Parallel axis*
 - *Converging axis*
 - *Arbitrary (Axis do not cross in space)*

Converging axis



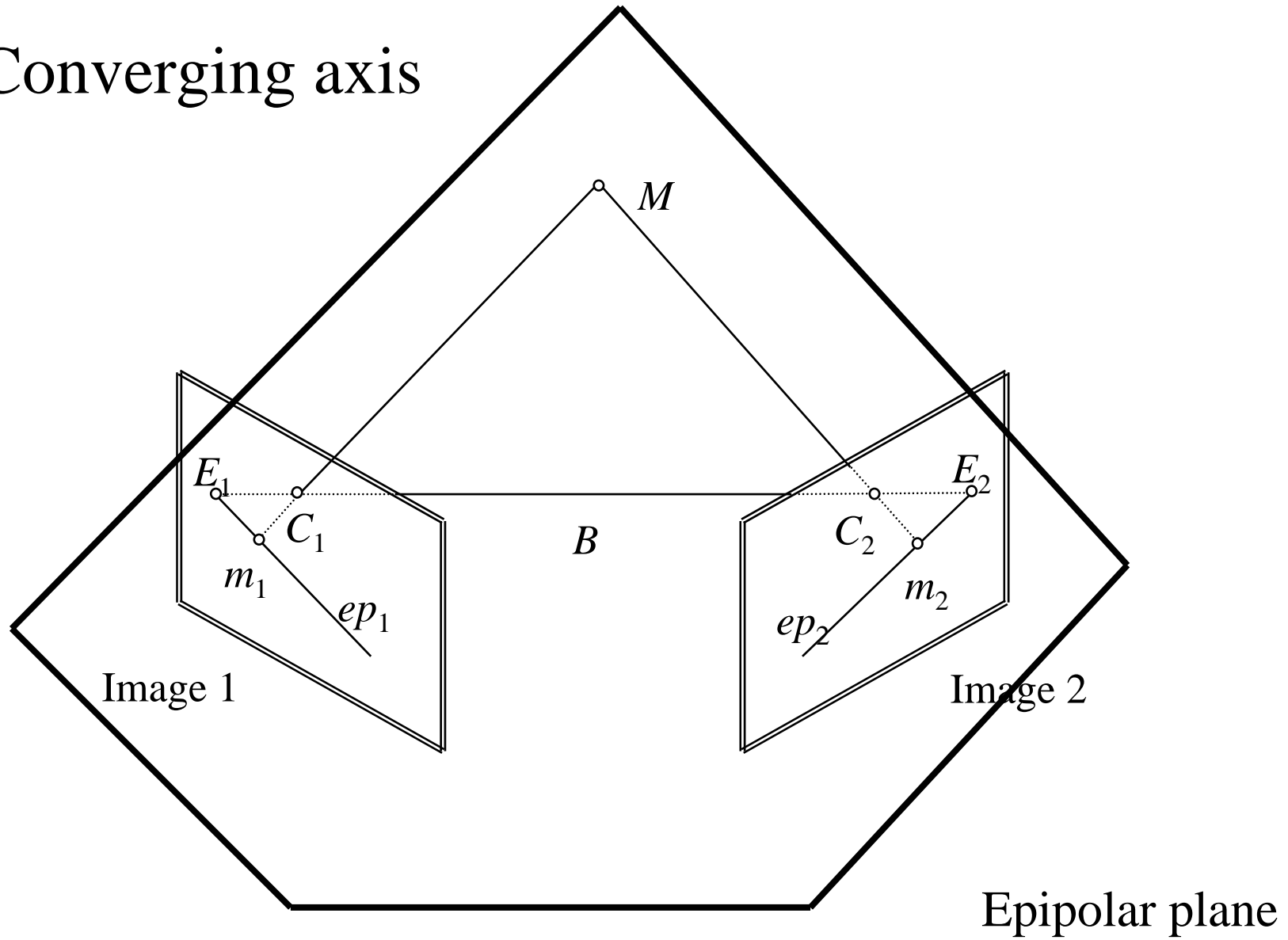


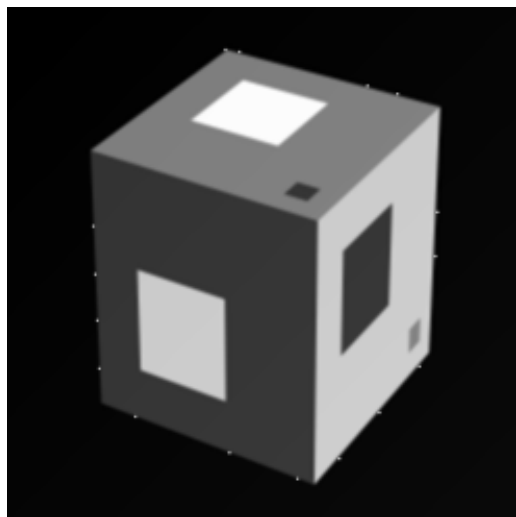
Baseline: Line through both optical centers

Epipolar plane: Plane that contains the optical centers C_1 and C_2 and the point M .

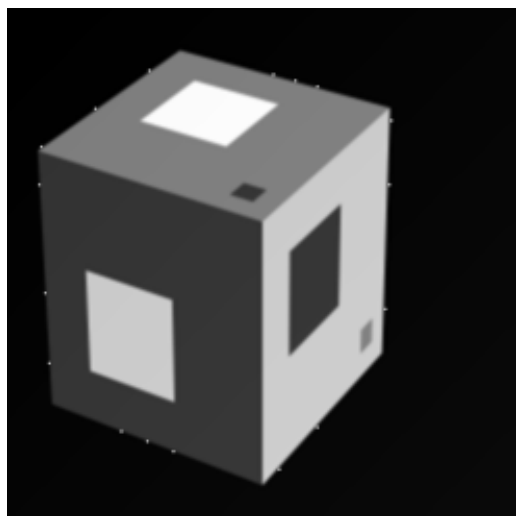
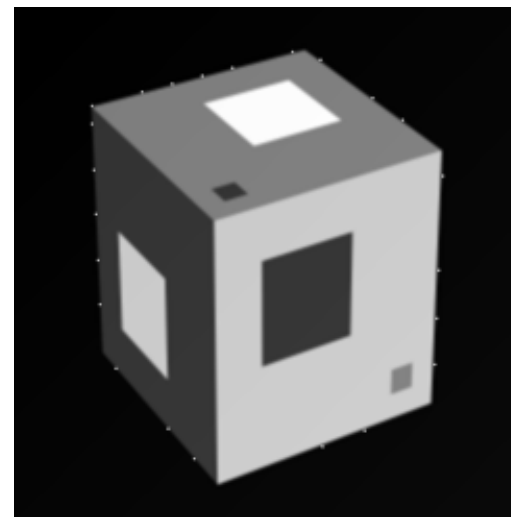
Epipolar line: Intersection of the epipolar plane and the each image plane.

Converging axis

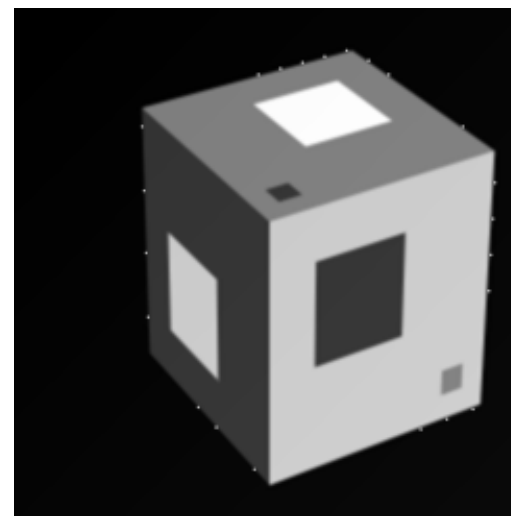




Parallel



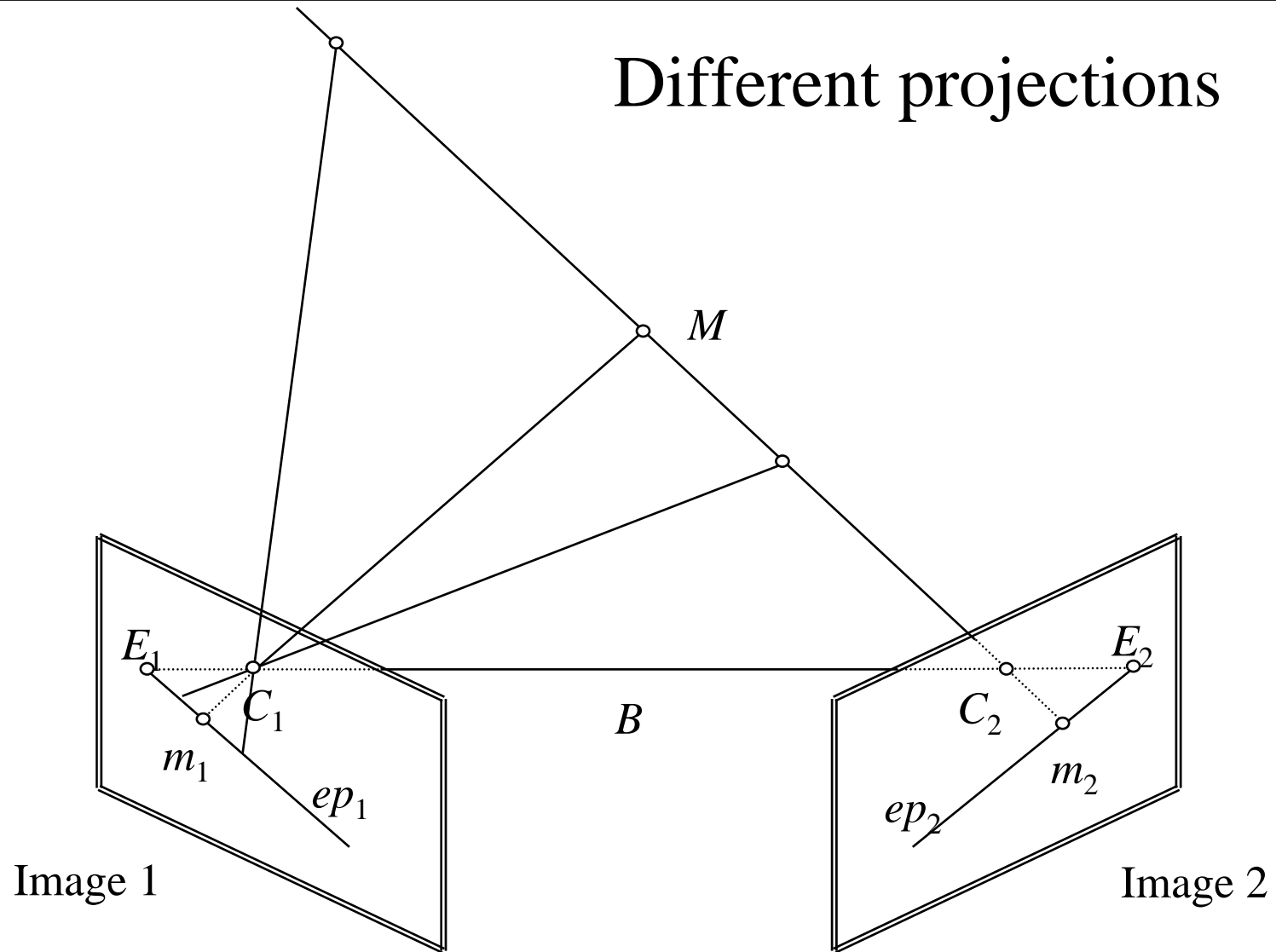
Converging



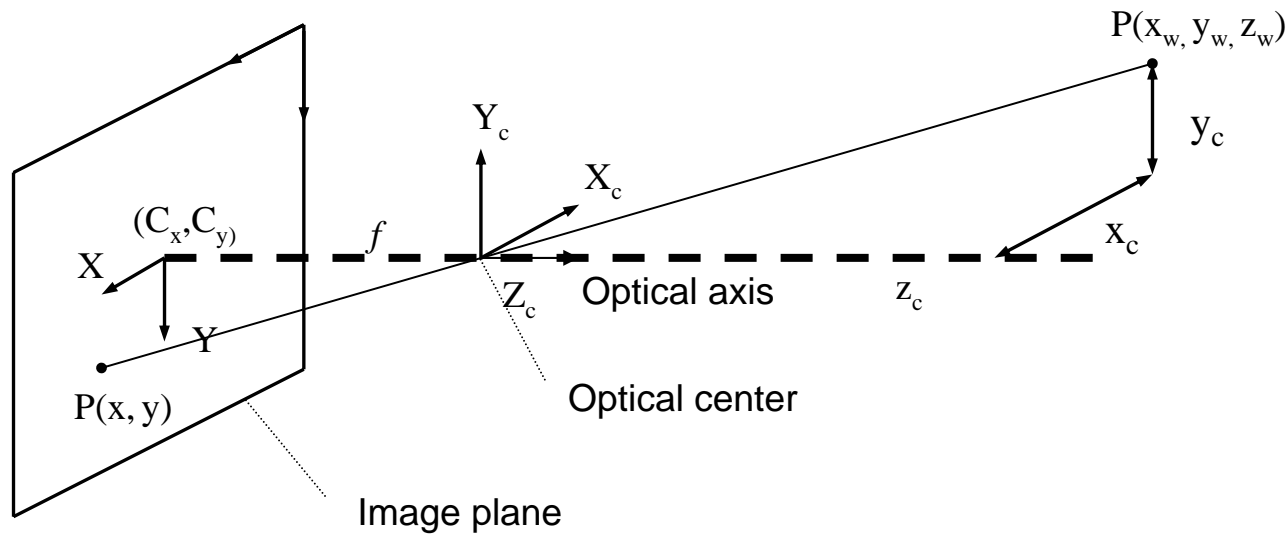
Left camera

Right camera

Different projections



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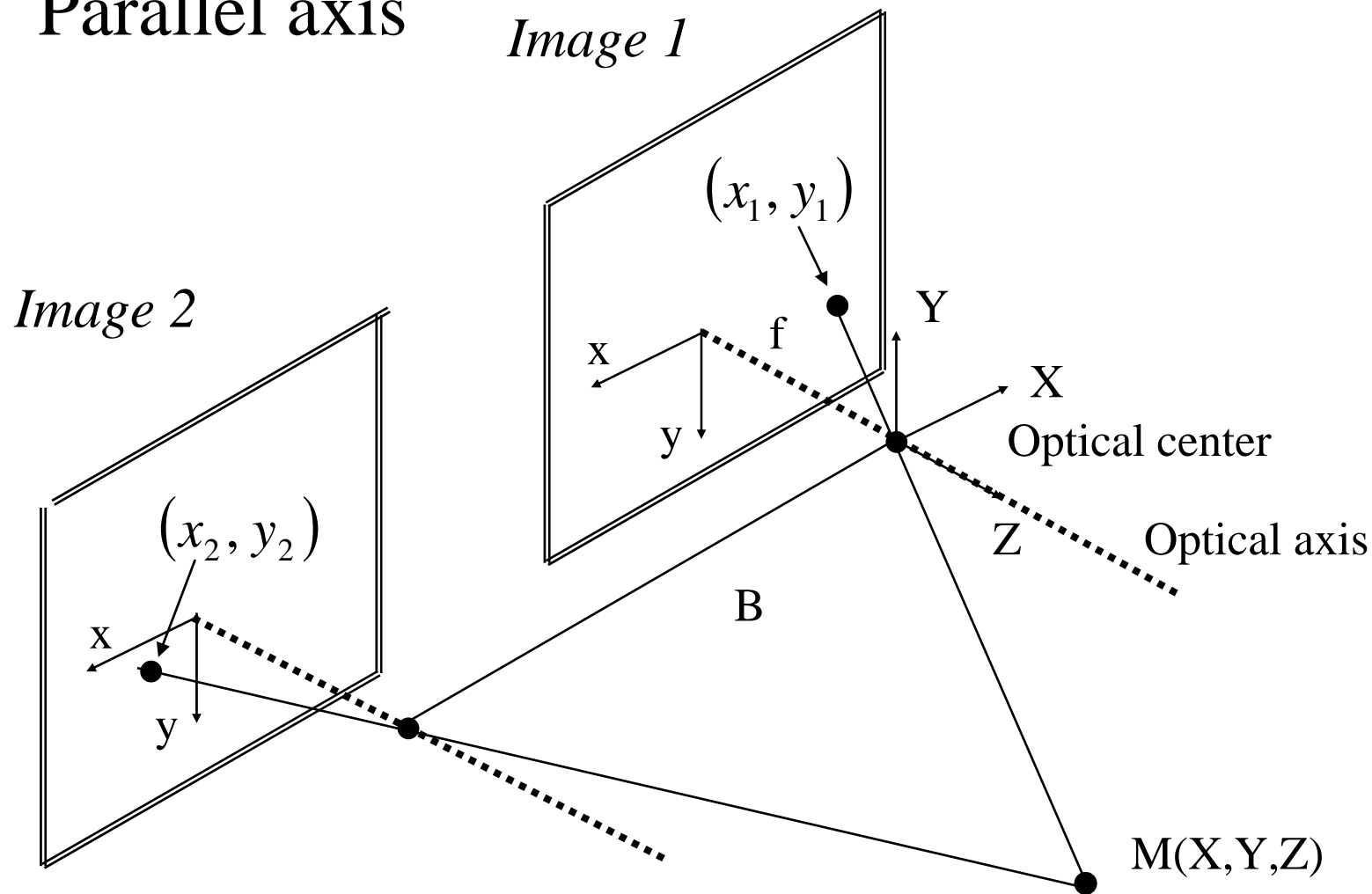
$$\frac{X}{f} = \frac{X_c}{Z_c}$$

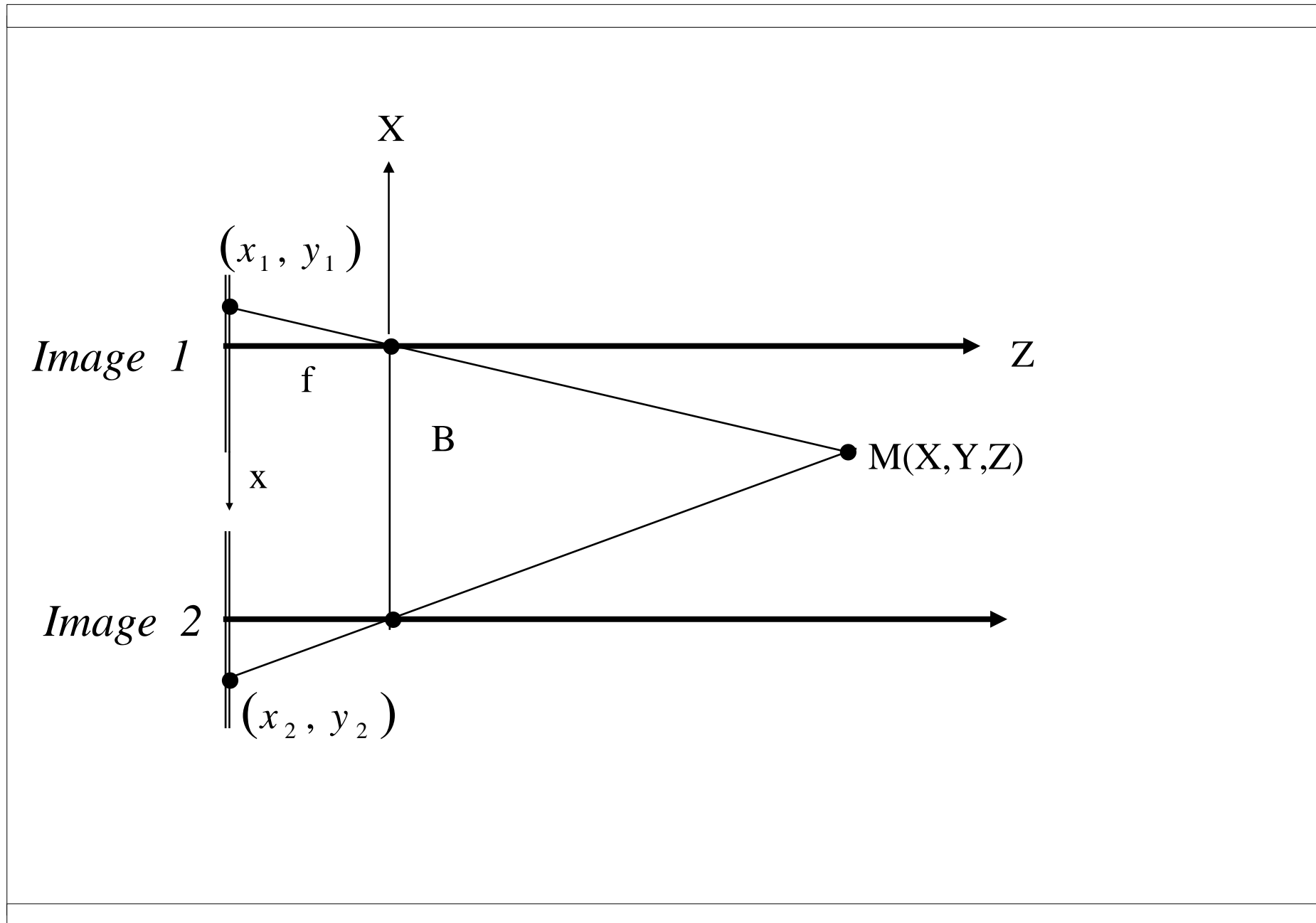
$$\frac{Y}{f} = \frac{Y_c}{Z_c}$$

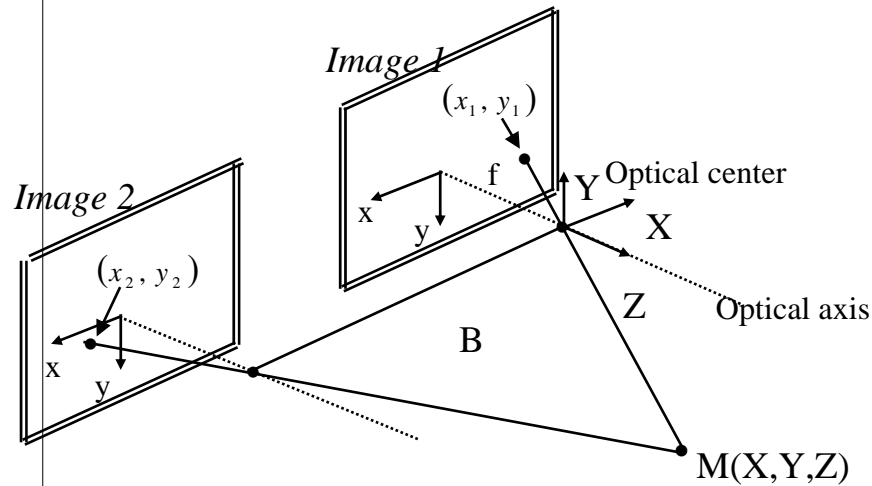
$$\begin{bmatrix} nX \\ nY \\ n \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1/f & 0 \end{bmatrix} \begin{bmatrix} X_c \\ Y_c \\ Z_c \\ 1 \end{bmatrix}$$

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Parallel axis







For the first image

$$\begin{cases} X = \frac{x_1}{f} Z \\ Y = \frac{y_1}{f} Z \end{cases}$$

$$X_2 = X_1 - B$$

For the second image

$$\begin{cases} X = \frac{x_2}{f} Z + B \\ Y = \frac{y_2}{f} Z \end{cases}$$

X should be equal,

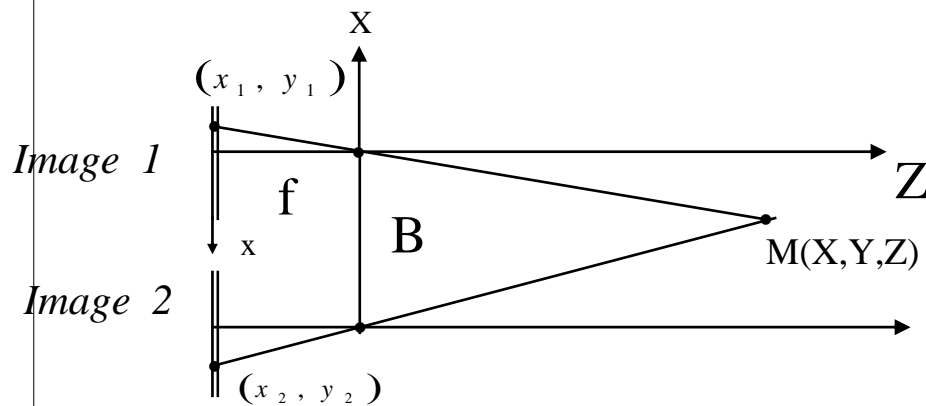
this lets us obtain Z :

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

$$X = \frac{B x_1}{x_1 - x_2}$$

Finally :

$$Y = \frac{B y_1}{x_1 - x_2}$$



- Notice that we are able to obtain the 3D position of point M in space, related to the coordinate system fixed at left camera.
- We can trivially relate this point to a new coordinate system, as long as we know a transformation that relates both systems.

Point coordinates in left camera (image 1)

$$\left(x_f^l, y_f^l\right) \Rightarrow \left(x_u^l, y_u^l\right) \equiv \left(x_1, y_1\right) = m_1$$

Point coordinates in right camera (image 2)

$$\left(x_f^r, y_f^r\right) \Rightarrow \left(x_u^r, y_u^r\right) \equiv \left(x_2, y_2\right) = m_2$$

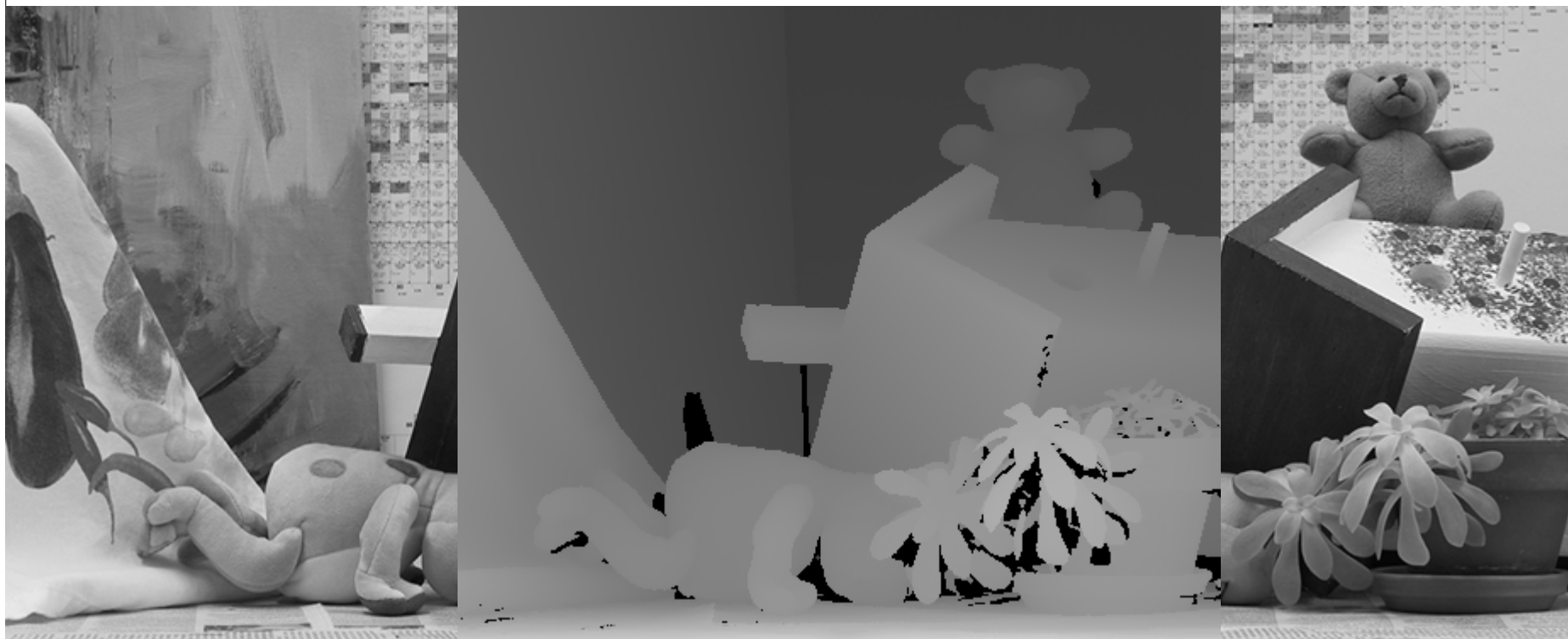
- Parallel axis: $d = x_1 - x_2$ **notice that** $y_1 = y_2$

- d is called disparity:

$$Z = \frac{f B}{d}$$

- Converging axis $d = \left(d_1, d_2\right) = \left(x_f^l - x_f^r, y_f^l - y_f^r\right)$

- There is an inverse relation between Z and d . Fixed B and f : Points with equal disparity possess the same depth Z .
- If B is too big, then it is more difficult to find the correspondence of a point accross images.
- If B is too short, then d won't be very sensitive to changes in Z .



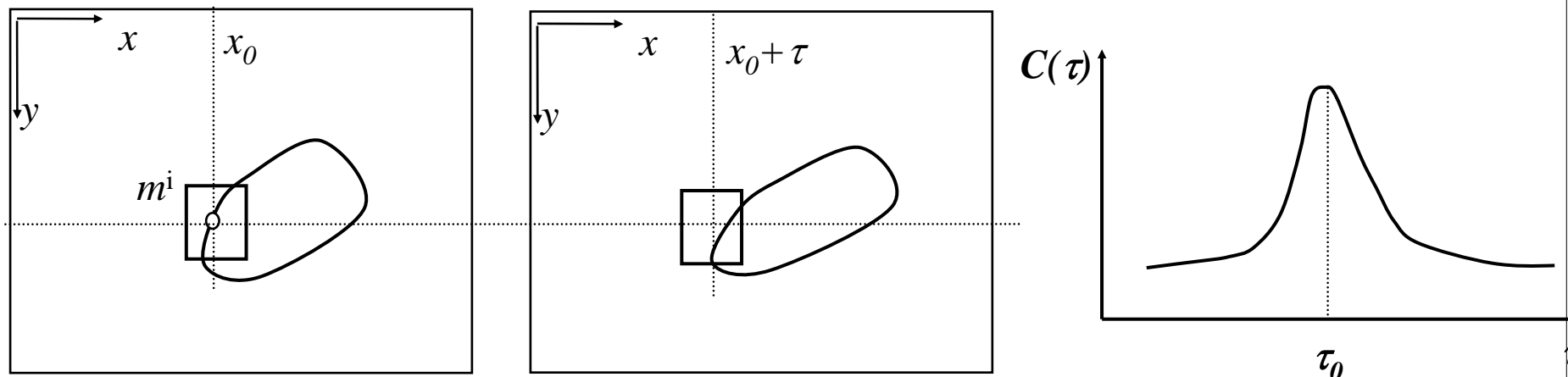
Left eye

Right eye

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- But... how can we find corresponding points in both images?
- We will make the following assumptions:
 - Most scene points are visible from both viewpoints.
 - Corresponding image regions are similar.
- Correspondence methods can be roughly classified in two groups:
 - Correlation-based methods.
 - Feature-based methods.

- Think... How can we look for a similar point in the other image?
- If we admit that corresponding points look similar in both images we can use a correlation measure to find correspondence.

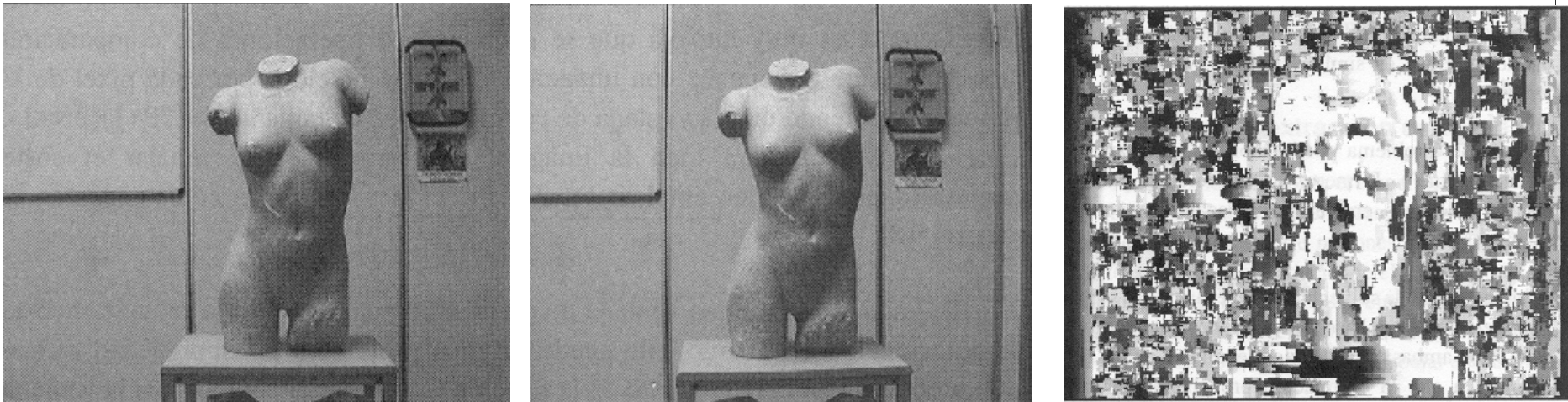


- That is: For each window m^i in the left image we calculate a correlation along the right epipolar line.
- The corresponding point maximizes correlation.

- The function to minimize is:

$$SSD(m^l(x, y), m^r(x', y')) = -\frac{1}{N} \sum_{\forall i, j \perp m(x+i, y+j) \in W(m)} (I^l(x+i, y+j) - I^r(x'+i, y'+j))^2$$

- We obtain a dense disparity map.



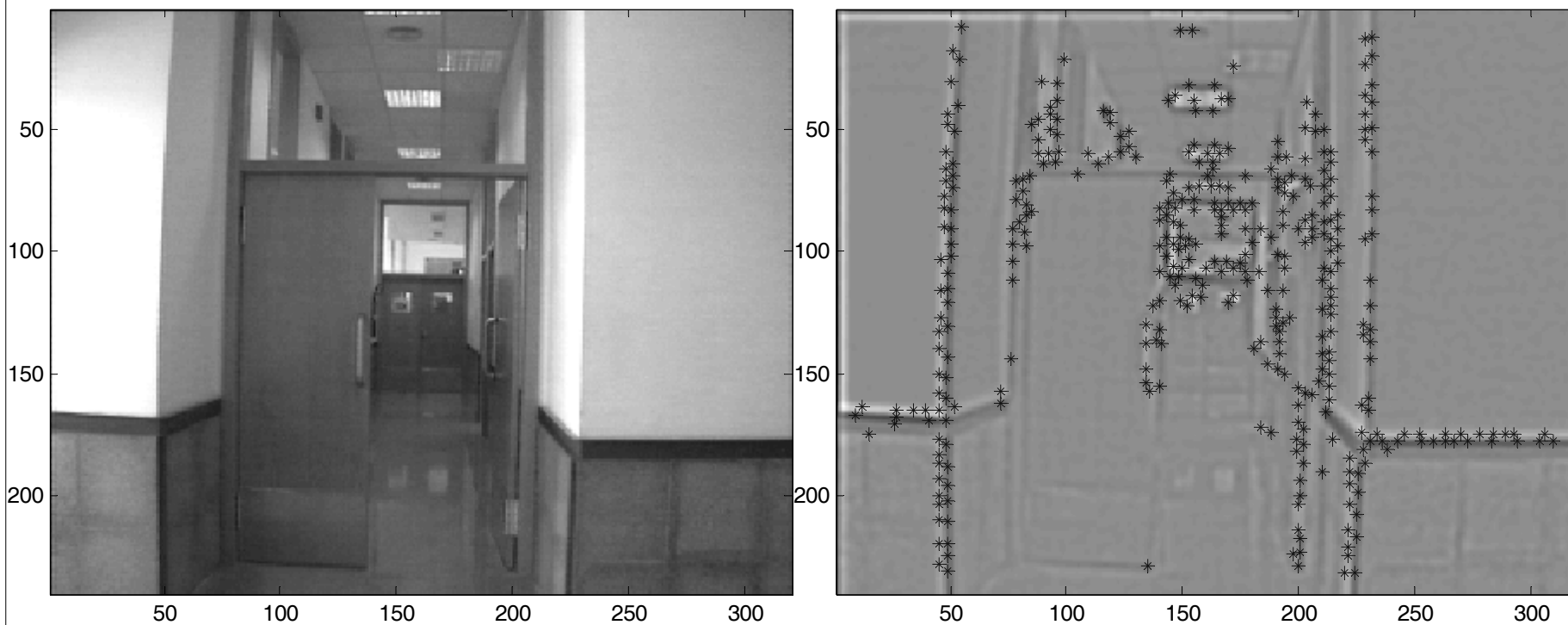
- Problems arising:
 - Illumination differing for both views.
 - Low textured areas can make the method fail.
 - Points can look similar and yet not be correspondant.
 - Occlusions: Points seen in one image are not visible in the other image.

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- Concept:
 - Restrict our search to some very characteristic points.
- First:
 - Extract those characteristic points: Edges, corners... etc, which are of great importance in images.
- Second:
 - Match them accross images.
- Finally:
 - Extract 3D information from correspondences.
- Result: We obtain sparse measures but highly reliable.

- Example:

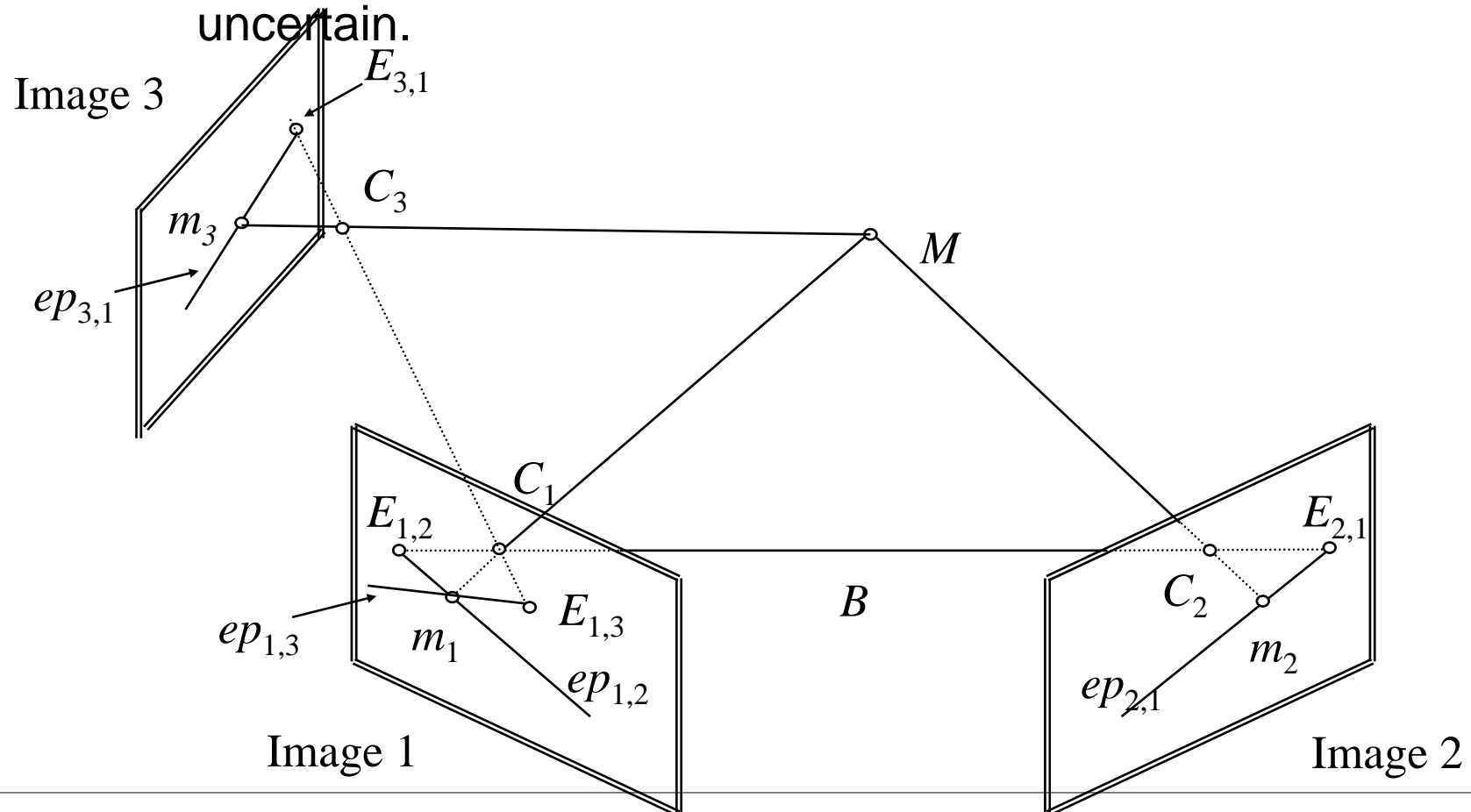
blanco y negro



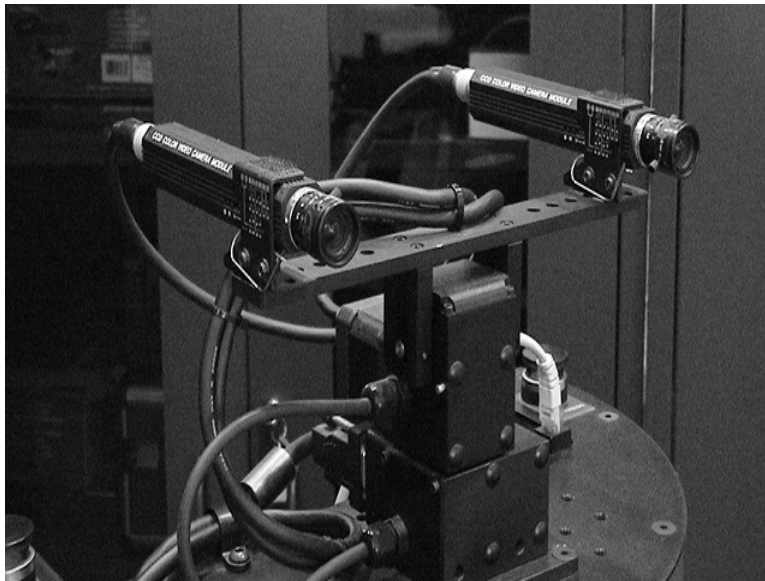
- Example:




- Trinocular vision:
 - Adds more information with an extra camera.
 - It adds up computational cost.
 - It can be useful if correspondence across two images is uncertain.



- Applications: Why?
 - Mobile robots: Try to emulate animal stereo-vision. Navigation in indoor environments.



- Monocular vision!



Real-Time
Camera Tracking
in Unknown Scenes