

Computer Vision and Machine Learning (Visual system and Camera)

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Computer Vision -- Intro

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Image formation

- Basic components
 - Light (Wave length, intensity)
 - Camera (intrinsic and extrinsic parameters)
 - Scene (reflectance, orientation, curvature)

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Shape from X

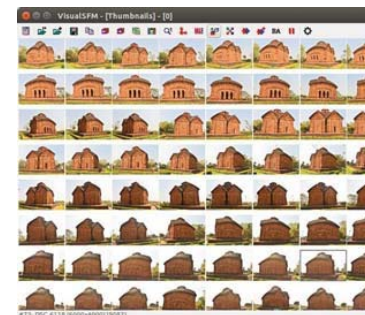
- Reconstructing 3D object from 2D images
 - Stereo
 - Motion
 - Shading
 - Texture
 - Contour

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Geometry: 3D reconstruction (cond.)



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Geometry: 3D reconstruction



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Human Visual system and Camera

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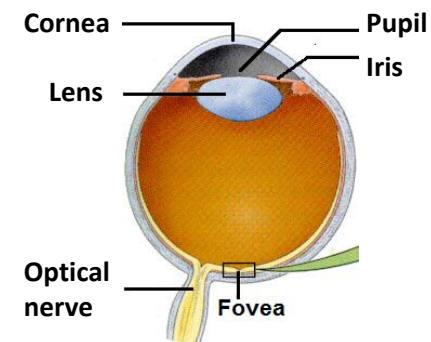
Human vision

- Most important and powerful sensor in human body.
- The **image formation** process is well understood.
- The **image understanding** is the one that remains **mysterious**.

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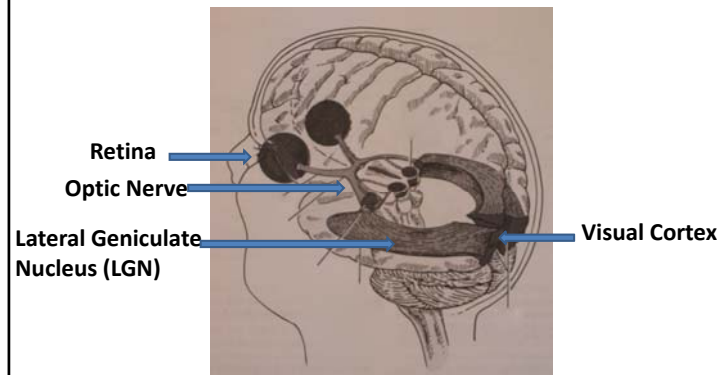
Human eye structure



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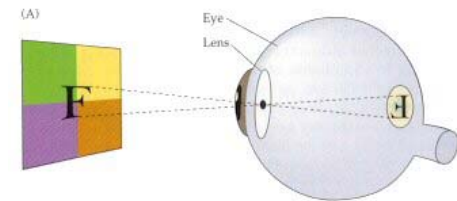
Pathways to the brain



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IMAGE FORMATION

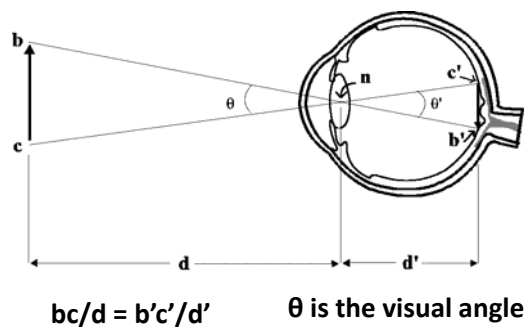


Camera model for eye

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PERSPECTIVE PROJECTION



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Light sensors in the eye

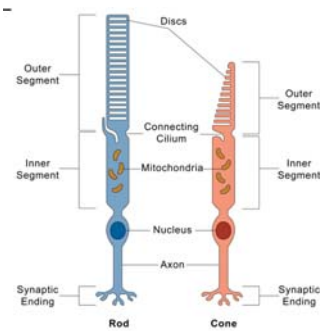
- Image formed on the retina is sensed by special type of cells called *photoreceptors*.
- Photoreceptors convert physical light signal into biological signal that passes through optical nerves.
- There are two types of photoreceptors:
 - Cones
 - Rods

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Rod and cone cells in eye

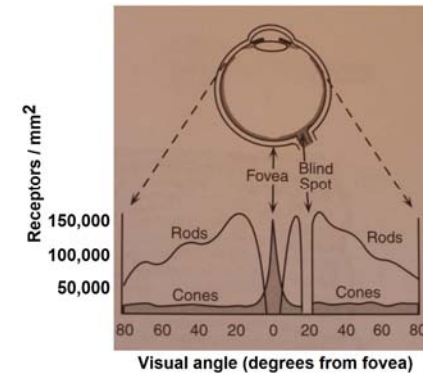


- **Cones:** Color-vision with higher intensity light.
- **Rods:** Low-intensity light vision, e.g. night vision.
- Cones are larger in size and are more densely packed in fovea region.
- Rods are smaller in size, much higher number in periphery.

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Receptor cell distribution

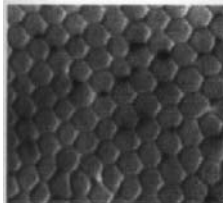


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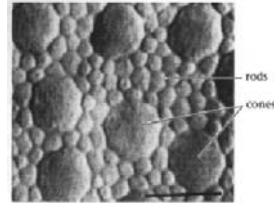
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Cell packing at fovea and periphery

Receptor cells at fovea



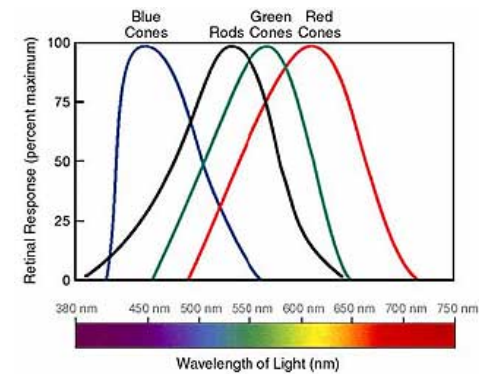
Receptor cells at periphery



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Color sensitivity of receptors



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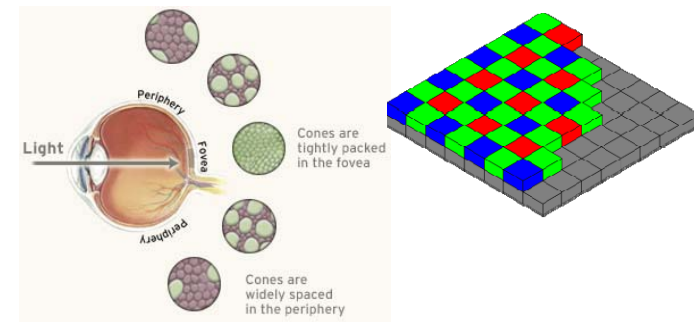
Colour sensitivity

- Wavelength of visible spectrum ranges from 0.38 μm to 0.76 μm (approx.)
- Response characteristics $h_B(v)$ for **blue** attains maximum at about 6.8×10^4 Hz or **0.44 μm**
- Response characteristics $h_G(v)$ for **green** attains maximum at about 5.8×10^4 Hz or **0.52 μm**
- Response characteristics $h_R(v)$ for **red** attains maximum at about 4.3×10^4 Hz or **0.70 μm**

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Retina and camera sensor



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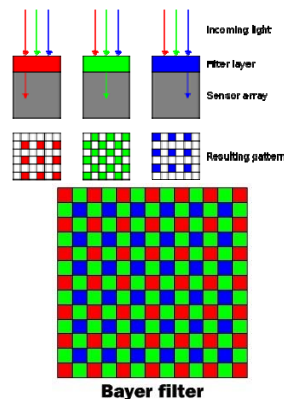
Bayer filter

- Bayer filter is a color filter array used with sensor of digital camera.

– Arranged in square grid with 50% G, 25% R and 25% B.

– Each pixel gets one particular color that the filter allows to reach the sensor.

– Other color components are interpolated (averaged) from the neighborhood.



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Human eye: Summary

Frontal part

- Cornea
- Pupil and Iris
- Lens

Retina:

- Rods (low-intensity light, night vision)
- Cones (color-vision)
- Synapses and ganglions
- Optic nerve fibers

Does sensing and low-level processing.

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Human vision vs. Computer Vision

The camera replaces the eye:

- Eye lens → Camera Optics
- Cones and Rods → CCD array
- Ganglion cells → Filter banks

The computer replaces the brain:

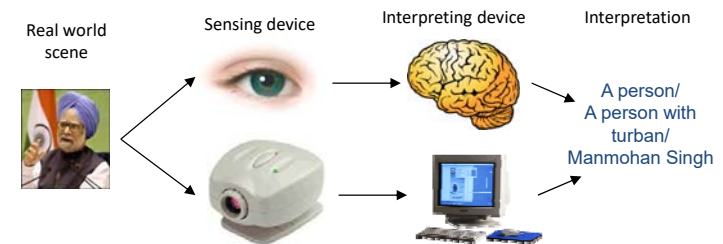
- But how?

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The Objective

- Want to make a computer understand images
- We know it is possible – we do it effortlessly!



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Camera model



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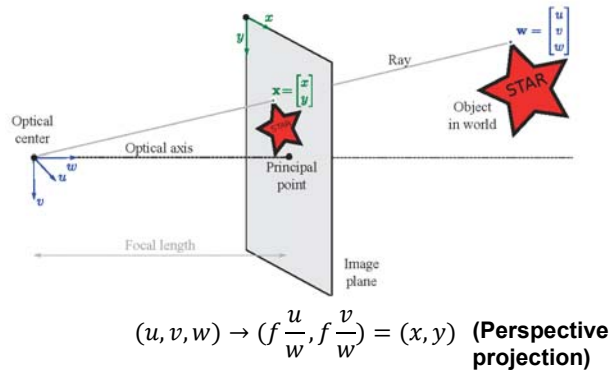
Projection properties

- **Many-to-one**: any point along same *visual ray* map to same point in image
- **Point → point**
- **Line → line** (collinearity is preserved)
 - But line through focal point (visual ray) projects to a point
- **Plane → plane** (or half-planes)
 - But plane through focal point projects to line

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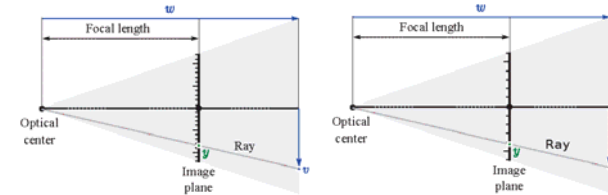
Image and scene coordinate



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Photoreceptor density

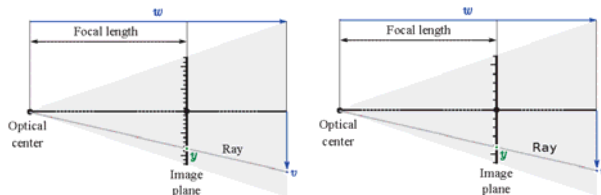


- We measure x and y (image coordinates) in terms of number of pixels, so should be for depth (w) also.
- With same image size, depth **value** varies with photoreceptor density.

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Photoreceptor density



- Hence, photoreceptor density should be combined with focal length.
- x unit of length $\equiv x \varphi_x$ number of pixels
- y unit of length $\equiv y \varphi_y$ number of pixels

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Shift of origin

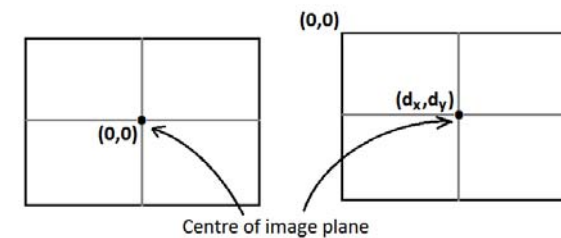


Image size in mm or inch: $(x, y) = (f \frac{u}{w}, f \frac{v}{w})$

With shift of origin: $(x, y) = (f \frac{u}{w} + S_x, f \frac{v}{w} + S_y)$

where $d_x = \varphi_x S_x$, $d_y = \varphi_y S_y$

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Photoreceptor density and shift

$$(u, v, w) \rightarrow (f \frac{u}{w}, f \frac{v}{w}) = (x, y)$$

turns to $(u, v, w) \rightarrow (f \frac{\varphi_x u}{w}, f \frac{\varphi_y v}{w}) = (x, y)$

Instead of origin (0,0) at the centre of image it usually at top-left corner, i.e., principal point is at (d_x, d_y) . Thus

$$x = f \frac{\varphi_x u}{w} + d_x$$

$$y = f \frac{\varphi_y v}{w} + d_y$$

$(f, \varphi_x, \varphi_y, d_x, d_y)$ are **intrinsic parameters** of the camera.

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Projection matrix for camera

Projection is a matrix multiplication using homogeneous coordinates:

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1/f & 0 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix} = \begin{bmatrix} u \\ v \\ w/f \end{bmatrix} \Rightarrow (f \frac{u}{w}, f \frac{v}{w})$$

Considering photoreceptor density and shift:

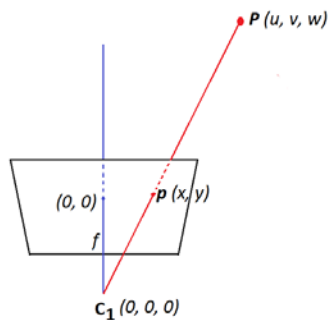
$$\begin{bmatrix} \varphi_x & 0 & d_x \\ 0 & \varphi_y & d_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1/f & 0 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix} = \begin{bmatrix} u\varphi_x + wd_x/f \\ v\varphi_y + wd_y/f \\ w/f \end{bmatrix}$$

$$\Rightarrow (f \frac{\varphi_x u}{w} + d_x, f \frac{\varphi_y v}{w} + d_y)$$

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Depth from disparity

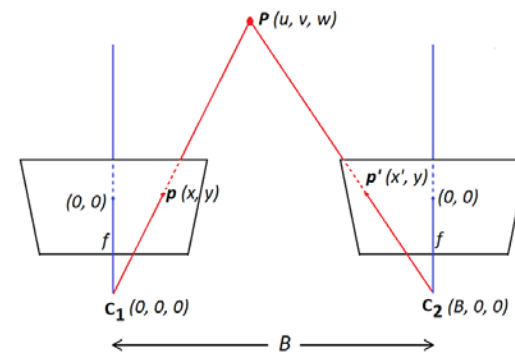


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Depth from disparity

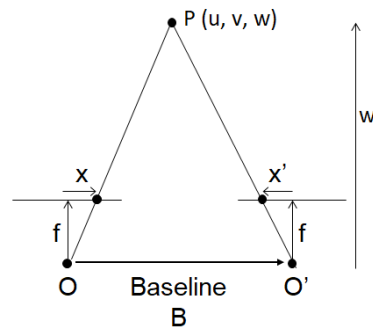


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Depth from disparity



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Depth from disparity

- Camera-2 is shifted from camera-1 only along horizontal direction and no rotation.
- Camera-1:
 - Optical centre $O_1: (0,0,0)$
 - World point $P: (u, v, w)$
 - Image coordinate: $x = f \frac{u}{w} \Rightarrow \frac{x}{f} = \frac{u}{w}$ (1)
- Camera-2:
 - Optical centre $O_2: (B, 0, 0)$
 - World point $P: (u - B, v, w)$
 - Image coordinate: $x' = f \frac{u-B}{w} \Rightarrow \frac{x'}{f} = \frac{u-B}{w}$ (2)

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Depth from disparity (contd.)

- Subtracting equation (2) from equation (1):

$$\frac{x - x'}{f} = \frac{B}{w}$$

- This implies

$$w = \frac{B \cdot f}{x - x'}$$

- Once we know w , u can be easily determined as

$$u = w \frac{x}{f}$$

Similarly y .

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Position and orientation of camera

- Camera may be placed anywhere in a scene.
- Same object may appear differently while captured by same camera at different position and orientation.
- It would be easier to handle if the relation between their geometry is known.

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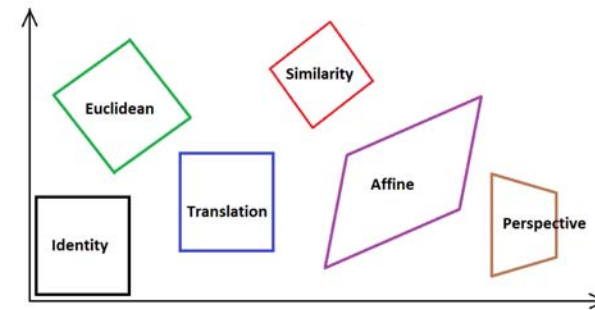
Coordinate transformation

- World coordinate system
 - World (object) centric scene coordinates
- Camera coordinate system
 - Camera centric scene coordinates
- Transforming world (scene) coordinate to camera coordinate
 - Rotation, translation and (optional) scaling
 - Linear transformation is advantageous.

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2D transforms



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2D transforms

Rotation: $\begin{bmatrix} u' \\ v' \end{bmatrix} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} \equiv \mathbf{p}' = \mathbf{R}\mathbf{p}$

(Coordinate axis is rotated anti-clockwise by an angle θ .)

Scaling: $\begin{bmatrix} u' \\ v' \end{bmatrix} = \begin{bmatrix} S_u & 0 \\ 0 & S_v \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} \equiv \mathbf{p}' = \mathbf{S}\mathbf{p}$

(Coordinate axis is squeezed / expanded $1/S_u$ and $1/S_v$.)

Translation: $\begin{bmatrix} u' \\ v' \end{bmatrix} = \begin{bmatrix} u \\ v \end{bmatrix} + \begin{bmatrix} t_u \\ t_v \end{bmatrix} \quad \mathbf{p}' = \mathbf{p} + \mathbf{t}$

(Coordinate axis is translated by $-t_u$ and $-t_v$.)

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3D transforms

Rotation: $R_u = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \alpha & \sin \alpha \\ 0 & -\sin \alpha & \cos \alpha \end{bmatrix}$

$R_v = \begin{bmatrix} \cos \beta & 0 & -\sin \beta \\ 0 & 1 & 0 \\ \sin \beta & 0 & \cos \beta \end{bmatrix} \quad R_w = \begin{bmatrix} \cos \gamma & \sin \gamma & 0 \\ -\sin \gamma & \cos \gamma & 0 \\ 0 & 0 & 1 \end{bmatrix}$

$R = R_u R_v R_w \equiv \mathbf{p}' = \mathbf{R}\mathbf{p}$, where $R = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$

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3D transforms

Scaling: $\begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} = \begin{bmatrix} s_u & 0 & 0 \\ 0 & s_v & 0 \\ 0 & 0 & s_w \end{bmatrix} \begin{bmatrix} u \\ v \\ w \end{bmatrix} \equiv \mathbf{p}' = \mathbf{S}\mathbf{p}$

Translation: $\begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} = \begin{bmatrix} u \\ v \\ w \end{bmatrix} + \begin{bmatrix} t_u \\ t_v \\ t_w \end{bmatrix} \equiv \mathbf{p}' = \mathbf{p} + \mathbf{t}$

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3D translation

Translation:

$$\begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} = \begin{bmatrix} u + t_u \\ v + t_v \\ w + t_w \end{bmatrix} = \begin{bmatrix} u \\ v \\ w \end{bmatrix} + \begin{bmatrix} t_u \\ t_v \\ t_w \end{bmatrix} \equiv \mathbf{p}' = \mathbf{p} + \mathbf{t}$$

In homogeneous coordinate system

$$\begin{bmatrix} u' \\ v' \\ w' \\ 1 \end{bmatrix} = \begin{bmatrix} u + t_u \\ v + t_v \\ w + t_w \\ 1 + 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & t_u \\ 0 & 1 & 0 & t_v \\ 0 & 0 & 1 & t_w \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix} \equiv \mathbf{p}'_h = \mathbf{T}\mathbf{p}_h$$

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Projection matrix for camera

- World to camera co-ordinate system using homogeneous coordinates:

$$\begin{bmatrix} u^c \\ v^c \\ w^c \\ 1 \end{bmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_u \\ r_{32} & r_{22} & r_{23} & t_v \\ r_{31} & r_{32} & r_{33} & t_w \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix}$$

$$\mathbf{p}^c = \mathbf{T}\mathbf{p} = \begin{bmatrix} \mathbf{R} & \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix} \mathbf{p}$$

- The elements of the matrix \mathbf{R} , i.e., $r_{i,j}$ and t_k are **extrinsic parameters**.

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Projection matrix for image formation

In practice: lots of coordinate transformations...

$$\begin{bmatrix} \text{2D point} \\ (3 \times 1) \end{bmatrix} = \begin{bmatrix} \text{Camera to pixel coord.} \\ \text{trans. matrix} \\ (3 \times 3) \end{bmatrix} \begin{bmatrix} \text{Perspective} \\ \text{projection matrix} \\ (3 \times 4) \end{bmatrix} \begin{bmatrix} \text{World to camera coord.} \\ \text{trans. matrix} \\ (4 \times 4) \end{bmatrix} \begin{bmatrix} \text{3D point} \\ (4 \times 1) \end{bmatrix}$$

$$\begin{bmatrix} u^c \varphi_x + w^c d_x / f \\ v^c \varphi_y + w^c d_y / f \\ w^c / f \end{bmatrix} = \begin{bmatrix} \varphi_x & 0 & d_x \\ 0 & \varphi_y & d_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1/f & 0 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_u \\ r_{32} & r_{22} & r_{23} & t_v \\ r_{31} & r_{32} & r_{33} & t_w \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} u^c \varphi_x + w^c d_x / f \\ v^c \varphi_y + w^c d_y / f \\ w^c / f \end{bmatrix} = \begin{bmatrix} \varphi_x & 0 & d_x / f & 0 \\ 0 & \varphi_y & d_y / f & 0 \\ 0 & 0 & 1/f & 0 \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} & r_{13} & t_u \\ r_{32} & r_{22} & r_{23} & t_v \\ r_{31} & r_{32} & r_{33} & t_w \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \\ 1 \end{bmatrix}$$

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Thank you!

Any question?