

Lecture 5.1

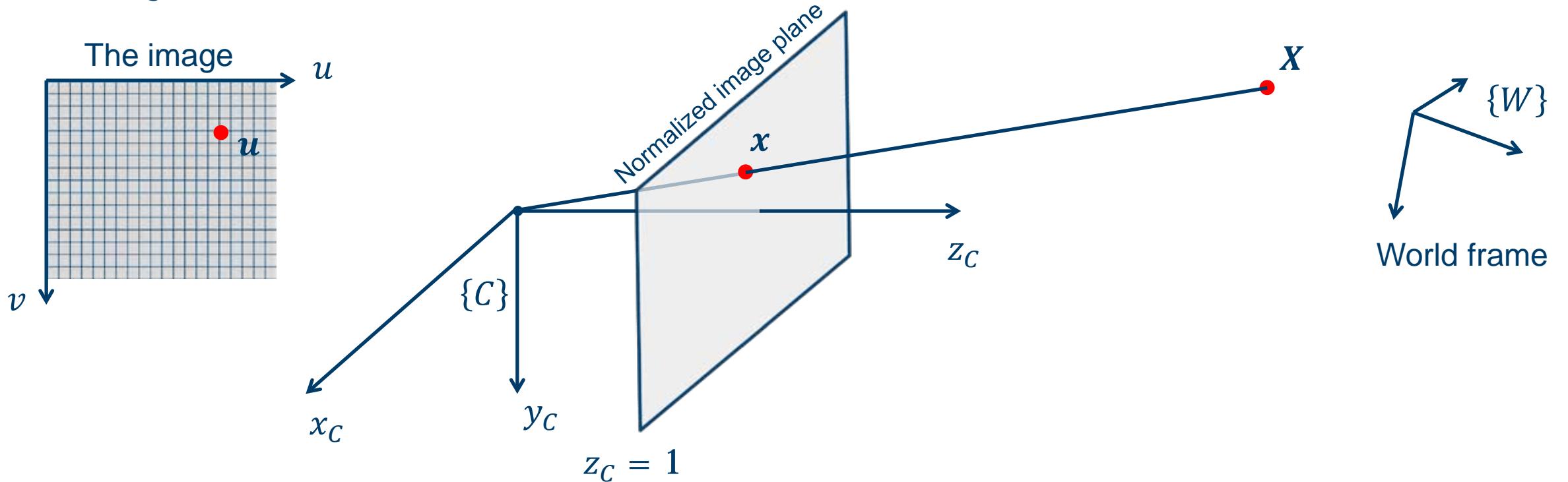
The camera matrix P

Thomas Opsahl

Weekly overview – Single View Geometry

- **The camera matrix P**
 - Repetition on perspective cameras
 - Dissecting the general camera matrix P
 - Vanishing points
 - Vanishing lines
 - Estimating P from known world-image correspondences
- **Camera calibration**
 - Zhangs method briefly
 - Calibration in Matlab
 - Undistortion
- **Pose from known 3D points**
 - Decompose P to K, R and t
 - PnP

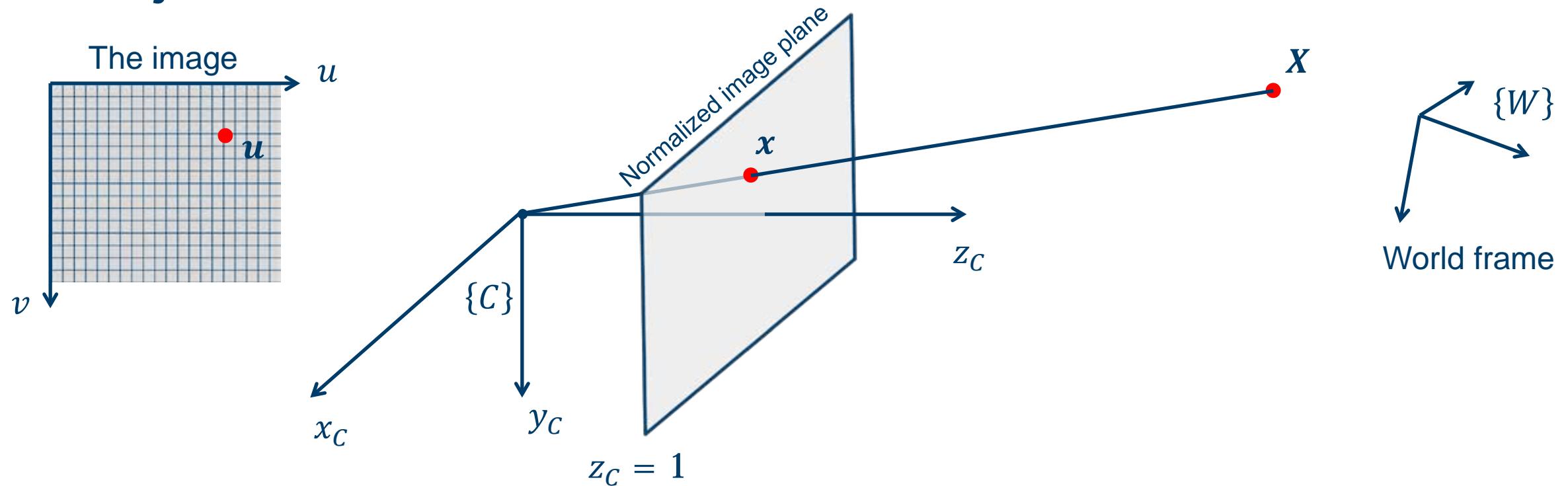
Projective cameras



- The perspective camera model describes the correspondence between points in the world and points in the image

$$\tilde{\mathbf{u}} = P^W \tilde{\mathbf{X}}$$

Projective cameras



Pinhole camera

$$P = \begin{bmatrix} f & 0 & 0 \\ 0 & f & 0 \\ 0 & 0 & 1 \end{bmatrix} [R \quad t]$$

Finite projective camera

$$P = \begin{bmatrix} f_u & s & c_u \\ 0 & f_v & c_v \\ 0 & 0 & 1 \end{bmatrix} [R \quad t]$$

General projective camera

$$P = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix}$$

with rank = 3 and 11 dof

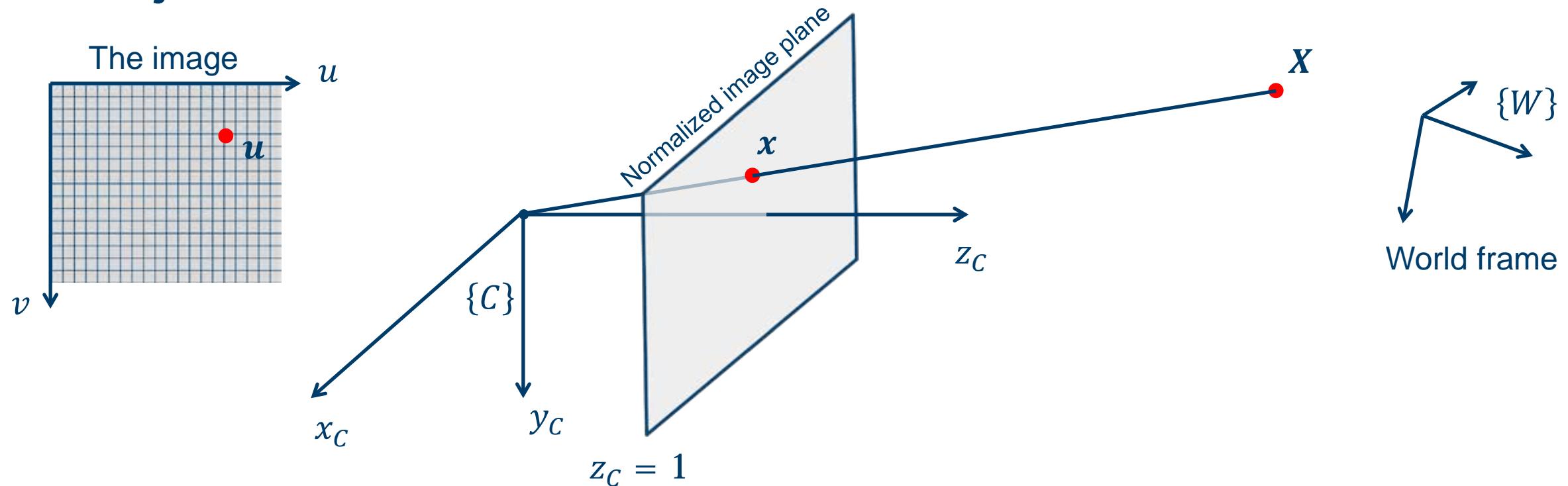
The camera matrix P

- From our lecture on the perspective camera model, we know that the camera matrix contains geometrical information about the image, the camera-frame and the world-frame
- For finite cameras, $P = K[R \ t]$ we recall that
 - The camera calibration matrix K describes the relationship between normalized image coordinates and image coordinates
 - The rotation matrix R describes how the world-frame $\{W\}$ is rotated/oriented relative to the camera-frame $\{C\}$ and is a change-of-basis-matrix for vectors (${}^C R_W : \{W\} \rightarrow \{C\}$)

$$\begin{aligned} \hat{\mathbf{e}}^C w_1 & \quad {}^C w_2 & \quad {}^C w_3 \hat{\mathbf{u}} = {}^C R_W = \begin{pmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{pmatrix} \hat{\mathbf{u}} = \begin{matrix} \hat{\mathbf{e}}^W c_1^T \\ \hat{\mathbf{e}}^W c_2^T \\ \hat{\mathbf{e}}^W c_3^T \end{matrix} \hat{\mathbf{u}} \end{aligned}$$

- The vector t describes how the world-frame is translated/positioned relative to the camera-frame
- The matrix ${}^C \xi_W = [R \ t]$ represents the pose of the world-frame relative to the camera-frame and transforms points from $\{W\}$ to $\{C\}$

Projective cameras



- What geometric information can we extract from a general camera matrix?

$$P = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix}$$

Dissecting the camera matrix

- First some practical notations
- Rows and columns

$$P = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix} = [\mathbf{p}_1 \quad \mathbf{p}_2 \quad \mathbf{p}_3 \quad \mathbf{p}_4] = \begin{bmatrix} \mathbf{P}^{1T} \\ \mathbf{P}^{2T} \\ \mathbf{P}^{3T} \end{bmatrix}$$

- Left 3×3 matrix in P

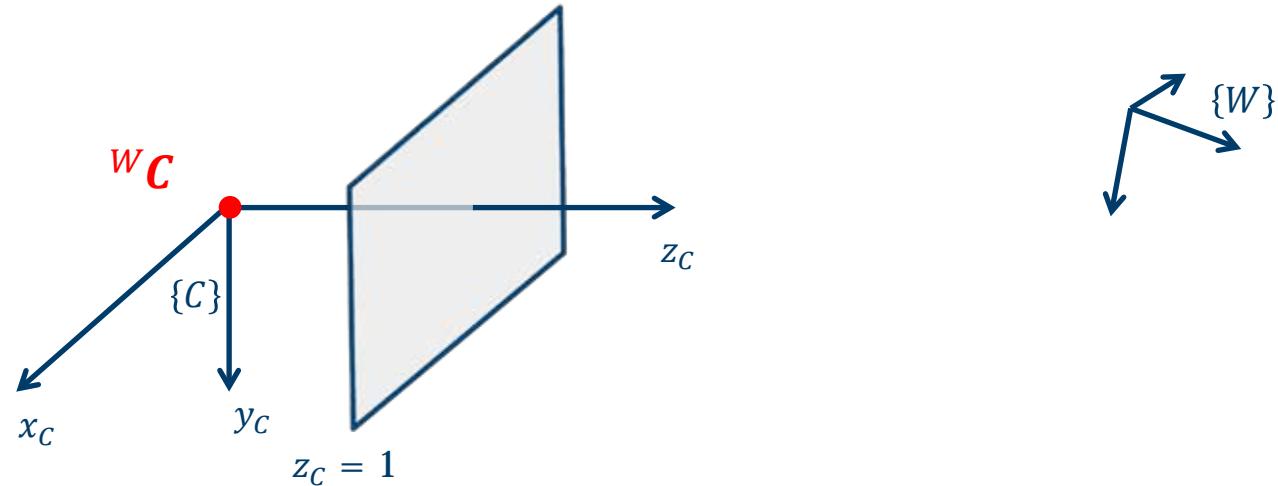
$$M = \begin{bmatrix} p_{11} & p_{12} & p_{13} \\ p_{21} & p_{22} & p_{23} \\ p_{31} & p_{32} & p_{33} \end{bmatrix} \Rightarrow P = [M \quad \mathbf{p}_4]$$

- $\det(M) \neq 0 \Rightarrow$ Finite projective camera
- $\det(M) = 0 \Rightarrow$ Camera center at infinity

NOTE

In the following slides $\mathbf{p}_1, \mathbf{p}_2, \mathbf{p}_3, \mathbf{p}_4, \mathbf{P}^1, \mathbf{P}^2$ and \mathbf{P}^3 are homogeneous vectors, $\mathbf{p}_i \in \mathbb{P}^2$ while $\mathbf{P}^j \in \mathbb{P}^3$

Dissecting the camera matrix

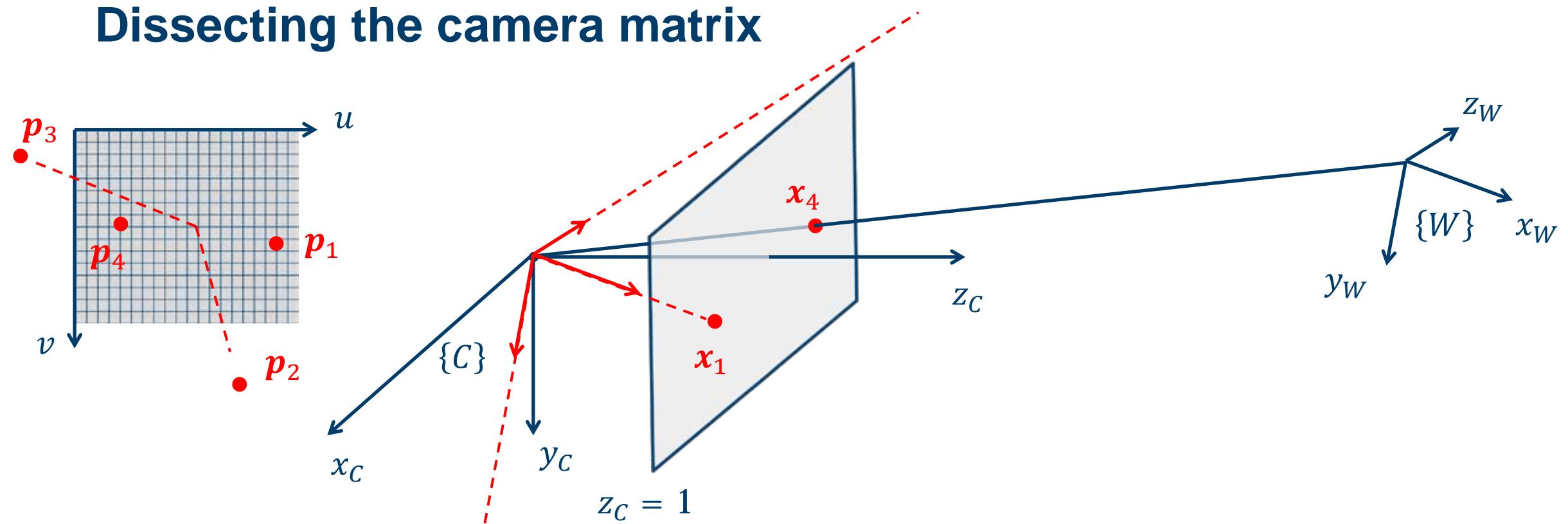


- The **camera center** ${}^W\mathbf{C}$ is given by the 1-dimensional right null-space of P
 - Finite camera
 - Camera at infinity

$${}^W\tilde{\mathbf{C}} = \begin{bmatrix} -M^{-1}\mathbf{p}_4 \\ 1 \end{bmatrix} \Rightarrow {}^W\mathbf{C} = -M^{-1}\mathbf{p}_4$$

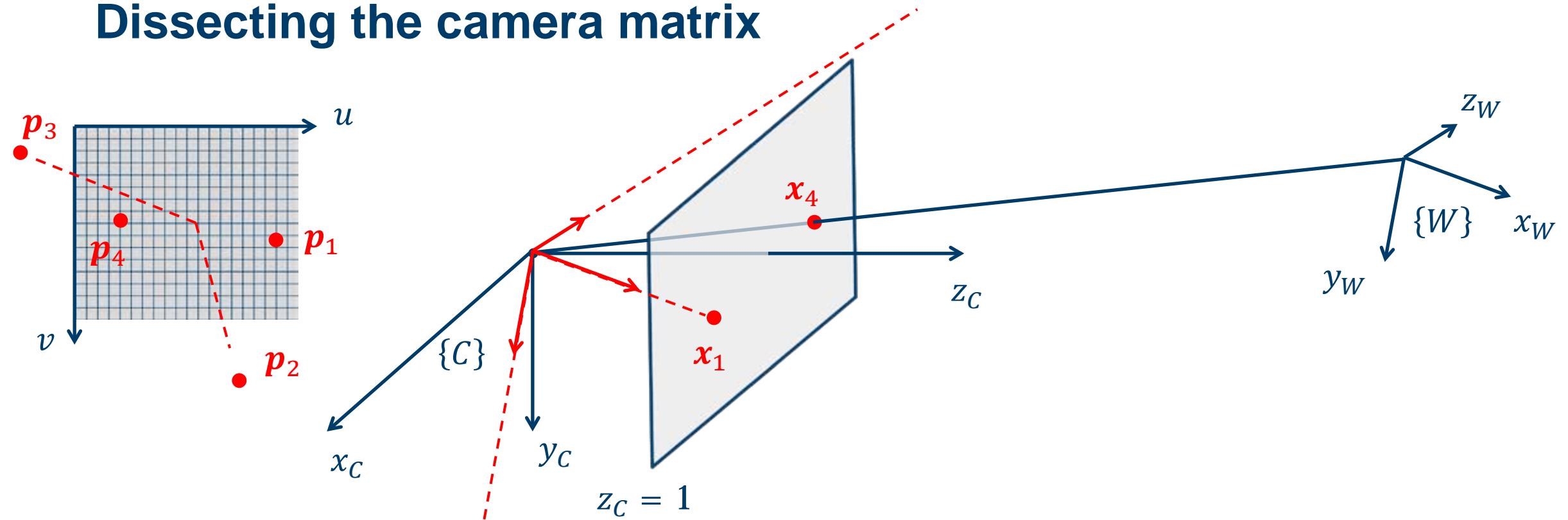
$${}^W\tilde{\mathbf{C}} = \begin{bmatrix} \mathbf{d} \\ 0 \end{bmatrix} \text{ where } \mathbf{d} \in \text{null}(M), \text{ i.e. } M\mathbf{d} = 0$$

Dissecting the camera matrix



- The **column vectors** p_1 , p_2 and p_3 are the vanishing points in the image corresponding to the X -, Y - and Z -axis of $\{W\}$ respectively
 - If a coordinate axis is parallel to the xy -plane of $\{C\}$, the corresponding column vector will be an infinite point
- The **column vector** p_4 is the image of the origin of $\{W\}$

Dissecting the camera matrix



- Observe

Points at infinity:
Homogeneous representation
of the directions of the three
coordinate axes of $\{W\}$

$$p_1 = P \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad p_2 = P \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad p_3 = P \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \quad p_4 = P \begin{pmatrix} 0 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$$

Finite point:
Homogeneous representation
of $\{W\}$'s origin

Source: Flickr.com
Photographer: Melita

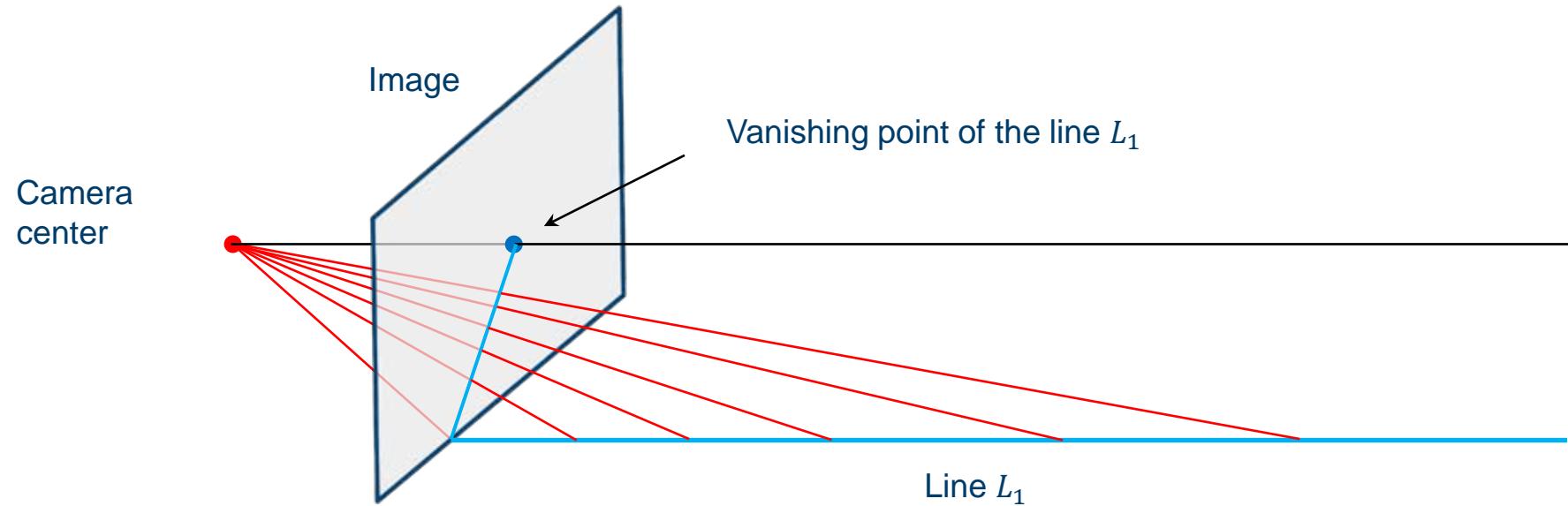


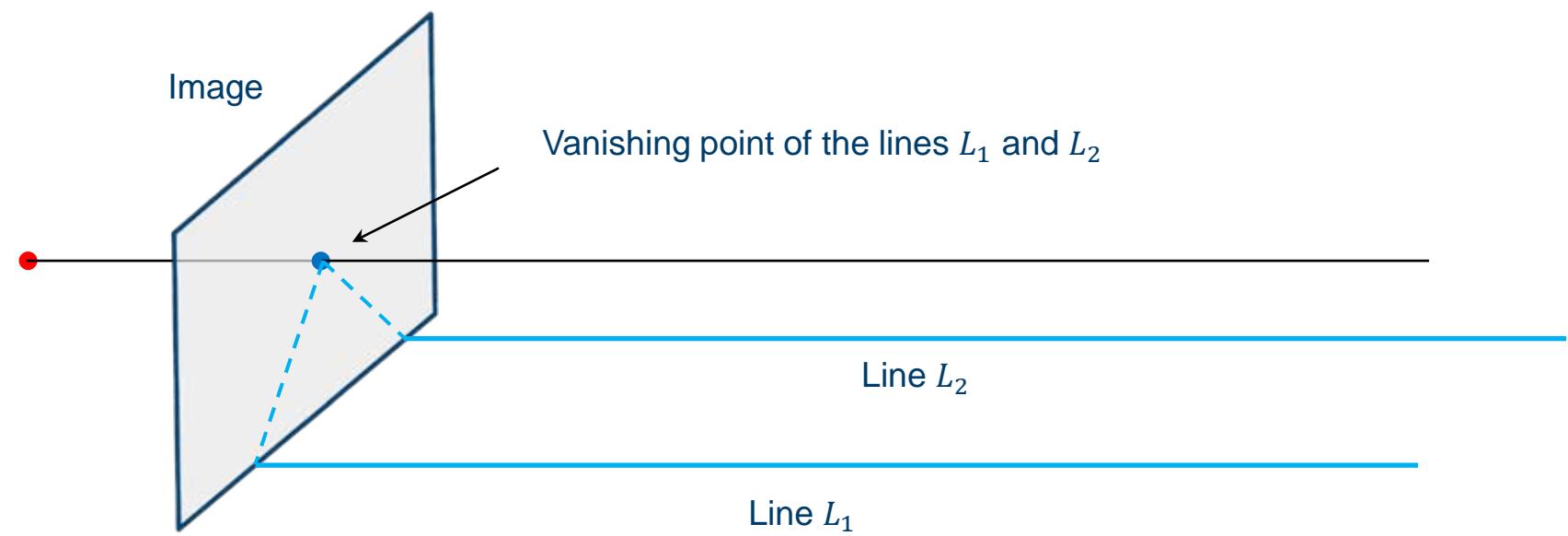
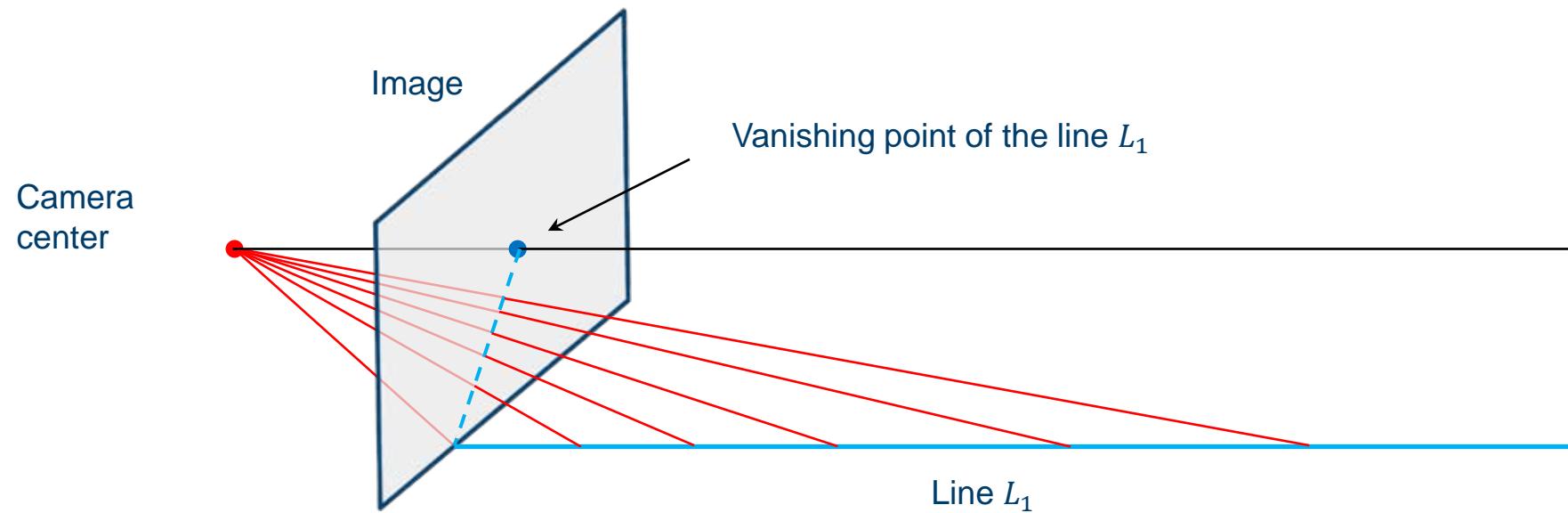


Parallel lines in 3D
have the same
vanishing point on the
line at infinity

The vanishing line
for the ground plane







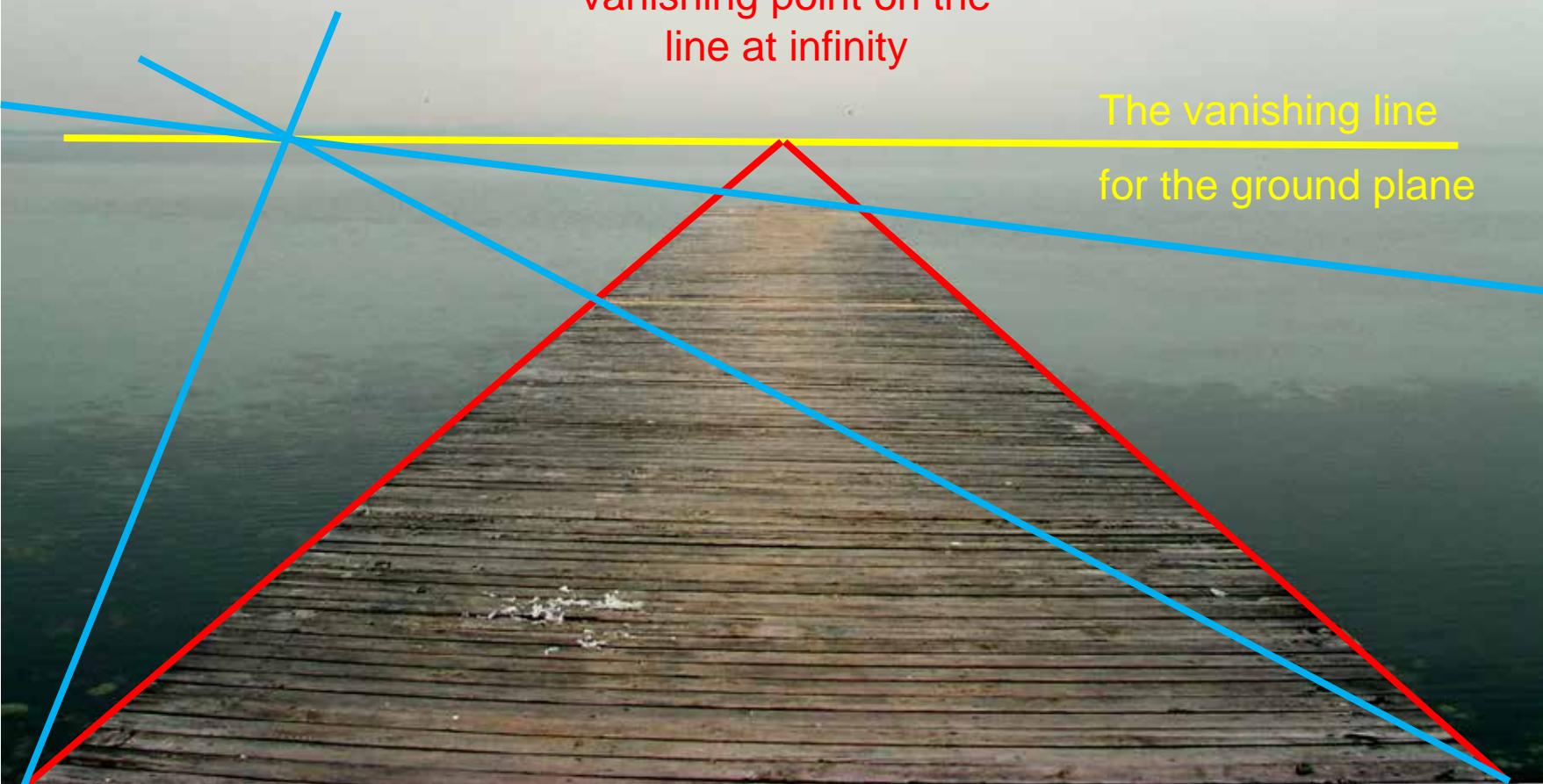
Any two parallel lines in 3D
have the same vanishing point

The line through the camera
center and the vanishing point
is parallel to the lines

Another family of parallel lines in the same plane has a different vanishing point on the same line

Parallel lines in 3D have the same vanishing point on the line at infinity

The vanishing line for the ground plane

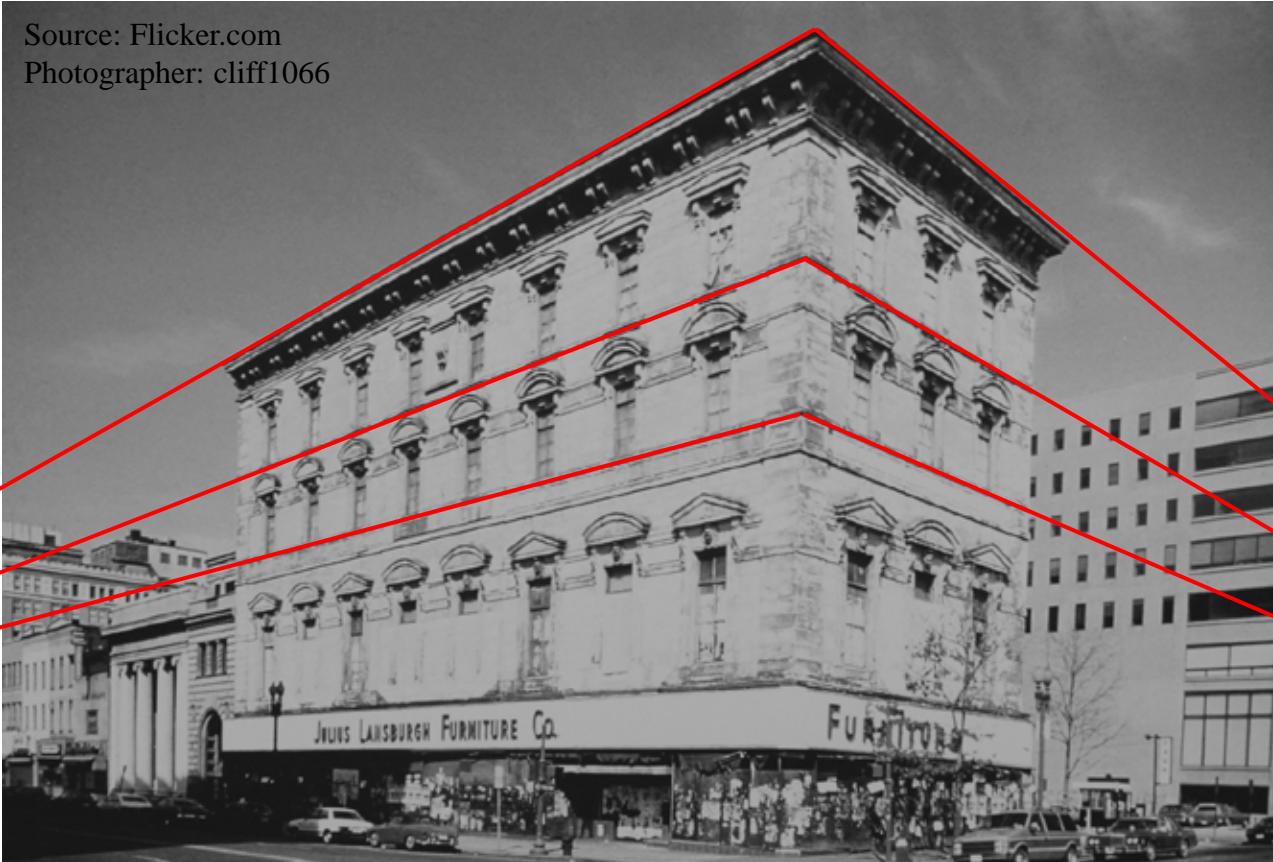


Source: Flickr.com
Photographer: cliff1066

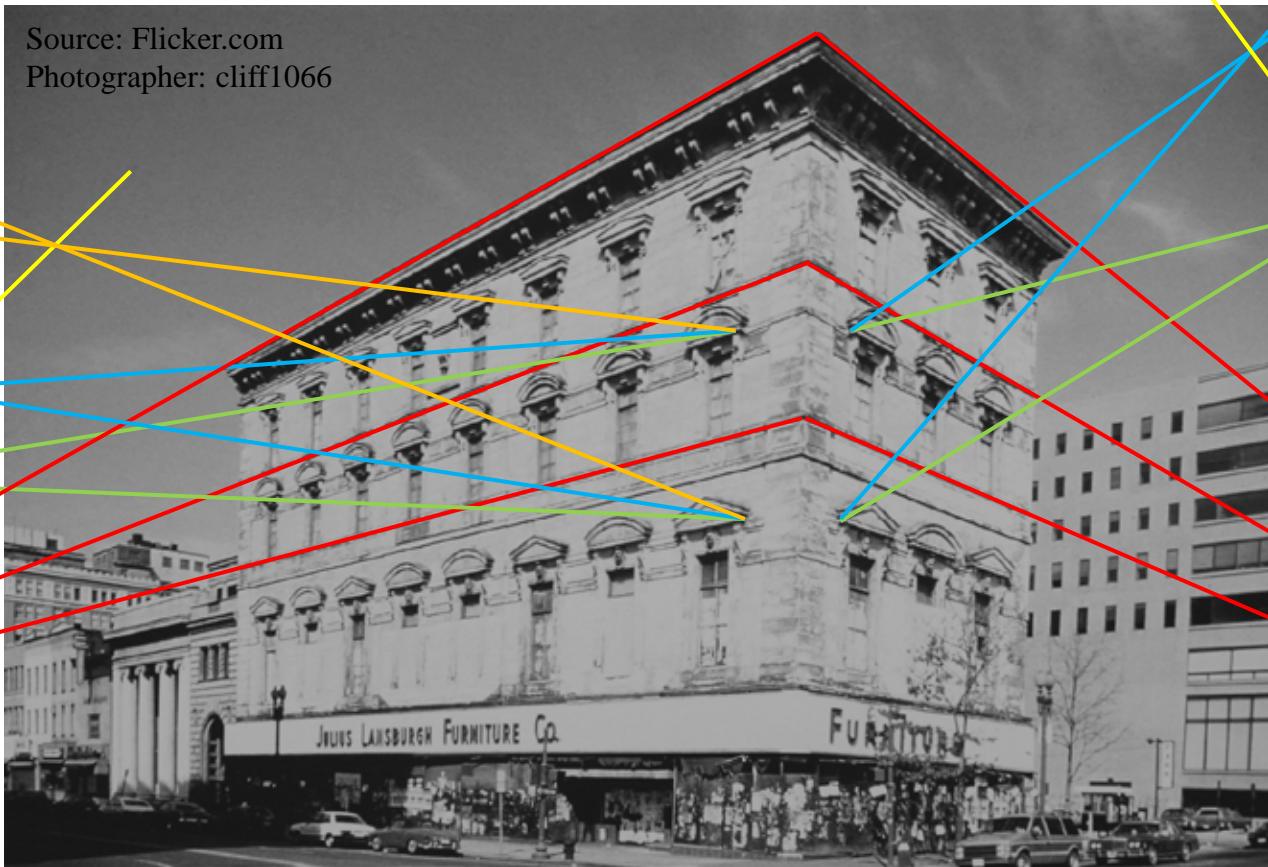


- Vanishing points and vanishing lines are often not directly visible, but can be estimated from lines known to be parallel in the image

Source: Flickr.com
Photographer: cliff1066



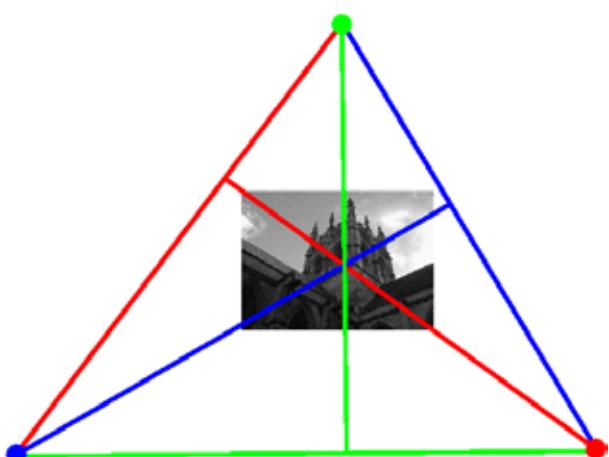
- Vanishing points and vanishing lines are often not directly visible, but their position in the image plane can be estimated from lines known to be parallel in the image



- Vanishing points and vanishing lines are often not directly visible, but their position in the image plane can be estimated from lines known to be parallel in the image
- These quantities are useful in several ways

Camera calibration

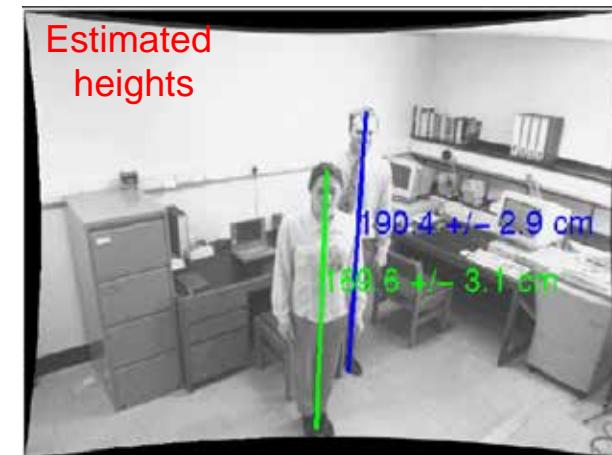
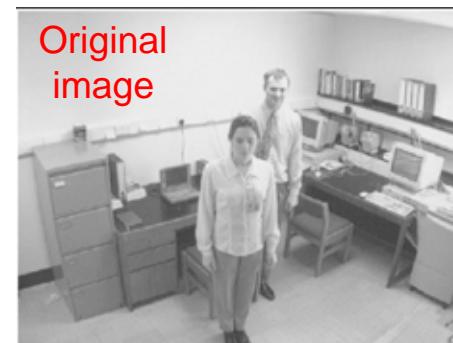
For instance it is possible to calibrate a camera from only three orthogonal vanishing points (if we also assume zero skew and square pixels)



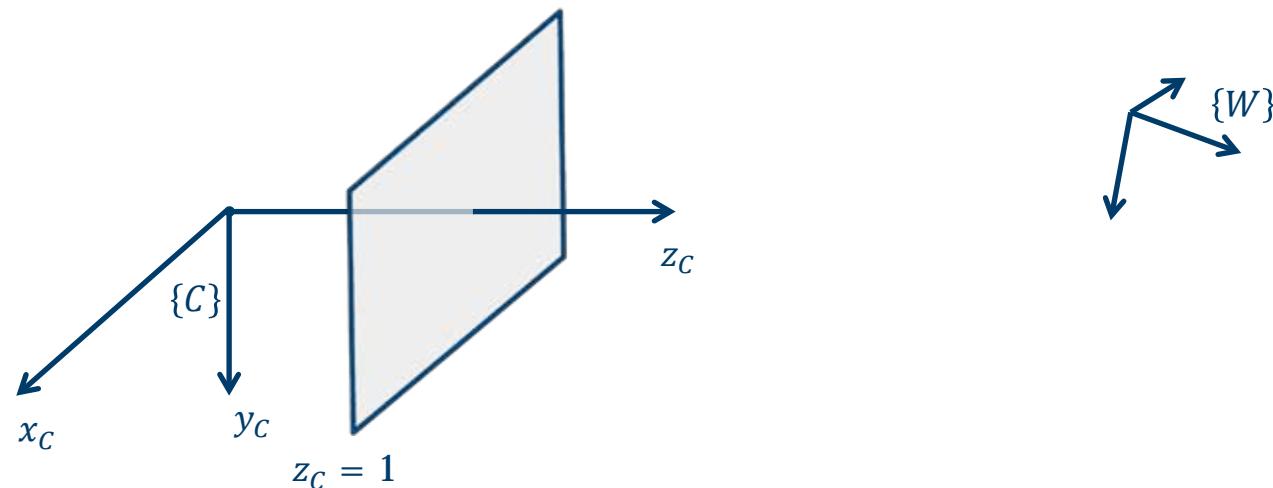
$$K = \begin{bmatrix} 1163 & 0 & 548 \\ 0 & 1163 & 404 \\ 0 & 0 & 1 \end{bmatrix}$$

Measurements from a single image

Knowing the vanishing line of a plane, its vertical vanishing point and the size of vertical line segment, it is possible to determine the size of other vertical segments

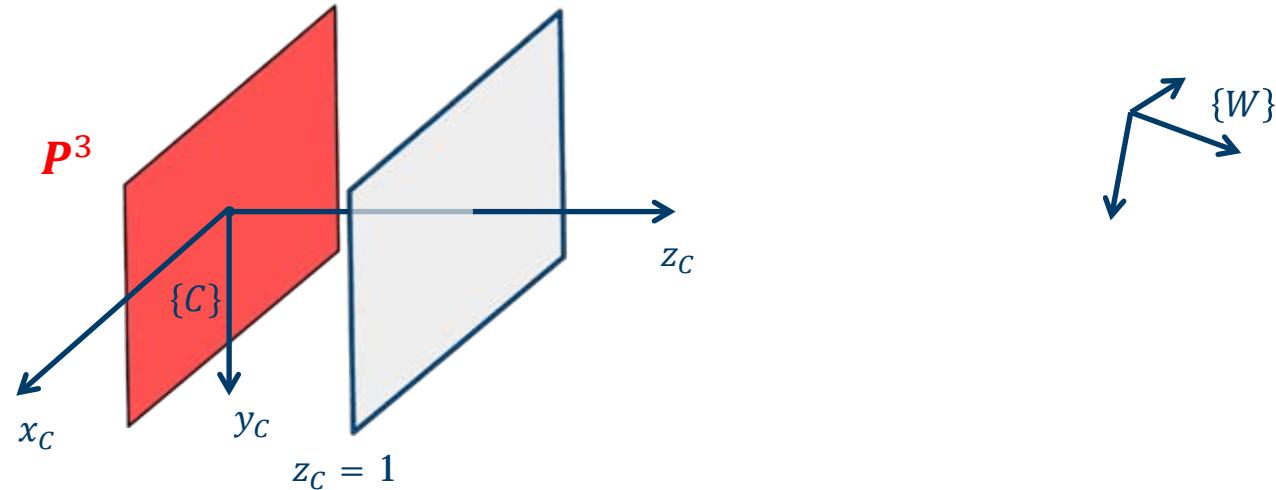


Dissecting the camera matrix



- The **row vectors** P^1 , P^2 and P^3 are homogeneous 4-vectors that can be interpreted as planes in $\{W\}$
 - Recall that in \mathbb{P}^3 points and planes are dual objects in the same way that points and lines are dual in \mathbb{P}^2

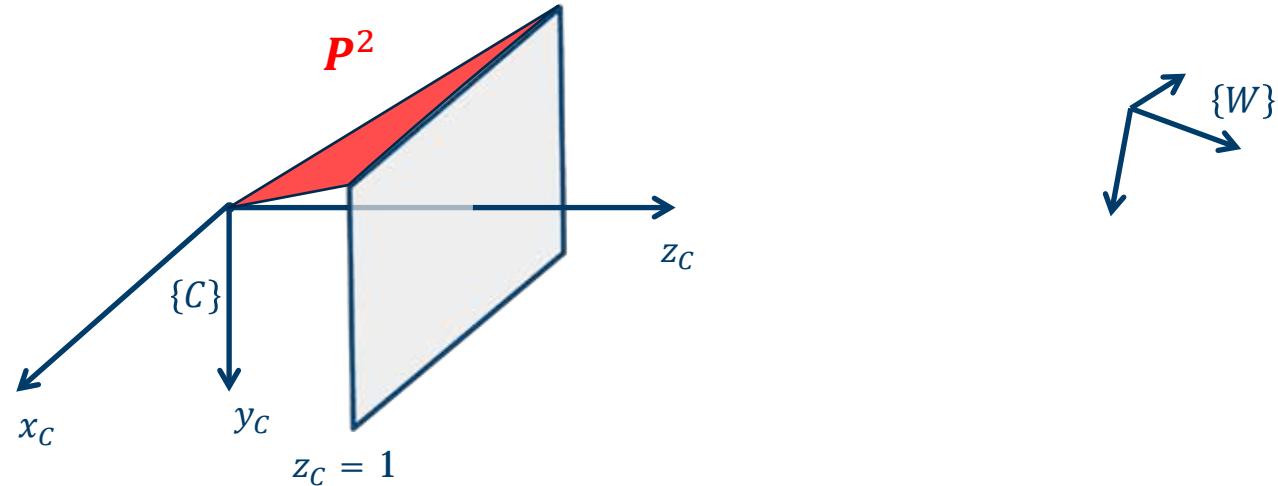
Dissecting the camera matrix



- The **row vectors** P^1 , P^2 and P^3 are homogeneous 4-vectors that can be interpreted as planes in $\{W\}$
 - Recall that in \mathbb{P}^3 points and planes are dual objects in the same way that points and lines are dual in \mathbb{P}^2
- P^3 represents the xy -plane of $\{C\}$ in $\{W\}$ coordinates

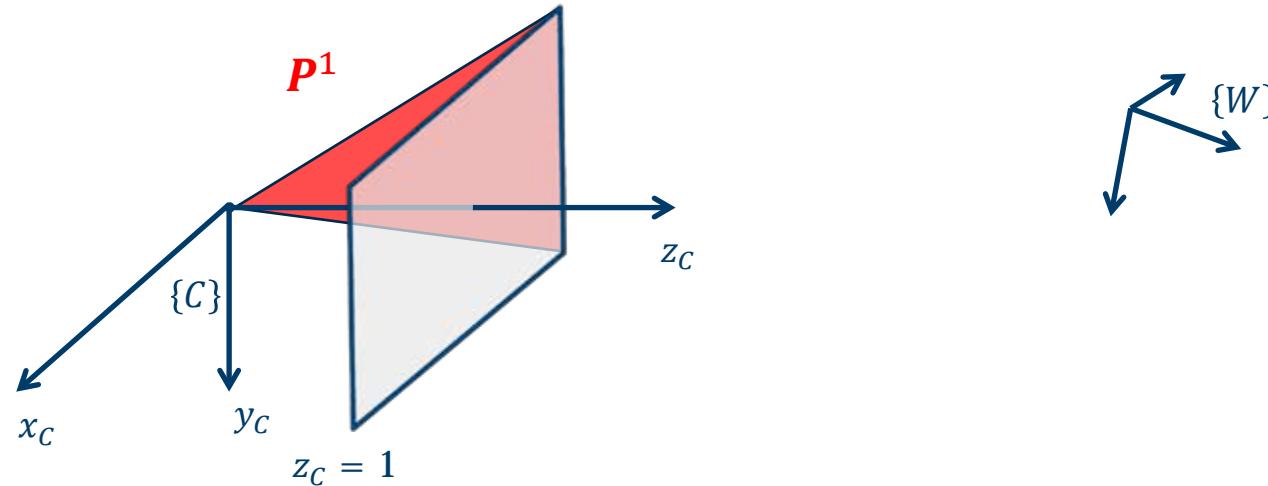
Observe that the xy -plane of $\{C\}$ consists of points ${}^W\tilde{X}$ that are imaged on to the line at infinity of the image, i.e. $P^W\tilde{X} = [u, v, 0]^T$ which is equivalent with the homogeneous plane equation $P^3{}^W\tilde{X} = 0$

Dissecting the camera matrix



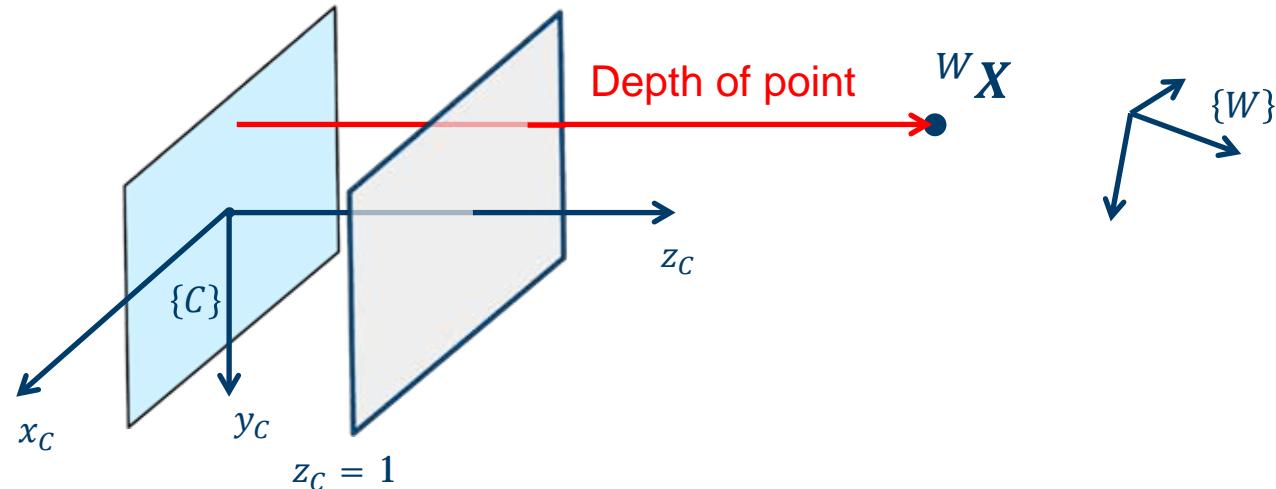
- The **row vectors** P^1 , P^2 and P^3 are homogeneous 4-vectors that can be interpreted as planes in $\{W\}$
 - Recall that in \mathbb{P}^3 points and planes are dual objects in the same way that points and lines are dual in \mathbb{P}^2
- P^3 represents the xy -plane of $\{C\}$ in $\{W\}$ coordinates
- P^2 represents the plane in $\{W\}$ that projects onto the u -axis of the image
Observe that the points ${}^W\bar{X}$ that maps to the u -axis in the image must satisfy $P^W\bar{X} = [u, 0, w]^T$ which is equivalent with the homogeneous plane equation $P^2{}^W\bar{X} = 0$

Dissecting the camera matrix



- The **row vectors** P^1 , P^2 and P^3 are homogeneous 4-vectors that can be interpreted as planes in $\{W\}$
 - Recall that in \mathbb{P}^3 points and planes are dual objects in the same way that points and lines are dual in \mathbb{P}^2
- P^3 represents the xy -plane of $\{C\}$ in $\{W\}$ coordinates
- P^2 represents the plane in $\{W\}$ that projects onto the u -axis of the image
- P^1 represents the plane in $\{W\}$ that projects onto the v -axis of the image

Dissecting the camera matrix



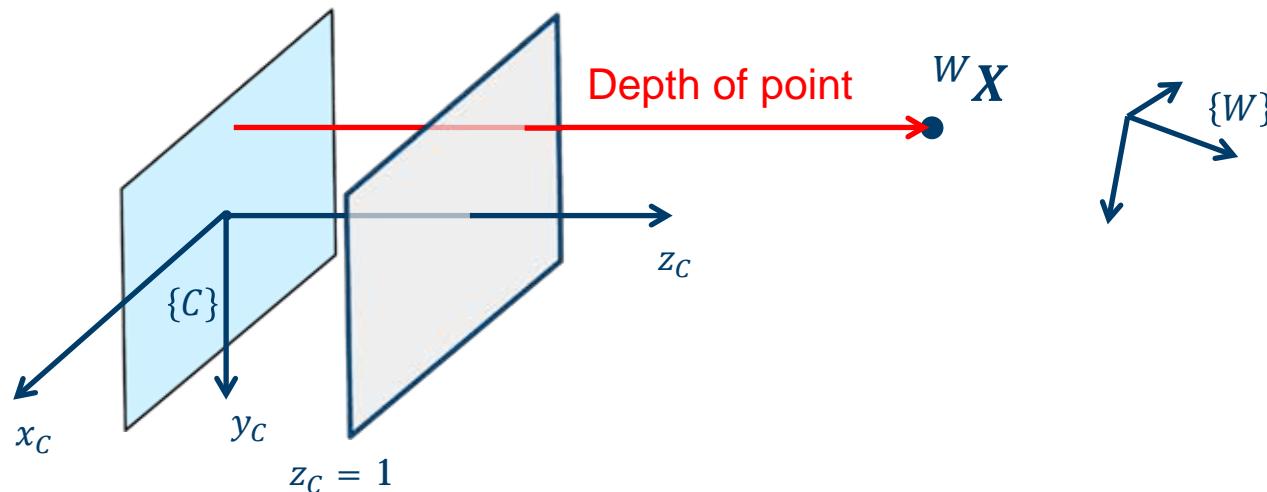
- The projective camera model $\tilde{u} = P^W \tilde{X}$ does not require world points ${}^W X$ to be in front of the camera
 - Points behind the camera will also project onto the image plane along a line through the projective center
- The depth of point formula allow us to easily check whether or not a point ${}^W X$ is in front of the camera,

$$depth({}^W X; P) = \frac{w \cdot sign(\det(M))}{W \|m^3\|}$$

where m^3 is the 3'rd row of M and

$$P[X \ Y \ Z \ W]^T = w[u \ v \ 1]^T$$

Dissecting the camera matrix



- The projective camera model $\tilde{u} = P^W \tilde{X}$ does not require world points ${}^W X$ to be in front of the camera
 - Points behind the camera will also project onto the image plane along a line through the projective center
- The depth of point formula allow us to easily check whether or not a point ${}^W X$ is in front of the camera,

$$\text{depth}({}^W X; P) = \frac{w \cdot \text{sign}(\det(M))}{W \|m^3\|}$$

where m^3 is the 3'rd row of M and

$$P[X \ Y \ Z \ W]^T = w[u \ v \ 1]^T$$

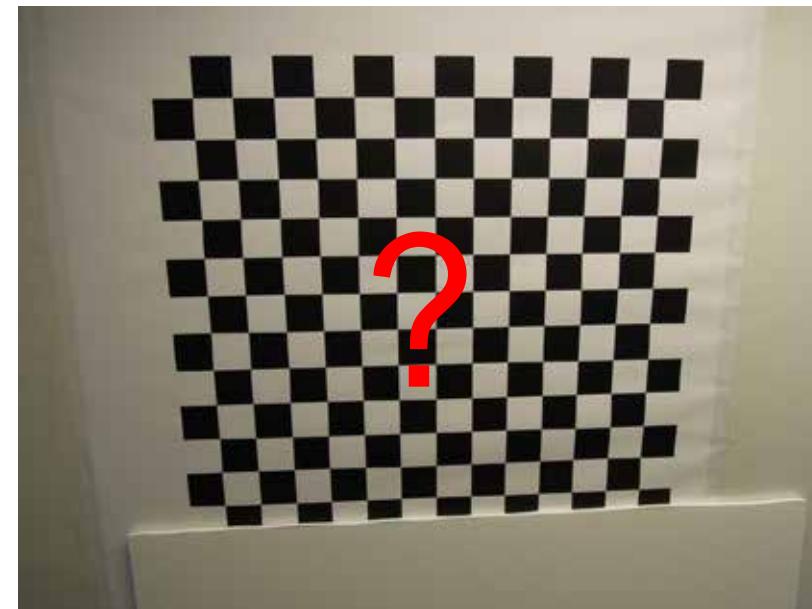
$$P = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ \color{red}{p_{31}} & \color{red}{p_{32}} & \color{red}{p_{33}} & \color{red}{p_{34}} \end{bmatrix} = \color{red}{m^3}$$

Dissecting the camera matrix

- So the general camera matrix P allow us to extract several geometrical bits of information about the camera
- Given the camera matrix P for a finite camera, it is even possible to retrieve K , R and t
 - Another lecture

Estimating the camera matrix

- Apart from constructing it from K , R and t , how can we find the camera matrix P for a given camera?
- Can we do it from an image???



Estimating the camera matrix

- Apart from constructing it from K , R and \mathbf{t} , how can we find the camera matrix P for a given camera?
- Can we do it from an image???
 - Yes, but we need to know the 3D coordinates of at least 6 points in the world and their corresponding 2D image coordinates
 - All 3D points can not be coplanar
 - Given such correspondences ${}^W\!X_i \leftrightarrow \mathbf{u}_i$, we can estimate P from the equation $\tilde{\mathbf{u}} = P {}^W\!\tilde{\mathbf{X}}$
- Solving this problem with respect to P is very similar to how we estimated the homography matrix between two images

Estimating the camera matrix P

- We separate the known quantities u, v, X, Y, Z from the unknown quantities p_{11}, \dots, p_{34}

$$\mathbf{u} = P\mathbf{x}$$

$$\begin{matrix} u \\ v \\ 1 \end{matrix} = \begin{matrix} p_{11} \\ p_{21} \\ p_{31} \end{matrix} \quad \begin{matrix} p_{12} & p_{13} & p_{14} \\ p_{22} & p_{23} & p_{24} \\ p_{32} & p_{33} & p_{34} \end{matrix} \begin{matrix} X \\ Y \\ Z \end{matrix} + \begin{matrix} 0 \\ 0 \\ 0 \end{matrix}$$

$$\begin{matrix} X & Y & Z & 1 & 0 & 0 & 0 & -uX & -uY & -uZ & -u \\ 0 & 0 & 0 & X & Y & Z & 1 & -vX & -vY & -vZ & -v \end{matrix} \mathbf{p} = \begin{matrix} 0 \\ 0 \\ 0 \end{matrix}$$

- Where

$$\mathbf{p} = [p_{11} \quad p_{12} \quad p_{13} \quad p_{14} \quad p_{21} \quad p_{22} \quad p_{23} \quad p_{24} \quad p_{31} \quad p_{32} \quad p_{33} \quad p_{34}]^T$$

- Each correspondence ${}^W\mathbf{X}_i \leftrightarrow \mathbf{u}_i$ contribute with two equations(rows)

Estimating the camera matrix P

- Hence from several 3D-2D correspondences we get a linear system of equations much like how we estimated homographies

$$\begin{matrix} \hat{x}_1 & Y_1 & Z_1 & 1 & 0 & 0 & 0 & -u_1 X_1 & -u_1 Y_1 & -u_1 Z_1 & -u_1 u & \hat{e}_0 u \\ \hat{e}_0 & 0 & 0 & 0 & X_1 & Y_1 & Z_1 & 1 & -v_1 X_1 & -v_1 Y_1 & -v_1 Z_1 & -v_1 u & \hat{e}_0 u \\ \hat{e}_1 & X_2 & Y_2 & Z_2 & 1 & 0 & 0 & 0 & -u_2 X_2 & -u_2 Y_2 & -u_2 Z_2 & -u_2 u & \hat{e}_0 u \\ \hat{e}_2 & 0 & 0 & 0 & X_2 & Y_2 & Z_2 & 1 & -v_2 X_2 & -v_2 Y_2 & -v_2 Z_2 & -v_2 u & \hat{e}_0 u \\ \hat{e}_3 & M & M & M & M & M & M & M & M & M & M & M & \hat{e}_M u \end{matrix} p = \hat{e}_0 u$$
$$Ap = 0$$

- So we get the basic DLT algorithm for estimating P from 3D-2D correspondences
 - Build matrix A (from at least 6 point-correspondences)
 - Obtain SVD of A : $A = UDV^T$
 - The non-trivial solution to $Ap = 0$ corresponds to the right singular vector with the smallest singular value
Typically the last column of V
 - Reconstruct the camera matrix P from p

Estimating the camera matrix P

- As usual, the DLT algorithm performs better if we work on normalized points
 - 2D image points should be normalized so that their centroid is at the origin and their root-mean-square distance from the origin is $\sqrt{2}$
 - 3D reference points should be normalized so that their centroid is at the origin and their RMS distance from the origin is $\sqrt{3}$
- Normalized DLT algorithm for estimating P from 3D-2D correspondences
 1. Normalize the 2D and 3D point sets with the transforms T_1 and T_2
 2. Estimate the matrix \hat{P} from normalized points using the basic DLT algorithm
Ideally one should refine the estimate by minimizing the reprojection error $\sum_i d(\tilde{\mathbf{u}}_i, \tilde{P}\tilde{\mathbf{X}}_i)^2$ iteratively
 3. Denormalize $P = T_1^{-1}\hat{P}T_2$

Summary

- The camera matrix P
 - Contains useful geometrical information
 - Vanishing points
 - Vanishing lines
 - Can be estimated from 3D-2D correspondences
- Additional reading
 - Szeliski: 6.2.1, 6.3.3
- Optional reading
 - *Single View Metrology* by A. Criminisi, I. Reid and A. Zisserman