### 9.3.4 Note on homographies that are rotations

First notice that homogeneous coordinats of points and lines constructed as combinations of joins and meets indeed behave under a homography as homogeneous coordinates constructed from affine coordinates of points.

Secondly, when the homography is a rotation and homogeneous coordinates are unit vecors, all $\lambda$ 's become equal to one, the determinant of $H$ is one and $\mathrm{H}^{-\top}=\mathrm{H}$. Therefore, all homogeneous coordinates in the previous formulas become related just by H .

### 9.4 Vanishing points

When modeling perspective projection in the affine space with affine projection planes, we meet somewhat unpleasant situations. For instance, imagine a projection of two parallel lines $K, L$, which are in a plane $\tau$ in the space into the projection plane $\pi$ through the center C, Figure 9.10

The lines $K, L$ project to image lines $k, l$. As we go with two points $X, Y$ along the lines $k, l$ away from the projection plane, their images $x, y$ get closer and closer to the point $v$ in the image but they do not reach point $v$. We shall call this point of convergence of lines $K, L$ the vanishing point

### 9.5 Vanishing line and horizon

If we take all sets of parallel lines in $\tau$, each set with a different direction, then all the points of convergence in the image will fill a complete line $h$.

The line $h$ is called the vanishing line or the horizon ${ }^{[8}$ when $\tau$ is the ground plane.

Now, imagine that we project all points from $\tau$ to $\pi$ using the affine geometrical projection model. Then, no point from $\tau$ will project to $h$.

[^0]

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Figure 9.11: Vanishing line (horizon) $h$ is the line of vanishing points.

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Similarly, when projecting in the opposite direction, i.e. $\pi$ to $\tau$, line $h$ has no image, i.e. it does not project anywhere to $\tau$.

When using the affine geometrical projection model with the real projective plane to model the perspective projection (which is equivalent to the algebraic model in $\mathbb{R}^{3}$ ), all points of the projective plane $\tau$ (obtained as the projective completion of the affine plane $\tau$ ) will have exactly one image in the projective plane $\pi$ (obtained as the projective completion of the affine plane $\pi$ ) and vice versa. This total symmetry is useful and beautiful.

## 10 Projective space

### 10.1 Motivation - the union of ideal points of all affine planes

Figure 10.1(a) shows a perspective image of three sets of parallel lines generated by sides of a cube in the three-dimensional real affine space. The images of the three sets of parallel lines converge to vanishing points $V_{1}, V_{2}$ and $V_{3}$. The cube has six faces. Each face generates two pairs of parallel lines and hence two vanishing points. Each face generates an affine plane which can be extended into a projective plane by adding the line of ideal points of that plane. The projection of the three ideal lines are vanishing lines $l_{12}=V_{1} \vee V_{2}, l_{23}=V_{2} \vee V_{3}$ and $l_{31}=V_{3} \vee V_{1}$. Imagine now all possible affine planes of the three-dimensional affine space and their corresponding ideal points. Let us take the union $V$ of the sets of ideal points of all such planes. There is exactly one ideal point for every set of parallel lines in $V$, i.e. there is a one-to-one correspondence between elements of $V$ (ideal points) and directions in the three-dimensional affine space. Notice also that every plane $\pi$ generates one ideal line $l_{\infty}$ of its ideal points and that all other planes parallel with $\pi$ generate the same $l_{\infty}$, Figure 10.1.

It suggests itself to extend the three-dimensional affine space by adding the set $V$ to it, analogically to how we have extended the affine plane. In this new space, all parallel lines will intersect. We will call this space the three-dimensional real projective space and denote it $\mathbb{P}^{3}$. Let us develop an algebraic model of $\mathbb{P}^{3}$. It is practical to require this model to encompass the model of the real projective plane. The real projective plane is modeled


Figure 10.1: (a) A perspective image of a cube generates three vanishing points $V_{1}, V_{2}$ and $V_{3}$ and hence also three vanishing lines $l_{12}$, $l_{23}$ and $l_{31}$. (b) Every plane adds one line of ideal points to the three-dimensional affine space. Every ideal point corresponds to one direction, i.e. to a set of parallel lines. Each ideal line corresponds to a set of parallel planes.

$$
\rho^{2} \equiv \mathbb{R}^{3}\left(\begin{array}{l}
10 \\
\text { shospoces } \\
\text { of } \mathbb{R}^{3}
\end{array}\right)
$$

Points $\equiv 1 D$ subspaces of $\mathbb{R}^{4}$

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algebraically by subspaces of $\mathbb{R}^{3}$. Let us observe that subspaces of $\mathbb{R}^{4}$ will be a convenient algebraic model of $\mathbb{P}^{3}$.

We start with the three-dimensional real affine space $\mathbb{A}^{3}$ and fix a coordinate system $(O, \delta)$ with $\delta=\left(\vec{d}_{1}, \vec{d}_{2}, \vec{d}_{3}\right)$. An affine plane $\pi$ is a set of points of $\mathbb{A}^{3}$ represented in $(O, \delta)$ by the set of vectors

$$
\begin{equation*}
\pi=\left\{[x, y, z]^{\top} \mid a x+b y+c z+d=0, a, b, c, d \in \mathbb{R}, a^{2}+b^{2}+c^{2} \neq 0\right\} \tag{10.1}
\end{equation*}
$$

We see that the point of $\pi$ represented by vector $[x, y, z]^{\top}$ can also be represented by one-dimensional subspace $\left\{\lambda[x, y, z, 1]^{\top} \mid \lambda \in \mathbb{R}\right\}$ of $\mathbb{R}^{4}$ and hence $\pi$ can be seen as the set
$\pi=\left\{\left\{\lambda[x, y, z, 1]^{\top} \mid \lambda \in \mathbb{R}\right\} \mid[a, b, c, d][x, y, z, 1]^{\top}=0, a, b, c, d \in \mathbb{R}, a^{2}+b^{2}+c^{2} \neq 0\right\}$
of one-dimensional subspaces of $\mathbb{R}^{4}$.
Notice that we did not require $\lambda \neq 0$ in the above definition. This is because we establish the correspondence between a vector $[x, y, z]$ and the corresponding complete one-dimensional subspace $\left\{\lambda[x, y, z, 1]^{\top}, \lambda \in \mathbb{R}\right\}$ of $\mathbb{R}^{4}$ and since every linear space contains zero vector, we admit zero $\lambda$.

Every $[x, y, z]^{\top} \in \mathbb{R}^{3}$ represents in $(O, \delta)$ a point of $\mathbb{A}^{3}$ and hence the subset

$$
\begin{equation*}
\mathbb{A}^{3}=\left\{\left\{\lambda[x, y, z, 1]^{\top} \mid \lambda \in \mathbb{R}\right\} \mid x, y, z \in \mathbb{R}\right\} \tag{10.3}
\end{equation*}
$$

of one-dimensional subspaces of $\mathbb{R}^{4}$ represents $\mathbb{A}^{3}$.
We observe that we have not used all one-dimensional subspaces of $\mathbb{R}^{4}$ to represent $\mathbb{A}^{3}$. The subset

$$
\begin{equation*}
\pi_{\infty}=\left\{\left\{\lambda[x, y, z, 0]^{\top} \mid \lambda \in \mathbb{R}\right\} \mid x, y, z \in \mathbb{R}, x^{2}+y^{2}+z^{2} \neq 0\right\} \tag{10.4}
\end{equation*}
$$

of one-dimensional subspaces of $\mathbb{R}^{4}$ is in one-to-one correspondence with all non-zero vectors of $\mathbb{R}^{3}$, i.e. in one-to-one correspondence with the set of directions in $\mathbb{A}^{3}$. This is the set of ideal points which we add to $\mathbb{A}^{3}$ to get the three-dimensional real projective space

$$
\begin{equation*}
\mathbb{P}^{3}=\left\{\left\{\lambda[x, y, z, w]^{\top} \mid \lambda \in \mathbb{R}\right\} \mid x, y, z, w \in \mathbb{R}, x^{2}+y^{2}+z^{2}+w^{2} \neq 0\right\} \tag{10.5}
\end{equation*}
$$

which is the set of all one-dimensional subspaces of $\mathbb{R}^{4}$. Notice that $\mathbb{P}^{3}=\mathbb{A}^{3} \cup \pi_{\infty}$.

Proves in $P^{3}$
§1 Points Every non-zero vector of $\mathbb{R}^{4}$ generates a one-dimensional subspace and thus represents a point of $\mathbb{P}^{3}$. The zero vector $[0,0,0,0]^{\top}$ does not represent any point.
$\S 2$ Planes Affine planes $\pi_{\mathbb{A}^{3}}$, Equation 10.2 , are in one-to-one correspondence to the subset

$$
\begin{equation*}
\pi_{\mathbb{A}^{3}}=\left\{\left\{\lambda[a, b, c, d]^{\top} \mid \lambda \in \mathbb{R}\right\} \mid a, b, c, d \in \mathbb{R}, a^{2}+b^{2}+c^{2} \neq 0\right\} \tag{10.6}
\end{equation*}
$$

of the set of one-dimensional subspaces of $\mathbb{R}^{4}$. There is only one onedimensional subspace of $\mathbb{R}^{4},\left\{\lambda[0,0,0,1]^{\top} \mid \lambda \in \mathbb{R}\right\}$ missing in $\pi_{\mathbb{A}^{3}}$. It is exactly the one-dimensional subspace corresponding to the set $\pi_{\infty}$ of ideal points of $\mathcal{P}^{3}$

$$
\begin{equation*}
\pi_{\infty}=\left\{\left\{\lambda[x, y, z, w]^{\top} \mid \lambda \in \mathbb{R}\right\} \mid x, y, z, w \in \mathbb{R}, x^{2}+y^{2}+z^{2} \neq 0,[0,0,0,1] \underset{\left.(10, y, z, w]^{\top}=0\right\}}{[x,}\right. \tag{10.7}
\end{equation*}
$$

We can take another view upon planes and observe that affine planes are in one-to-one correspondence with the three-dimensional subspaces of $\mathbb{R}^{4}$. The set $\pi_{\infty}$ also corresponds to a three-dimensional subspace of $\mathbb{R}^{4}$. Hence $\pi_{\infty}$ can be considered another plane, the ideal plane of $\mathbb{P}^{3}$.

The set of planes of $\mathbb{P}^{3}$ can be hence represented by the set of onedimensional subspaces of $\mathbb{R}^{4}$

$$
\begin{equation*}
\pi_{\mathbb{P}^{3}}=\left\{\left\{\lambda[a, b, c, d]^{\top} \mid \lambda \in \mathbb{R}\right\} \mid a, b, c, d \in \mathbb{R}, a^{2}+b^{2}+c^{2}+d^{2} \neq 0\right\} \tag{10.8}
\end{equation*}
$$

but can also be viewed as the set of three-dimensional subspaces of $\mathbb{R}^{4}$.
We see that there is a duality between points and planes of $\mathbb{P}^{3}$. They both are represented by one-dimensional subspaces of $\mathbb{R}^{4}$ and we see that point

Determined by 3 pts


$$
\operatorname{vank}=3=1
$$

$$
\text { Plow es in } P^{3} \equiv 1 D \text { subip }
$$

$$
\text { of } \mathbb{R}^{4}
$$

$X$ represented by vector $\vec{X}=[x, y, x, w]^{\top}$ is incident to plane $\pi$ represented by vector $\vec{\pi}=[a, b, c, d]^{\top}$, i.e. $X \circ \pi$, when

$$
\vec{\pi}^{\top} \vec{X}=\left[\begin{array}{llll}
a & b & c & d
\end{array}\right]\left[\begin{array}{c}
x  \tag{10.9}\\
y \\
z \\
w
\end{array}\right]=a x+b y+c z+d w=0
$$

§3 Lines Lines in $\mathbb{P}^{3}$ are represented by two-dimensional subspaces of $\mathbb{R}^{4}$. Unlike in $\mathbb{P}^{2}$, lines are not dual to points.

Lines in $P^{3}$ are a new r kind of animals.


Representations.

$$
\begin{aligned}
& \epsilon A^{3} \rightarrow \rho^{3} \\
& l \cap k=\phi
\end{aligned}
$$

lines $P^{3} \equiv 2 D$ subspaces of $\mathbb{R}^{4}$

$$
l \equiv\left\langle\vec{x}_{\phi} \vec{T}_{\phi}\right\rangle \rightarrow\langle\vec{p}\rangle \in \mathbb{R}^{6}\langle l\rangle \equiv\left\langle\vec{x}_{\phi}, \vec{Y}_{\phi}\right\rangle \rightarrow \overrightarrow{0}
$$

Plunder coordinates

$$
\langle h\rangle=\left\langle\vec{z}_{\phi}, \vec{W}_{\phi}\right\rangle \ni \overrightarrow{0}
$$

$$
\langle l\rangle \cap\langle l\rangle=\overrightarrow{0}
$$ vepresuts No point

How r to find $k^{2}$ ?

1) 6 ph ls in spoce $\rightarrow P \rightarrow\left[\left.\begin{array}{l}K R\end{array} \right\rvert\,-K R \vec{C}_{\delta}\right]$

11 Camera auto calibration

Camera auto-calibration is a process when the parameters of image formation are determined from properties of the observed scene or knowledge of camera motions. We will study camera auto-calibration methods and tasks related to metrology in images. We have seen in Chapter 7 that to measure the angle between projection rays we needed only matrix K. Actually, it is enough to know matrix ${ }^{1}$

$\varepsilon \ldots$ orthonormal bases
Knowing $\omega$ is however (almost) equivalent to knowing $K$ since $K$ can be recovered from $\omega$ up to two signs as follows.
$\S 1$ Recovering K from $\omega$ Let us give a procedure for recovering K from $\omega$. Assuming

$$
\eta \rightarrow \omega \stackrel{\imath}{\sim} \mathrm{K}=\left[\begin{array}{ccc}
k_{11} & k_{12} & k_{13}  \tag{11.2}\\
0 & k_{22} & k_{23} \\
0 & 0 & 1
\end{array}\right]
$$

N... orthogonal boisis ( $f$.)

$$
\begin{aligned}
\cos \alpha & =\frac{\vec{x}_{1 \varepsilon}^{\top} \vec{x}_{2 \varepsilon}}{\left\|\vec{x}_{1 \varepsilon}\right\|\left\|\vec{x}_{2 \varepsilon}\right\|}=\frac{\hat{f}_{f} \vec{x}_{1 \gamma}^{\top} \frac{1}{f} \vec{x}_{2 r}}{\left\|\frac{1}{f} \vec{x}_{1 r}\right\|\left\|\overrightarrow{1}_{f}^{1} \vec{x}_{2}\right\|} \\
& =*
\end{aligned}
$$

${ }^{1}$ In [15], $\omega$ is called the image of the absolute conic.

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we get
$\left.\mathrm{K}^{-1}=\left[\begin{array}{ccc}k_{11} & k_{12} & k_{13} \\ 0 & k_{22} & k_{23} \\ 0 & 1\end{array}\right]\right]^{-1}=\left[\begin{array}{ccc}\frac{1}{k_{11}} & \frac{-k_{12}}{k_{11} k_{22}} & \frac{k_{12} k_{23}-k_{13} k_{22}}{k_{11} k_{22}} \\ 0 & \frac{1}{k_{22}} & \frac{-k_{23}}{k_{22}} \\ 0 & 0 & 1\end{array}\right]=\left[\begin{array}{ccc}m_{11} & m_{12} & m_{13} \\ 0 & m_{22} & m_{23} \\ 0 & 0 & 1\end{array}\right] \stackrel{\omega}{\mathrm{K}} \mathrm{K}$
for some real $m_{11}, m_{12}, m_{13}, m_{22}$ and $m_{23}$. Equivalently, we get
$\mathrm{K}=\left[\begin{array}{ccc}\frac{1}{m_{11}} & \frac{-m_{12}}{m_{11} m_{22}} & \frac{m_{12} m_{23}-m_{13} m_{22}}{m_{11} m_{22} m_{23}} \\ 0 & \frac{1}{m_{22}} & \frac{-m_{23}}{m_{22}} \\ 0 & 0 & 1\end{array}\right] \quad \mathrm{M}$
Introducing the following notation

$$
\omega=\mathrm{K}^{-\top} \mathrm{K}^{-1}=\left[\begin{array}{lll}
\omega_{11} & \omega_{12} & \omega_{13}  \tag{11.5}\\
\omega_{12} & \omega_{22} & \omega_{23} \\
\omega_{13} & \omega_{23} & \omega_{33}
\end{array}\right]
$$

yields

$$
\left[\begin{array}{ccc}
\omega_{11} & \omega_{12} & \omega_{13} \\
\omega_{12} & \omega_{22} & \omega_{23} \\
\omega_{13} & \omega_{23} & \omega_{33}
\end{array}\right]=\left[\begin{array}{ccc}
m_{11}^{2} & m_{11} m_{12} & m_{11} m_{13} \\
m_{11} m_{12} & m_{12}^{2}+m_{22}^{2} & m_{12} m_{13}+m_{22} m_{23} \\
m_{11} m_{13} & m_{12} m_{13}+m_{22} m_{23} & m_{13}^{2}+m_{23}^{2}+1
\end{array}\right]
$$

$$
\omega=\quad M^{\top} M
$$

which can be solved for $\mathrm{K}^{-1}$ up to the sign of the rows of $\mathrm{K}^{-1}$ as follows. From w to K
which can be solved for $m_{i j}$ with $s_{1}= \pm 1$ and $s_{2}= \pm 1$. Hence


Signs $s_{1}, s_{2}$ are determined by the choice of the image coordinate system. The standard choice is $s_{1}=s_{2}=1$, which corresponds to $k_{11}>0$ and $k_{22}>0$.

Notice that $\sqrt{\omega_{11}}$ is never zero for a real camera since $m_{11}=\frac{1}{k_{11}} \neq 0$.

$$
\begin{aligned}
& s_{1}= \pm\left\{\begin{array}{l}
s_{1}=+ \\
s_{2}= \pm
\end{array}\right\} \begin{array}{l}
s_{2}=+
\end{array} \text { }
\end{aligned}
$$

$\sqrt{\omega_{22}-\omega_{12}^{2} / \omega_{11}}=\sqrt{m_{11}^{2}-m_{12}^{2}}=\sqrt{\frac{1}{k_{11}^{2}-\frac{k_{12}^{2}}{k_{11}^{2} k_{22}^{2}}}=\frac{1}{k_{11} k_{22}} \overbrace{\sqrt{k_{22}^{2}-k_{12}^{2}} \neq 0}^{\text {There also holds true }} 00}$ since $\left|k_{12}\right|$ is much smaller than $\left|k_{22}\right|$ for all real cameras.

Equations by comparing clements

$3 \times 3=9$ elements
knt only 6 index. equs.
$\Rightarrow 5$ unknowns
4 choices $\mathrm{N} \longrightarrow M$

$$
k=\left[\begin{array}{ccc}
>0 & k_{12} &  \tag{11.8}\\
0 & >0 & 1 \\
0 & 0 & 1
\end{array}\right]
$$

$$
\begin{aligned}
& \text { Equation } 11.6 \text { provides equations } \\
& \text { Equation 11.6 provides equations } \\
& \omega_{11}=m_{11}^{2} \xlongequal{\Rightarrow} m_{11}=s_{1} \sqrt{\omega_{11}} \\
& \omega_{12}=m_{11} m_{12} \xlongequal{\cong} \quad m_{12}=\omega_{12} /\left(s_{1} \sqrt{\omega_{11}}\right)=s_{1} \omega_{12} / \sqrt{\omega_{11}} \\
& \omega_{13}=m_{11} m_{13} \quad \Rightarrow \quad m_{13}=\omega_{13} /\left(s_{1} \sqrt{\omega_{11}}\right)=s_{1} \omega_{13} / \sqrt{\omega_{11}} \\
& \omega_{22}=m_{12}^{2}+m_{22}^{2} \stackrel{y}{\Rightarrow} m_{22}=s_{2} \sqrt{\omega_{22}-m_{12}^{2}}=s_{2} \sqrt{\omega_{22}-\omega_{12}^{2} / \omega_{11}} \\
& \omega_{23}=m_{12} m_{13}+m_{22} m_{23} \stackrel{\zeta}{\leftrightharpoons} \quad m_{23}=s_{2}\left(\omega_{23}-\omega_{12} \omega_{13} / \omega_{11}\right) / \sqrt{\omega_{22}-\omega_{12}^{2} / \omega_{11}} \\
& =s_{2}\left(\omega_{11} \omega_{23}-\omega_{12} \omega_{13}\right) / \sqrt{\omega_{11}^{2} \omega_{22}-\omega_{11} \omega_{12}^{2}}
\end{aligned}
$$

### 11.1 Constraints on $\omega$

Matrix $\omega$ is a $3 \times 3$ symmetric matrix and by this it has only six independent elements $\omega_{11}, \omega_{12}, \omega_{13}, \omega_{22}, \omega_{23}$ and $\omega_{33}$. Let us next investigate additional constratints on $\omega$, which follow from different choices of $K$.
§1 Constraints on $\omega$ for a general K Even a general K yields a constraint on $\omega$. Equation 11.6 relates the six parameters of $\omega$ to only five parameters $m_{11}, m_{12}, m_{13}, m_{22}$ and $m_{23}$ and hence the six parameters of $\omega$ can't be independent. Indeed, let us see that the following identity holds true

$$
\begin{align*}
& \left(\omega_{23}^{2}-\frac{\omega_{13}^{2} \omega_{12}^{2}}{\omega_{11}^{2}}-\left(\omega_{22}-\frac{\omega_{12}^{2}}{\omega_{11}}\right)\left(\omega_{33}-\frac{\omega_{13}^{2}}{\omega_{11}}-1\right)\right)^{2}-4 \frac{\omega_{13}^{2} \omega_{12}^{2}}{\omega_{11}^{2}}\left(\omega_{22}-\frac{\omega_{12}^{2}}{\omega_{11}}\right)\left(\omega_{33}-\frac{\omega_{13}^{2}}{\omega_{11}}-1\right) \\
= & \left(\left(m_{12} m_{13}+m_{22} m_{23}\right)^{2}-\frac{\left(m_{11} m_{13}\right)^{2}\left(m_{11} m_{12}\right)^{2}}{m_{11}^{4}}\right. \\
& \left.-\left(m_{12}^{2}+m_{22}^{2}-\frac{\left(m_{11} m_{12}\right)^{2}}{m_{11}^{2}}\right)\left(m_{13}^{2}+m_{23}^{2}+1-\frac{\left(m_{11} m_{13}\right)^{2}}{m_{11}^{4}}-1\right)\right)^{2} \\
- & 4 \frac{\left(m_{11} m_{13}\right)^{2}\left(m_{11} m_{12}\right)^{2}}{m_{11}^{4}}\left(m_{12}^{2}+m_{22}^{2}-\frac{\left(m_{11} m_{12}\right)^{2}}{m_{11}^{2}}\right)\left(m_{13}^{2}+m_{23}^{2}+1-\frac{\left(m_{11} m_{13}\right)^{2}}{m_{11}^{4}}-1\right) \\
= & \left(\left(m_{12} m_{13}+m_{22} m_{23}\right)^{2}-\left(m_{12} m_{13}\right)^{2}-\left(m_{22} m_{23}\right)^{2}\right)^{2}-4\left(m_{12} m_{13}\right)^{2}\left(m_{22} m_{23}\right)^{2} \\
= & \left(2\left(m_{12} m_{13}\right)\left(m_{22} m_{23}\right)\right)^{2}-4\left(m_{12} m_{13}\right)^{2}\left(m_{22} m_{23}\right)^{2} \\
= & 0 \tag{11.9}
\end{align*}
$$

Since $\omega_{11} \neq 0$, we get the following equivalent identity

$$
\begin{align*}
&\left(\omega_{11}^{2} \omega_{23}^{2}-\omega_{13}^{2} \omega_{12}^{2}-\left(\omega_{11} \omega_{22}-\omega_{12}^{2}\right)\left(\omega_{11} \omega_{33}-\omega_{13}^{2}-\omega_{11}\right)\right)^{2} \\
&-4 \omega_{13}^{2} \omega_{12}^{2}\left(\omega_{11} \omega_{22}-\omega_{12}^{2}\right)\left(\omega_{11} \omega_{33}-\omega_{13}^{2}-\omega_{11}\right)=0 \tag{11.10}
\end{align*}
$$

which is a polynomial equation of degree eight in elements of $\omega$.

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We shall see next that it makes sense to introduce a new matrix

$$
\Omega=\left[\begin{array}{lll}
(1) & \kappa_{12} & o_{13} \\
\sigma_{12} & o_{22} & o_{23} \\
o_{13} & o_{23} & o_{33}
\end{array}\right]=\left[\begin{array}{ccc}
1 & \frac{\omega_{12}}{\omega_{11}} & \frac{\omega_{13}}{\omega_{11}} \\
\boldsymbol{a} & \frac{\omega_{12}}{\omega_{11}} & \frac{\omega_{22}}{\omega_{11}} \\
\frac{\omega_{23}}{\omega_{11}} \\
\frac{\omega_{13}}{\omega_{11}} & \frac{\omega_{23}}{\omega_{11}} & \frac{\omega_{33}}{\omega_{11}}
\end{array}\right]=\frac{\omega}{\omega_{11}(11}
$$

which contains only five unknowns, and use Equation 11.10 to get the positive $\omega_{11}$ from $\Omega$ by solving the following quadratic equation
with

$$
\begin{align*}
a_{2}= & -4 o_{23}{ }^{2} o_{13}{ }^{2} o_{12}{ }^{2}+o_{23}{ }^{4}-2 o_{23}{ }^{2} o_{22} o_{33}+o_{13}{ }^{2} o_{12}{ }^{2} o_{22} o_{33}(1 \\
& -2 o_{22}{ }^{2} o_{33} o_{13}{ }^{2}+o_{12}{ }^{4} o_{33}{ }^{2}+2 o_{23}{ }^{2} o_{22} o_{13}{ }^{2}=2 o_{23}{ }^{2} o_{12}{ }^{2} o_{33} \\
& +o_{22}{ }^{2} o_{13}{ }^{4}+o_{22}{ }^{2} o_{33}{ }^{2}-2 o_{22} o_{33}{ }^{2} o_{12}{ }^{2} \\
a_{1}= & 2 o_{13}{ }^{2} o_{12}{ }^{2} o_{22}+2 o_{23}{ }^{2} o_{22}-2 o_{22}{ }^{2} o_{33}-2 o_{12}{ }^{4} o_{3}^{2}  \tag{11.14}\\
& +4 o_{22} o_{33} o_{12}^{2}-2 o_{23} o_{12}^{2}+2 o_{22}^{2} o_{13}{ }^{2}
\end{align*}
$$

$$
\begin{equation*}
a_{2} \omega_{11}^{2}+a_{1} \omega_{11}+a_{0}=0 \tag{11.12}
\end{equation*}
$$


§2 Constraints on $\omega$ for K from square pixels Cameras have often square pixels, i.e. $\left\|\vec{b}_{1}\right\|=\mid \overrightarrow{b_{2}} \|=1$ and $\angle\left(\vec{b}_{1}, \vec{b}_{2}\right)=\pi / 2$, which implies, Equations 叉.13, 7.15, 7.16, a simplified

$$
\mathrm{K}=\left[\right]
$$




This gives also simpler

$$
\begin{aligned}
& \text { simpler } \\
& \omega=\frac{1}{k_{11}^{2}}\left[\begin{array}{ccc}
1 & 0 & -k_{13} \\
0 & 1 & -k_{23} \\
-k_{13} & -k_{23} & k_{11}^{2}+k_{13}^{2}+k_{23}^{2}
\end{array}\right]<\frac{1}{k_{11}^{2}} \Omega(11.17) \\
& \omega \rightarrow K
\end{aligned}
$$

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We see that we get the following three identities

$$
\begin{align*}
\omega_{12} & =0  \tag{11.18}\\
\omega_{22}-\omega_{11} & =0  \tag{11.19}\\
\omega_{13}^{2}+\omega_{23}^{2}-\omega_{11} \omega_{33}+\omega_{11} & =0 \tag{11.20}
\end{align*}
$$

We also get simpler
$\Omega\left[\begin{array}{ccc}1 & 0 & \left(o_{13}\right. \\ 0 & 1 & o_{23} \\ o_{13} & o_{23} & o_{33}\end{array}\right]=k_{11}^{2} \omega=\left[\begin{array}{ccc}1 & 0 & -k_{13} \\ 0 & 1 & -k_{23} \\ -k_{13} & -k_{23} & k_{11}^{2}+k_{13}^{2}+k_{23}^{2}\end{array}\right]$
and use Equation 11.21 to get

$$
\begin{align*}
& k_{11}^{2}=o_{33}-o_{13}^{2}-o_{23}^{2}  \tag{11.22}\\
& k_{13}=-o_{13}  \tag{11.23}\\
& k_{23}=-o_{23} \tag{11.24}
\end{align*}
$$

11.2 Camera calibration from angles between projection rays

We will now show how to calibrate a camera by finding the matrix $\omega=\mathrm{K}^{-\top} \mathrm{K}^{-1}$.

In general, matrix $\omega$ is constrained by knowing angles contained between pairs of projection rays. Consider two projection rays with direction vectors $\vec{x}_{1}, \vec{x}_{2}$. Then the angle between them is related to $\omega$ and $\Omega$ by

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Squaring the above and clearing the denominators gives unkur~~

which is a second order equation in elements of $\Omega$. Tिfind $\Omega$, which has five independent parameters for a general $K$, we need to be able to establish five pairs of rays with known angles and solve a system of five quadratic equations 11.26 above.
§1 Camera with square pixels A simpler situation arises when the camera has square pixels. Then, we can use constraints from $\$ 2$ to recover $\omega$ and K from three pairs of rays containing known angles. That amounts to solving three second order equations 11.26 in $o_{13}, o_{23}, o_{33}$.

However, this is actually exactly the same problem as we have already solved in Section 7.3. Figure 11.2 shows an image plane $\pi$ with a coordinate system $\left(0, \delta^{\prime}\right)$ with $\delta^{\prime}=\left(\vec{b}_{1}, \vec{b}_{2}, \vec{b}_{3}^{\prime}\right)$ derived from the image coordinate system $(o, \alpha)$. Having square pixels, vectors $\vec{b}_{1}, \vec{b}_{2}$ can be complemented with $\vec{b}_{3}^{\prime}$ to form an orthogonal coordinates system $\left(O=0, \delta^{\prime}\right)$. Next, we choose the global orthonormal coordinate system, $(O=0, \delta), \delta=$ $\left(\overrightarrow{d_{1}}, \overrightarrow{d_{2}}, \overrightarrow{d_{3}}\right)$, such that

$$
\begin{equation*}
\vec{d}_{1}=\frac{\vec{b}_{1}}{\left\|\vec{b}_{1}\right\|}, \quad \vec{d}_{2}=\frac{\vec{b}_{2}}{\left\|\vec{b}_{1}\right\|}, \quad \text { and } \quad \vec{d}_{3}=\frac{\vec{b}_{3}^{\prime}}{\left\|\vec{b}_{1}\right\|} \tag{11.27}
\end{equation*}
$$

and hence

$$
\vec{x}_{\delta}=\left[\begin{array}{ccc}
\left\|\vec{b}_{1}\right\| & 0 & 0  \tag{11.28}\\
0 & \left\|\vec{b}_{1}\right\| & 0 \\
0 & 0 & \left\|\vec{b}_{1}\right\|
\end{array}\right] \vec{x}_{\delta^{\prime}}
$$

We know angles $\angle\left(\vec{x}_{1}, \vec{x}_{2}\right), \angle\left(\vec{x}_{2}, \vec{x}_{3}\right)$ and $\angle\left(\vec{x}_{3}, \vec{x}_{1}\right)$. We also know image points $\vec{u}_{1 \alpha}=\vec{X}_{1 \delta^{\prime}}, \vec{u}_{2 \alpha}=\vec{X}_{2 \delta^{\prime}}, \vec{u}_{3 \alpha}=\vec{X}_{3 \delta^{\prime}}$ and thus we can compute

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distances $d_{12}=\left\|\vec{X}_{2 \delta^{\prime}}-\vec{X}_{1 \delta^{\prime}}\right\|, d_{23}=\left\|\vec{X}_{3 \delta^{\prime}}-\vec{X}_{2 \delta^{\prime}}\right\|$ and $d_{31}=\left\|\vec{X}_{3 \delta^{\prime}}-\vec{X}_{1 \delta^{\prime}}\right\|$. Having that, we can find the pose $\vec{C}_{\delta^{\prime}}=\left[c_{1}, c_{2}, c_{3}\right]^{\top}$ of the camera center $C$ in $\left(O, \delta^{\prime}\right)$ by solving the absolute pose problem from Chapter 7.3. We will select a solution with $c_{3}<0$ and, if necessary, use a fourth point in $\pi$ to choose the right solution among them. To find K , we can form the following equation

$$
\left[\begin{array}{l}
0  \tag{11.29}\\
0 \\
1
\end{array}\right]=\frac{1}{f}\left[\mathrm{KR} \mid-\mathrm{KR} \overrightarrow{\mathrm{C}}_{\delta}\right]\left[\begin{array}{l}
0 \\
0 \\
0 \\
1
\end{array}\right]
$$

since point $o$ is represented by $[0,0,1]^{\top}$ in $\beta$ and by $[0,0,0]^{\top}$ in $\delta$. Coordinate system $(O, \delta)$ is chosen such that $R=I$ and $\vec{C}_{\delta}=\left\|\vec{b}_{1}\right\| \vec{C}_{\delta^{\prime}}$ and thus we get

$$
\mathrm{K}^{-1}\left[\begin{array}{l}
0  \tag{11.30}\\
0 \\
1
\end{array}\right]=-\frac{\left\|\overrightarrow{b_{1}}\right\|}{f} \vec{C}_{\delta^{\prime}}
$$

Now, let us consider matrix $K$ as in Equation 11.16 and use the interpretation of elements of K from Chapter 7 , Equations 7.16, 7.17 We can write

$$
\mathrm{K}=\left[\begin{array}{ccc}
\frac{f}{\left\|\vec{b}_{1}\right\|} & 0 & k_{13}  \tag{11.31}\\
0 & \frac{f}{\left\|\vec{b}_{1}\right\|} & k_{23} \\
0 & 0 & 1
\end{array}\right] \text { an thus } \mathrm{K}^{-1}=\left[\begin{array}{ccc}
\frac{\left\|\overrightarrow{b_{1}}\right\|}{f} & 0 & -\frac{\left\|\overrightarrow{b_{1}}\right\|}{f} k_{13} \\
0 & \frac{\left\|\vec{b}_{1}\right\|}{f} & -\frac{\left\|\vec{b}_{1}\right\|}{f} k_{23} \\
0 & 0 & 1
\end{array}\right]
$$ and use it in Equation 11.30 to get

$$
\left[\begin{array}{r}
k_{13}  \tag{11.32}\\
k_{23} \\
-\frac{f}{\left\|\overrightarrow{b_{1}}\right\|}
\end{array}\right]=\left[\begin{array}{l}
c_{1} \\
c_{2} \\
c_{3}
\end{array}\right]
$$

而

TPajdla. Elements of Geometry for Computer Vision and Graphics 2020-4-13 (pajdla@cvut . cz and thus

$$
K=\left[\begin{array}{rrr}
-c_{3} & 0 & c_{1}  \tag{11.33}\\
0 & -c_{3} & c_{2} \\
0 & 0 & 1
\end{array}\right]
$$

### 11.3 Camera calibration from vanishing points

Let us first make an interesting observation about parallel lines in space an its corresponding vanishing point in an image. Let us consider a pair of parallel lines $K, L$ in space as shown in Figure 11.1(a). There is an 2 ffine plane $\sigma$ containing the lines. The lines $K, L$ are projected to image plane $\pi$ into lines $k$, $l$, respectively.

Now, first extend affine plane $\sigma$ to a projective plane $\Sigma$ using the camera center $C$. Then, define a coordinate system $(C, \delta)$ with orthongrmal basis $\delta=\left(\overrightarrow{d_{1}}, \overrightarrow{d_{2}}, \overrightarrow{d_{3}}\right)$ such that vectors $\vec{d}_{1}, \vec{d}_{2}$ span affine plane $\sigma$.

Let $\vec{K}_{\bar{\delta}}, \vec{L}_{\bar{\delta}}$ be homogeneous coordinates of lines $K, L$ w.r.t. $\bar{\delta}$. Then

$$
\begin{equation*}
\vec{w}_{\delta}=\vec{K}_{\bar{\delta}} \times \vec{L}_{\bar{\delta}} \tag{11.34}
\end{equation*}
$$

are homogeneous coordinates of the intersection of lines $K, L$ in $\Sigma$.
Next, extend the affine plane $\pi$ to a prgjective plane $\Pi$ using the camera center $C$ with the (camera) coordinatesystem ( $C, \beta$ ).

Let $\vec{k}_{\bar{\beta}}, \vec{l}_{\bar{\beta}}$ be homogeneous coordinates of lines $k, l$ w.r.t. $\bar{\beta}$. Then

$$
\begin{equation*}
\vec{v}_{\beta}=\vec{k}_{\bar{\beta}} \times \vec{l}_{\bar{\beta}} \tag{11.35}
\end{equation*}
$$

are homogeneous coordinates of the intersection of lines $k, l$ in $\Pi$.
Now, consider Equation 8.14 for planes $\Sigma$ and $\Pi$. Since $\delta$ is orthonormal, we have $\mathrm{K}^{\prime}=\mathrm{I}$ and thus that there is a homoghraphy

$$
\begin{equation*}
\mathrm{H}=\mathrm{K} \mathrm{R} \tag{11.36}
\end{equation*}
$$


which maps plane $\Sigma$ to plane $\Pi$. Matrices $K$ and $R$ of the camera are here w.r.t. the world coordinate system $(C, \delta)$.

We see that there is a real $\lambda$ such that there holds

$$
\begin{equation*}
\lambda \vec{v}_{\beta}=\mathrm{KR} \vec{w}_{\delta} \tag{11.37}
\end{equation*}
$$

true.
§1 Pairs of "orthogonal" vanishing points and camera with square pixels Let us have two pairs of parallel lines in space, Figure 11.1(b), such that they are also orthogonal, i.e. let $K_{1}$ be parallel with $L_{1}$ and $K_{2}$ be parallel with $L_{2}$ and at the same time let $K_{1}$ be orthogonal to $K_{2}$ and $L_{1}$ be orthogonal to $L_{2}$. This, for instance, happens when lines $K_{1}, L_{1}, K_{2}, L_{2}$ form a rectangle but they also may be arranged in the three-dimensional space as non-intersecting.

Let lines $k_{1}, l_{1}, k_{2}, l_{2}$ be the projections of $K_{1}, L_{1}, K_{2}, L_{2}$, respectively, represented by the corresponding vectors $\vec{k}_{1 \overline{1}}, \vec{l}_{1 \bar{\beta}}, \vec{k}_{2 \bar{\beta}}, \vec{l}_{2 \bar{\beta}}$ in the camera coordinates system with (in general non-orthogonal) basis $\beta$. Lines $k_{1}$ and $l_{1}$, resp. $k_{2}$ and $l_{2}$, generate vanishing points

$$
\begin{aligned}
& \vec{v}_{1 \beta}=\vec{k}_{1 \bar{\beta}} \times \vec{l}_{1 \bar{\beta}} \\
& \vec{v}_{2 \beta}=\vec{k}_{2 \bar{\beta}} \times \vec{l}_{2 \bar{\beta}}
\end{aligned}
$$

The perpendicularity of $\vec{w}_{1}$ to $\vec{w}_{2}$, in the camera orthogonal basis $\delta$, modeled by

We therefore get from Equation 11.37


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which is a linear homogeneous equation in $\omega$. Assuming further square pixels, we get, §2,

$$
\begin{aligned}
& \begin{array}{l}
\vec{v}_{1 \beta}^{\top} \omega \vec{v}_{2 \beta}=0 \\
\vec{v}_{1 \beta}^{\top} \Omega \vec{v}_{2 \beta}=0
\end{array} \downarrow \text { pixels } \\
& {\left[\begin{array}{lll}
v_{11} & v_{12} & v_{13}
\end{array}\right]\left[\begin{array}{lll}
1 & 9 & o_{13} \\
0 & & o_{23} \\
o_{13} & v_{23} & o_{33}
\end{array}\right]\left[\begin{array}{l}
v_{21} \\
v_{22} \\
v_{23}
\end{array}\right]=0} \\
& {\left[\begin{array}{lll}
v_{23} v_{11}+v_{21} v_{13} & v_{23} v_{12}+v_{22} v_{13} & v_{23} v_{13}
\end{array}\right]\left[\begin{array}{l}
o_{13} \\
o_{23} \\
o_{33}
\end{array}\right]=-\left(v_{21} v_{11}+v_{22} v_{12}\right) \epsilon_{七}}
\end{aligned}
$$



Now, we need only 3 pairs of perpendicular vanishing points, e.g. to observe 3 rectangles not all in one plane to compute $o_{13}, 0_{23}, O_{33}$ and then

$$
\begin{aligned}
k_{13} & =-o_{13} \\
k_{23} & =-o_{23} \\
k_{11} & =\sqrt{o_{33}-k_{13}^{2}-k_{23}^{2}}
\end{aligned}
$$


which is a linear homogeneous equation in $\omega$. Assuming, further square pixels, we get, \$2.

$$
\vec{v}_{1 \beta}^{\top} \omega \vec{v}_{2 \beta}=0
$$

$$
\left[\begin{array}{lll}
v_{11} & v_{12} & v_{13}
\end{array}\right]
$$

$\left[\begin{array}{lll}v_{23} v_{11}+v_{21} v_{13} & v_{23} v_{12}+v_{22} v_{13} & v_{23}\end{array}\right]\left[\begin{array}{l}o_{13} \\ o_{23} \\ o_{33}\end{array}\right]=-\left(v_{21} v_{11}+v_{22} v_{12}\right)$
Now, we need only 3 pairs of perpendicular vanishing points, e.g. to observe 3 rectangles not all in one plane to compute $o_{13}, o_{23}, o_{33}$ and then

$$
\begin{aligned}
& k_{13}=-o_{13} \\
& k_{23}=-o_{23} \\
& k_{11}=\sqrt{o_{33}-k_{13}^{2}-k_{23}^{2}}
\end{aligned}
$$

### 11.4 Camera calibration from images of squares

Let us exploit the relationship between the coordinates of points $X$, which all lie in a plane $\sigma$ and are measured in a coordinate system $\left(O, \overrightarrow{d_{1}}, \overrightarrow{d_{2}}\right)$ in $\sigma$, Figure 8.2. The points $X$ are projected by a perspective camera with the camera coordinate system is $(C, \beta), \beta=\left(\vec{b}_{1}, \vec{b}_{2}, \vec{b}_{3}\right)$ and projection matrix P into image coordinates $\left[\begin{array}{ll}u & v\end{array}\right]^{\top}$, w.r.t. an image coordinate system $\left(0, \vec{b}_{1}, \vec{b}_{2}\right)$, Equation 8.16 See paragraph $\$ 1$ to recall that the columns of $P$ can be writen as

$$
\mathrm{P}=\left[\mathrm{KR} \mid-\mathrm{KR} \vec{C}_{\delta}\right]=\left[\begin{array}{llll}
\vec{d}_{181} & \vec{d}_{2 v} & \vec{d}_{3 v} & -\vec{C}_{v} \tag{11.42}
\end{array}\right]
$$



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$$
\begin{align*}
\mathrm{h}_{1}=\mathrm{p}_{1} & =\vec{d}_{1 v}  \tag{11.43}\\
\mathrm{~h}_{2}=\mathrm{p}_{2} & =\vec{d}_{2 v}  \tag{11.44}\\
\mathrm{~h}_{3}=\mathrm{p}_{4} & =-\vec{C}_{v} \tag{11.45}
\end{align*}
$$

of the homography H mapping $\sigma$ to $\pi$ as defined in Equation 8.17
Now imagine that we axe observing a square with 4 cфrner points $X_{1}$, $X_{2}, X_{3}$ and $X_{4}$ in the plane $\sigma$ and we construct the coordinate system in $\sigma$ by assigning coordinates to the cornersas

$$
\begin{align*}
\vec{X}_{1 \delta} & =\left[\begin{array}{lll}
0 & 0 & 0
\end{array}\right]  \tag{11.46}\\
\vec{d}_{2 \delta}=\vec{X}_{2 \delta} & =\left[\begin{array}{lll}
1 & 0 & 0
\end{array}\right]  \tag{11.47}\\
\vec{d}_{3 \delta}=\vec{X}_{3 \delta} & =\left[\begin{array}{lll}
0 & 1 & 0
\end{array}\right]  \tag{11.48}\\
\vec{X}_{4 \delta} & =\left[\begin{array}{lll}
1 & 1 & 0
\end{array}\right] \tag{11.49}
\end{align*}
$$

$$
\begin{align*}
& \text { We see that we get two constraints on } \vec{d}_{1 \delta}, \vec{d}_{2 \delta} \\
& \left.\begin{array}{r}
\qquad\|=\| \| \vec{d}_{1 \delta}^{\dagger} \vec{d}_{2 \delta}=0
\end{array}\right]  \tag{11.50}\\
& \vec{d}_{1 \beta}^{\top} \mathrm{K}^{-\top} \mathrm{K}^{-1} \vec{d}_{1 \beta}-\vec{d}_{2 v}^{\top} \mathrm{K}^{-\top} \mathrm{K}^{-1} \vec{d}_{2 v}=0 \tag{11.52}
\end{align*}
$$

which lead to
by using $\vec{d}_{i v}=\mathrm{K} \mathrm{R} \vec{d}_{i \delta}$ for $i=1,2$, and $\mathrm{R}^{\top} \mathrm{R}=\mathrm{I}$.
These are two linear equation ond hence also, see $\S 1$, on $\Omega$

$$
\left(\begin{array}{rl}
\overrightarrow{d_{1 v}^{\dagger}} \Omega \vec{d}_{2 v} & =0  \tag{11.54}\\
\overrightarrow{d_{1 v}^{\top}} \Omega \vec{d}_{1 v}-\vec{d}_{2 v}^{\dagger} \Omega \vec{d}_{2 v} & =0 \\
102
\end{array}\right.
$$

on $\omega$ in terms of estimated $\lambda \mathrm{H}$

$$
\begin{align*}
\mathrm{h}_{1}^{\top} \Omega \mathrm{h}_{2} & =0  \tag{11.56}\\
\mathrm{~h}_{1}^{\top} \Omega \mathrm{h}_{1}-\mathrm{h}_{2}^{\top} \Omega \mathrm{h}_{2} & =0 \tag{11.57}
\end{align*}
$$

One square provides two equations and therefore three squares in two planes in a general position suffice to calibrate full K. Actually, such three squares provide one more equations than necessary since $\Omega$ has only five parameters. Hence, it is enough observe two squares and one rectangle to get five constraints. Similarly, one square and one rectangle in a plane then suffice to calibrate K when pixels are square.

Notice also that we have never used the special choice of coordinates of $\vec{X}_{\delta}$. Indeed, point $X_{4}$ could be anywhere provided that we know how to assign it coordinates in $\left(O, \overrightarrow{d_{1}}, \overrightarrow{d_{2}}\right)$.

To calibrate the camera, we first assign coordinates to the corners of the square as above, then find the homography H from the plane to the image

$$
\begin{equation*}
\lambda_{i} \vec{x}_{i \beta}=\mathrm{H} \vec{X}_{i \delta} \tag{11.58}
\end{equation*}
$$

for $\alpha_{i}=1, \ldots, 4$ and finally use columns of H the find $\Omega$.


[^0]:    ${ }^{7}$ Úběžník in Czech.
    ${ }^{8}$ Horizont in Czech

