

Camera Calibration



Real-time Facial Reenactment



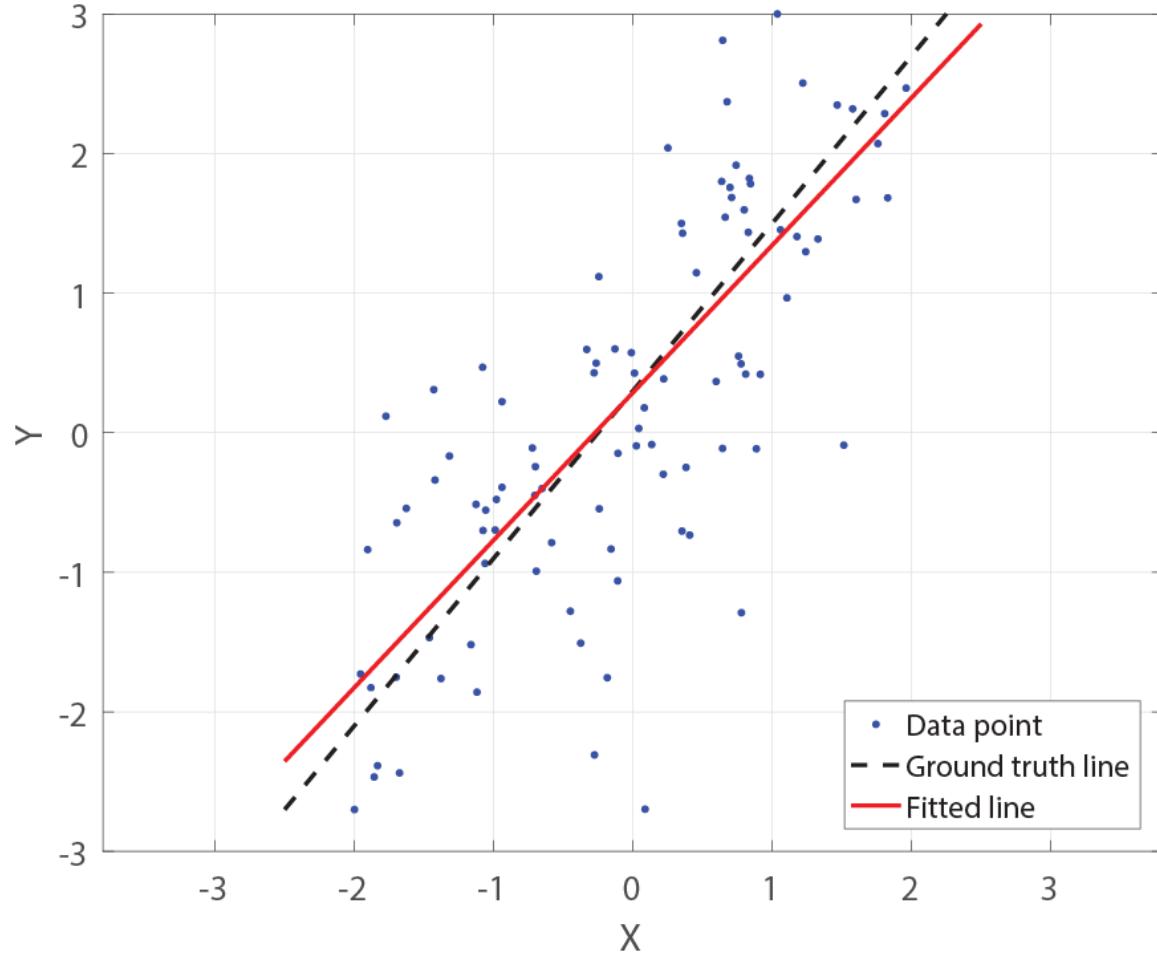
Live capture using a commodity webcam

Thies et al. "Face2Face: Realtime Face Capture and Reenactment of RGB Videos"

Announcement

- HW #2 due today
- HW #2 short presentation on next Thursday (share your panorama!)
- HW #3 will be out next Tuesday

Line Fitting ($Ax=b$)



Total error: $E = \sum_{i=1}^n (v_i - (mu_i + d))^2$

$$\begin{aligned} E &= (\mathbf{Ax} - \mathbf{b})^\top (\mathbf{Ax} - \mathbf{b}) = \|\mathbf{Ax} - \mathbf{b}\|^2 \\ &= \mathbf{x}^\top \mathbf{A}^\top \mathbf{Ax} - 2\mathbf{x}^\top \mathbf{A}^\top \mathbf{b} + \mathbf{b}^\top \mathbf{b} \end{aligned}$$

$$\begin{bmatrix} u_1 & 1 \\ u_2 & 1 \\ \vdots & \vdots \\ u_n & 1 \end{bmatrix} \mathbf{A} \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} \approx \begin{bmatrix} m \\ d \end{bmatrix} \mathbf{b}$$

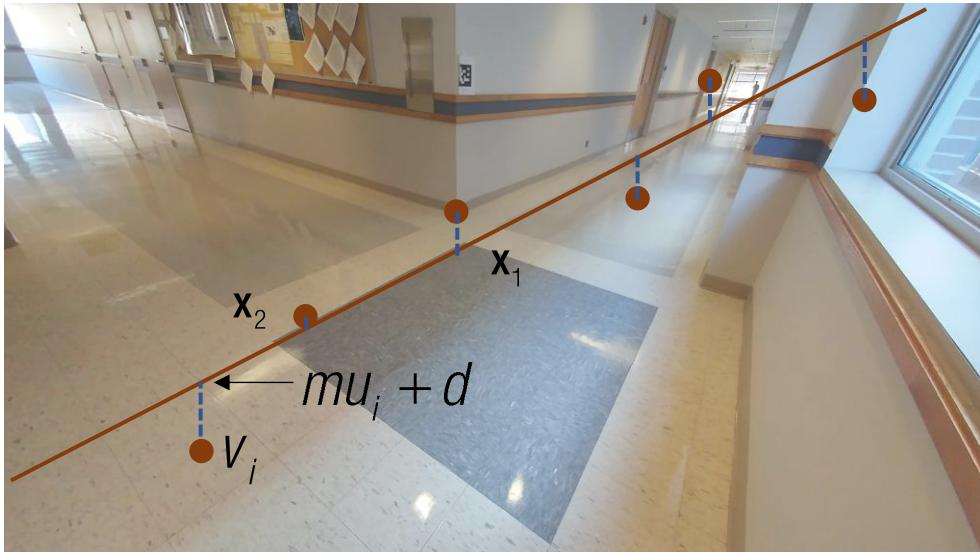
$$\frac{\partial E}{\partial \mathbf{x}} = 2\mathbf{A}^\top \mathbf{Ax} - 2\mathbf{A}^\top \mathbf{b} = \mathbf{0}$$

$$\rightarrow \mathbf{A}^\top \mathbf{Ax} = \mathbf{A}^\top \mathbf{b}$$

$$\mathbf{A}^\top \mathbf{A} \mathbf{x} = \mathbf{A}^\top \mathbf{b}$$

$$\mathbf{x} = [\mathbf{A}^\top \mathbf{A}]^{-1} \mathbf{A}^\top \mathbf{b}$$

Line Fitting ($Ax=b$)



Given points: $(u_1, v_1), (u_2, v_2), \dots, (u_n, v_n)$

Find the best line: $(a, b, c) \longrightarrow (\underline{m}, \underline{d})$

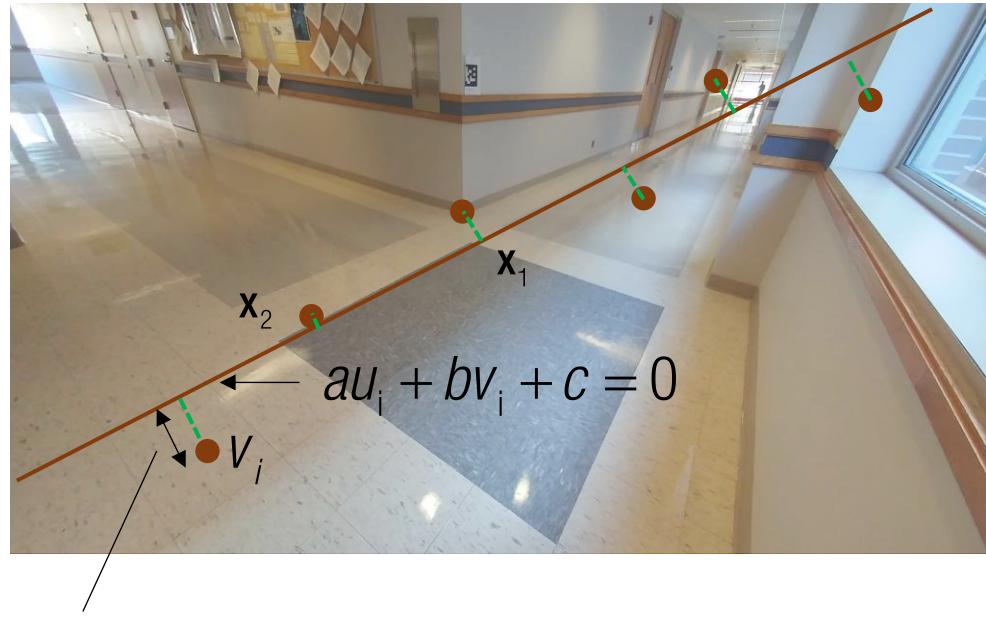
slope y-intercept

$$\begin{array}{l} au_1 + bv_1 + c \approx 0 \\ au_2 + bv_2 + c \approx 0 \\ \vdots \\ au_n + bv_n + c \approx 0 \end{array} \longrightarrow \begin{array}{l} v_1 \approx mu_1 + d \\ v_2 \approx mu_2 + d \\ \vdots \\ v_n \approx mu_n + d \end{array}$$

$$\mathbf{Ax = b}$$

What is different?

Line Fitting ($Ax=0$)



$$d_i = \frac{|au_i + bv_i + c|}{\sqrt{a^2 + b^2}}$$

Given points: $(u_1, v_1), (u_2, v_2), \dots, (u_n, v_n)$

Find the best line: (a, b, c)

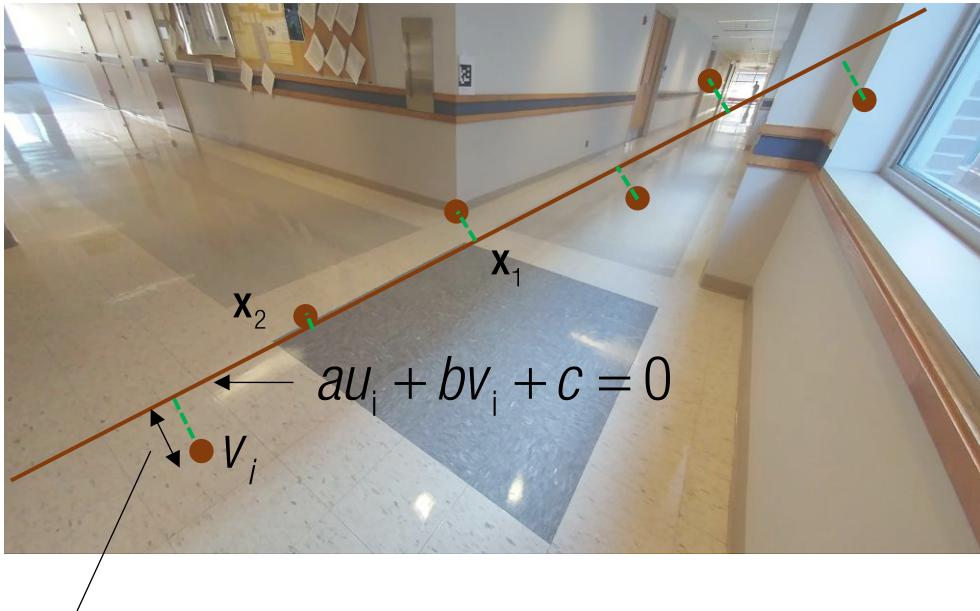
$$\begin{aligned} au_1 + bv_1 + c &\approx 0 \\ au_2 + bv_2 + c &\approx 0 \\ &\vdots \\ au_n + bv_n + c &\approx 0 \end{aligned}$$

$$Ax = 0$$

$$\begin{aligned} v_1 &\approx mu_1 + d \\ v_2 &\approx mu_2 + d \\ &\vdots \\ v_n &\approx mu_n + d \end{aligned}$$

$$Ax = b$$

Line Fitting ($\mathbf{Ax} = 0$)



$$d_i = \frac{|au_i + bv_i + c|}{\sqrt{a^2 + b^2}}$$

Given points: $(u_1, v_1), (u_2, v_2), \dots, (u_n, v_n)$

Find the best line: (a, b, c)

$$\begin{aligned} au_1 + bv_1 + c &\approx 0 \\ au_2 + bv_2 + c &\approx 0 \\ &\vdots \\ au_n + bv_n + c &\approx 0 \end{aligned}$$

$$\mathbf{Ax} = \mathbf{0}$$

Trivial solution: $\mathbf{x} = \mathbf{0}$

$$\underset{\mathbf{x}}{\text{minimize}} \|\mathbf{Ax}\|^2$$

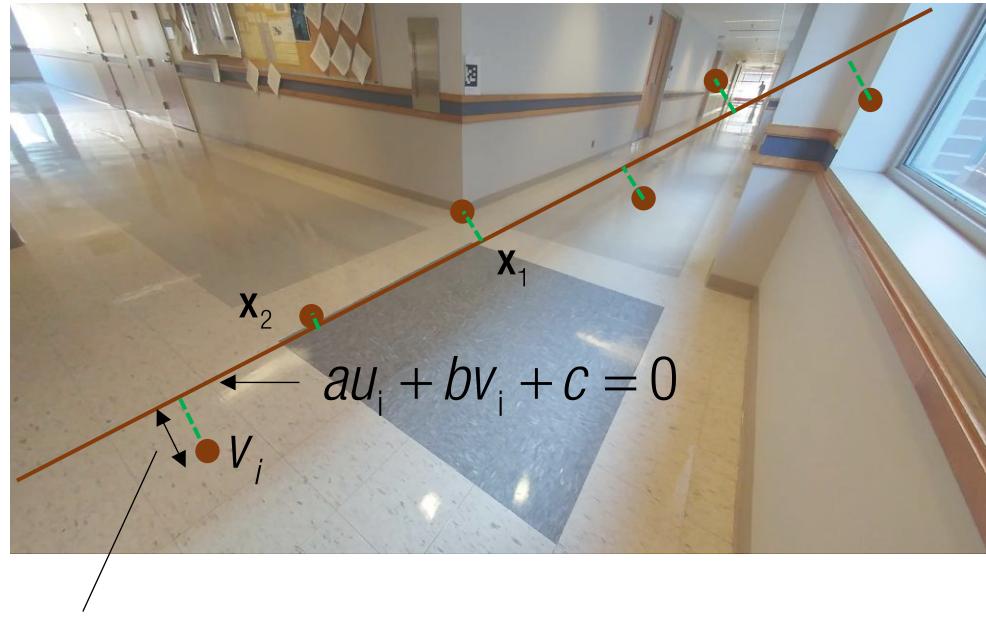
subject to $\|\mathbf{x}\| = 1$

Condition to avoid the trivial solution

$$\begin{aligned} v_1 &\approx mu_1 + d \\ v_2 &\approx mu_2 + d \\ &\vdots \\ v_n &\approx mu_n + d \end{aligned}$$

$$\mathbf{Ax} = \mathbf{b}$$

Line Fitting ($Ax=0$)



$$d_i = \frac{|au_i + bv_i + c|}{\sqrt{a^2 + b^2}}$$

Given points: $(u_1, v_1), (u_2, v_2), \dots, (u_n, v_n)$

Find the best line: (a, b, c)

$$\begin{aligned} au_1 + bv_1 + c &\approx 0 \\ au_2 + bv_2 + c &\approx 0 \\ &\vdots \\ au_n + bv_n + c &\approx 0 \end{aligned}$$

$$Ax = 0$$

$$\begin{aligned} v_1 &\approx mu_1 + d \\ v_2 &\approx mu_2 + d \\ &\vdots \\ v_n &\approx mu_n + d \end{aligned}$$

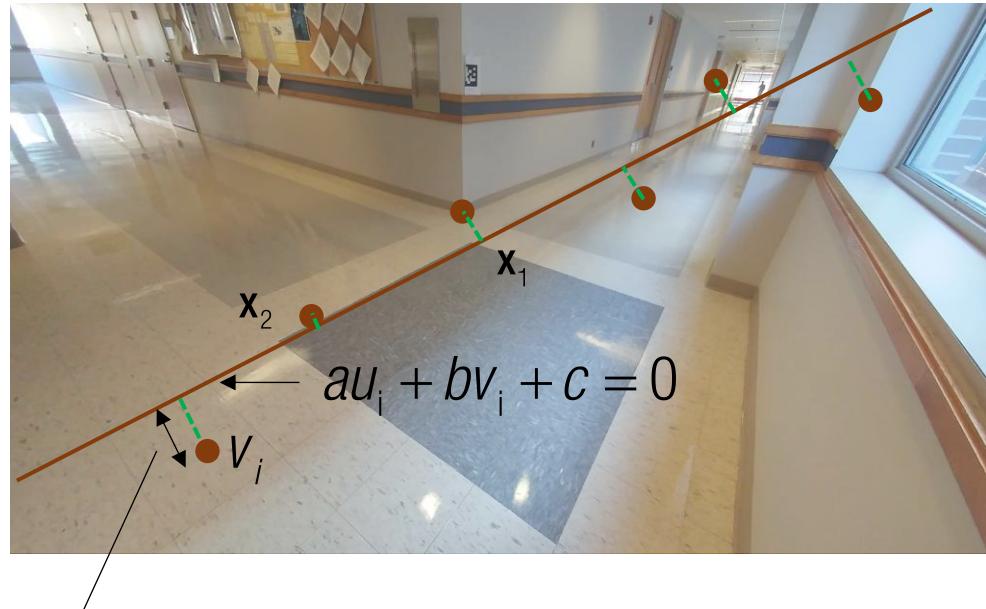
$$Ax = b$$

Trivial solution: $\mathbf{x} = \mathbf{0}$

$$\underset{\mathbf{x}}{\text{minimize}} \|\mathbf{Ax}\|^2 \quad \text{subject to } \|\mathbf{x}\| = 1$$

If A is $m \times n$ matrix where $m < n$,

Line Fitting ($Ax=0$)



$$d_i = \frac{|au_i + bv_i + c|}{\sqrt{a^2 + b^2}}$$

Given points: $(u_1, v_1), (u_2, v_2), \dots, (u_n, v_n)$

Find the best line: (a, b, c)

$$\begin{aligned} au_1 + bv_1 + c &\approx 0 \\ au_2 + bv_2 + c &\approx 0 \\ &\vdots \\ au_n + bv_n + c &\approx 0 \end{aligned}$$

$$Ax = 0$$

$$\begin{aligned} v_1 &\approx mu_1 + d \\ v_2 &\approx mu_2 + d \\ &\vdots \\ v_n &\approx mu_n + d \end{aligned}$$

$$Ax = b$$

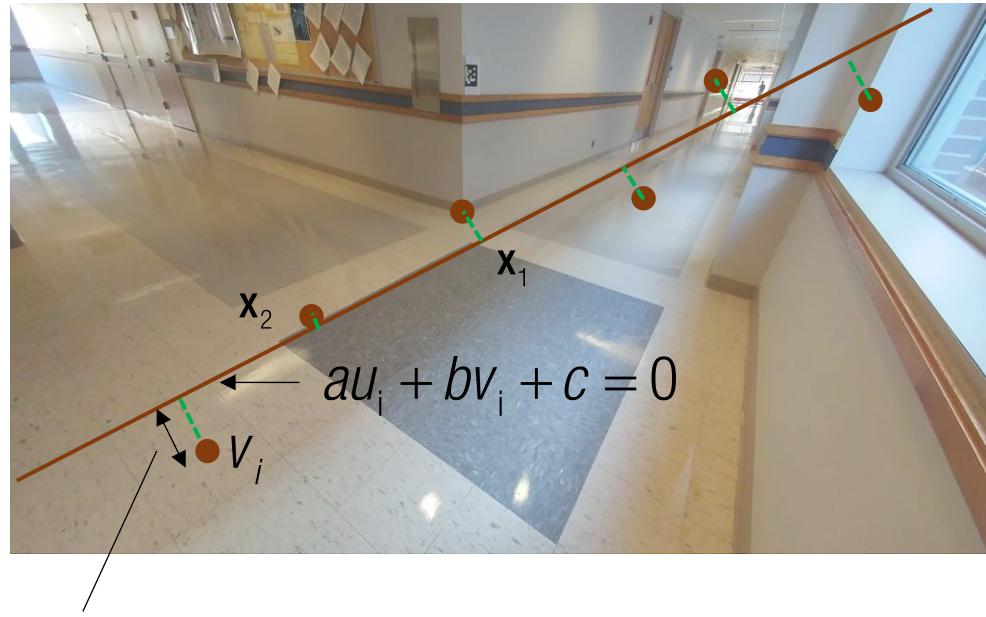
Trivial solution: $\mathbf{x} = \mathbf{0}$

$$\underset{\mathbf{x}}{\text{minimize}} \|\mathbf{Ax}\|^2 \quad \text{subject to } \|\mathbf{x}\| = 1$$

If A is $m \times n$ matrix where $m < n$,

$$\boxed{\mathbf{N}} = \text{null} \left(\boxed{\mathbf{A}} \right)$$

Line Fitting ($Ax=0$)



$$d_i = \frac{|au_i + bv_i + c|}{\sqrt{a^2 + b^2}}$$

Given points: $(u_1, v_1), (u_2, v_2), \dots, (u_n, v_n)$

Find the best line: (a, b, c)

$$\begin{aligned} au_1 + bv_1 + c &\approx 0 \\ au_2 + bv_2 + c &\approx 0 \\ &\vdots \\ au_n + bv_n + c &\approx 0 \end{aligned}$$

$$Ax = 0$$

$$\begin{aligned} v_1 &\approx mu_1 + d \\ v_2 &\approx mu_2 + d \\ &\vdots \\ v_n &\approx mu_n + d \end{aligned}$$

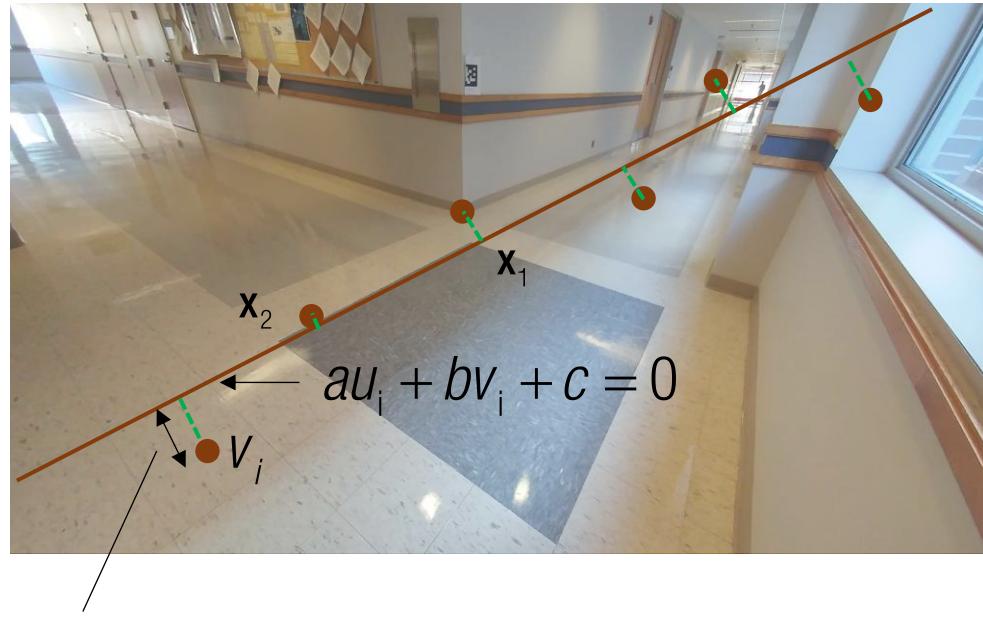
$$Ax = b$$

Trivial solution: $\mathbf{x} = \mathbf{0}$

$$\underset{\mathbf{x}}{\text{minimize}} \|Ax\|^2 \quad \text{subject to } \|\mathbf{x}\| = 1$$

If A is $m \times n$ matrix where $m > n$,

Line Fitting ($Ax=0$)



$$d_i = \frac{|au_i + bv_i + c|}{\sqrt{a^2 + b^2}}$$

Given points: $(u_1, v_1), (u_2, v_2), \dots, (u_n, v_n)$

Find the best line: (a, b, c)

$$\begin{aligned} au_1 + bv_1 + c &\approx 0 \\ au_2 + bv_2 + c &\approx 0 \\ &\vdots \\ au_n + bv_n + c &\approx 0 \end{aligned}$$

$$Ax = 0$$

Trivial solution: $\mathbf{x} = \mathbf{0}$

$$\underset{\mathbf{x}}{\text{minimize}} \|\mathbf{Ax}\|^2 \quad \text{subject to } \|\mathbf{x}\| = 1$$

If A is $m \times n$ matrix where $m > n$, $\boxed{\quad} = \text{approx.null}\left(\boxed{A}\right)$

Nullspace

eqs < # unknowns

$$\begin{array}{c|c|c} \text{A} & \text{x} & \text{0} \\ m \times n & n \times 1 & \\ \hline & = & \end{array}$$

Nullspace

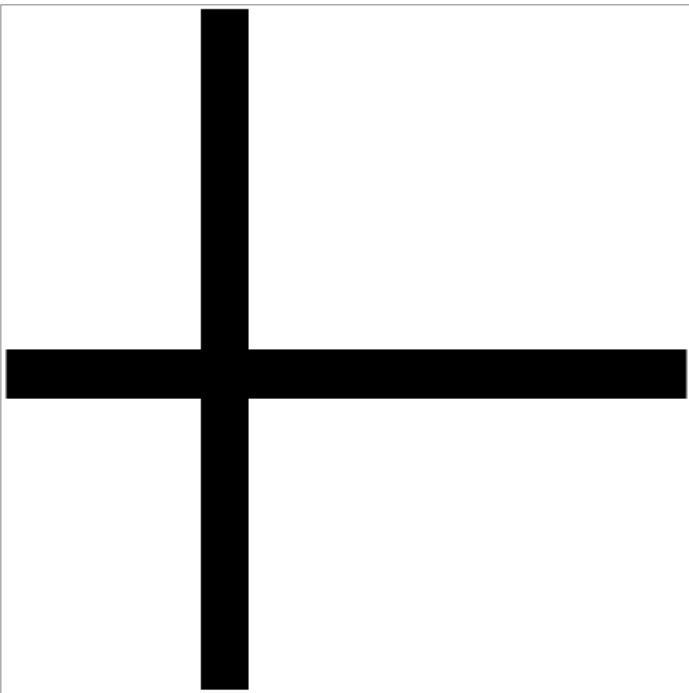
eqs < # unknowns

There exist at least $n-m$ vectors that makes $\mathbf{Ax}=0$.

$$\begin{array}{c|c|c} \mathbf{A} & \mathbf{x} & \mathbf{0} \\ m \times n & n \times 1 & \\ \hline m < n & & \end{array} =$$

$$\longrightarrow \mathbf{x} = \text{null}(\mathbf{A})$$

More SVD

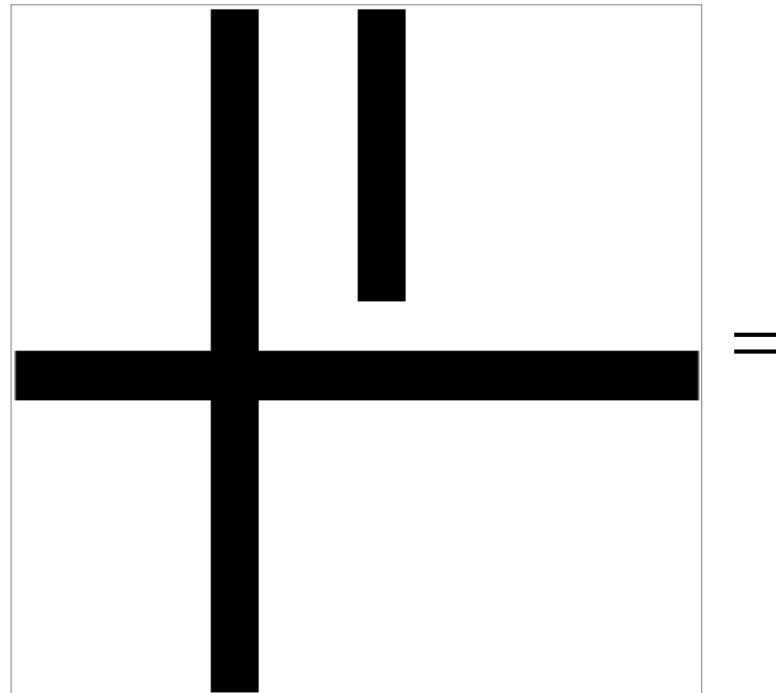


14x14

$$= \begin{matrix} & \text{red square} \\ \text{blue rectangle} & \end{matrix}$$



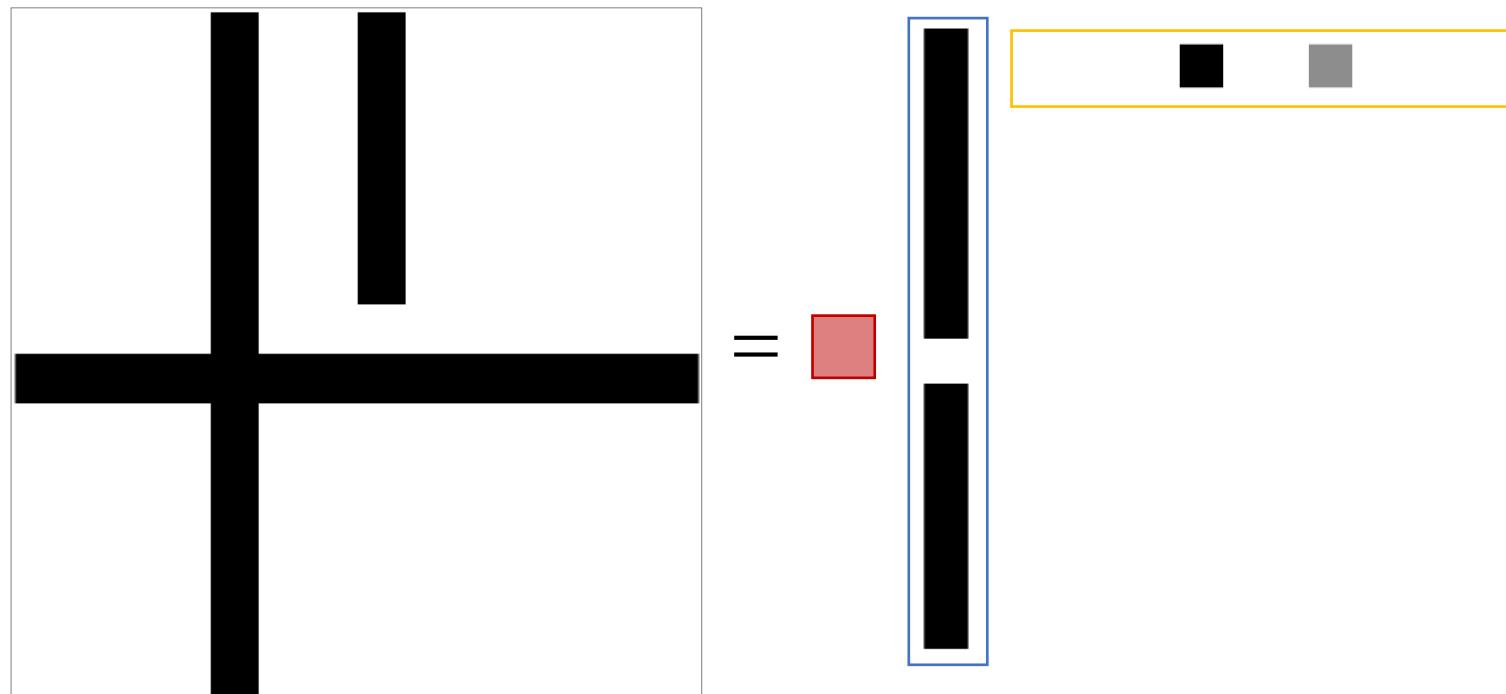
More SVD



=

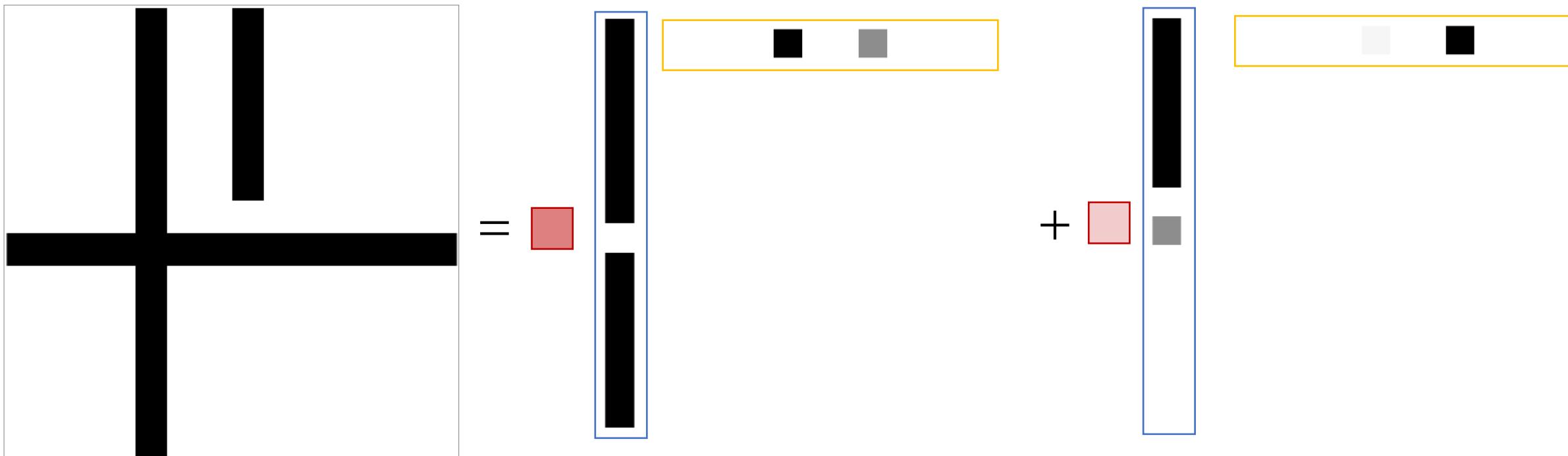
14x14

More SVD



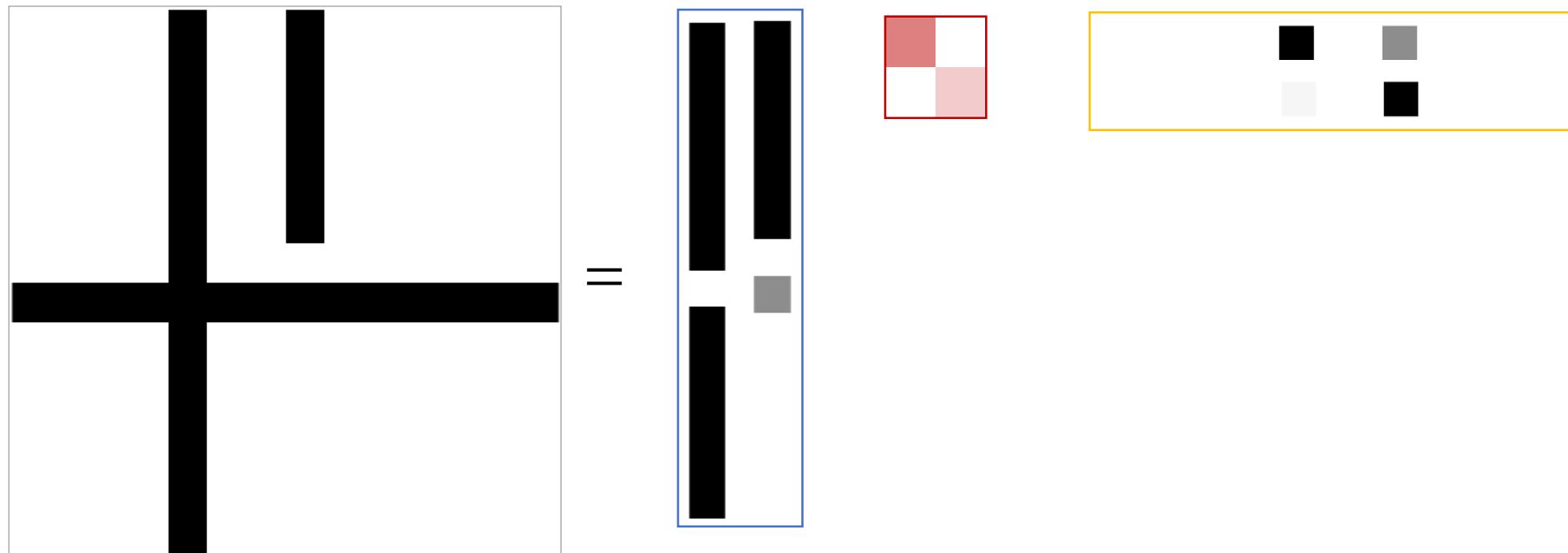
14x14

More SVD



14x14

More SVD



14x14

Singular Value Decomposition (SVD)

eqs < # unknowns

There exist at least $n-m$ vectors that makes $\mathbf{Ax}=0$.

$$\begin{array}{c|c|c} \mathbf{A} & \mathbf{x} & = \\ \text{m x n} & \text{n x 1} & \mathbf{0} \\ \hline & & \end{array} \quad \begin{array}{c|c} \mathbf{A} & = \\ \text{m x n} & \end{array}$$

$$\longrightarrow \mathbf{x} = \text{null}(\mathbf{A})$$

Singular Value Decomposition (SVD)

eqs < # unknowns

There exist at least $n-m$ vectors that makes $\mathbf{Ax}=0$.

$$\begin{array}{c|c|c} \text{A} & \mathbf{x} & \mathbf{0} \\ m \times n & n \times 1 & \\ \hline & = & \end{array}$$

Column space

$$\begin{array}{c|c} \text{A} & \mathbf{U} \\ m \times n & m \times m \\ \hline & = \end{array}$$

Orthogonal matrix

$$\longrightarrow \mathbf{x} = \text{null}(\mathbf{A})$$

Singular Value Decomposition (SVD)

eqs < # unknowns

There exist at least $n-m$ vectors that makes $\mathbf{Ax}=0$.

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} \quad \begin{matrix} \mathbf{x} \\ n \times 1 \end{matrix} = \begin{matrix} \mathbf{0} \\ m \times 1 \end{matrix}$$

$$\longrightarrow \mathbf{x} = \text{null}(\mathbf{A})$$

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} = \begin{matrix} \text{Column space} \\ \mathbf{U} \\ m \times m \end{matrix} \quad \begin{matrix} \mathbf{D} \\ m \times n \end{matrix}$$

Orthogonal matrix Diagonal matrix

Singular Value Decomposition (SVD)

eqs < # unknowns

There exist at least $n-m$ vectors that makes $\mathbf{Ax}=0$.

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} \quad \begin{matrix} \mathbf{x} \\ n \times 1 \end{matrix} = \begin{matrix} \mathbf{0} \\ 1 \times n \end{matrix}$$

$$\longrightarrow \mathbf{x} = \text{null}(\mathbf{A})$$

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} = \begin{matrix} \mathbf{U} \\ m \times m \end{matrix} \begin{matrix} \mathbf{D} \\ m \times n \end{matrix} \begin{matrix} \mathbf{V}^T \\ n \times n \end{matrix}$$

Column space Orthogonal matrix Diagonal matrix Orthogonal matrix

Row space

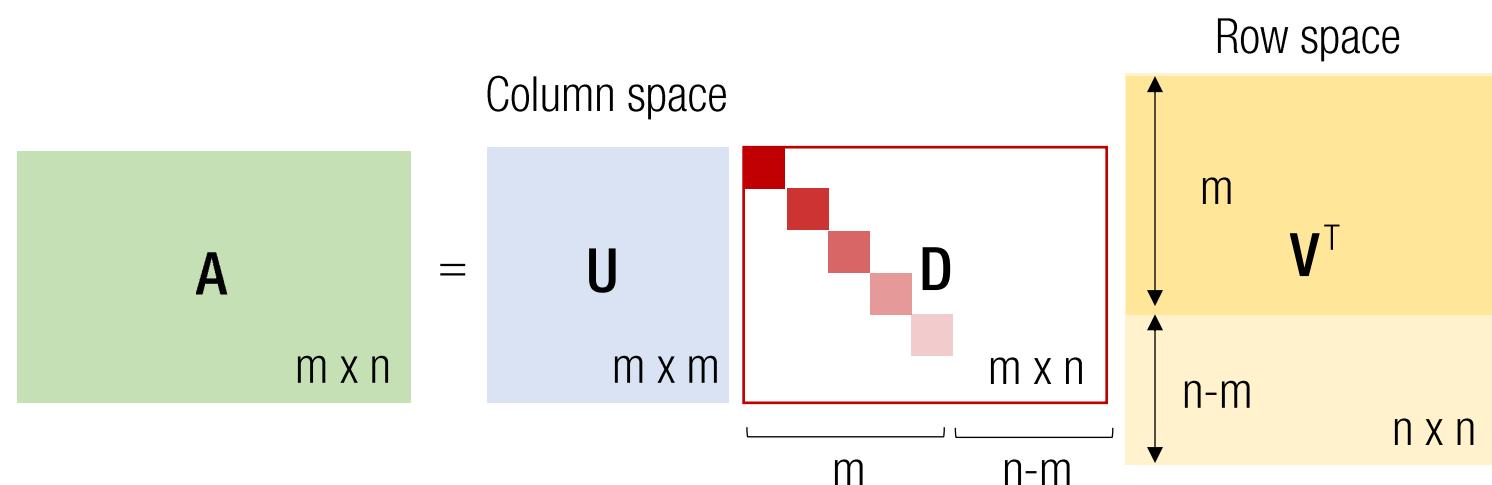
Singular Value Decomposition (SVD)

eqs < # unknowns

There exist at least $n-m$ vectors that makes $\mathbf{Ax}=0$.

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} \quad \begin{matrix} \mathbf{x} \\ n \times 1 \end{matrix} = \begin{matrix} \mathbf{0} \end{matrix}$$

$$\longrightarrow \mathbf{x} = \text{null}(\mathbf{A})$$



Singular Value Decomposition (SVD)

eqs < # unknowns

There exist at least $n-m$ vectors that makes $\mathbf{Ax}=0$.

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} \quad \begin{matrix} \mathbf{x} \\ n \times 1 \end{matrix} = \begin{matrix} \mathbf{0} \end{matrix}$$

$$\longrightarrow \mathbf{x} = \text{null}(\mathbf{A})$$

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} = \begin{matrix} \mathbf{U} \\ m \times m \end{matrix} \begin{matrix} \mathbf{D} \\ m \times n \end{matrix} \begin{matrix} \mathbf{V}^T \\ n \times n \end{matrix}$$

Column space

Row space

m $n-m$

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} \quad \begin{matrix} \mathbf{V}_{m+1:n} \\ n-m \times n \end{matrix} = \begin{matrix} \mathbf{0} \end{matrix} \quad \mathbf{V}_{m+1:n} = \text{null}(\mathbf{A})$$

Singular Value Decomposition (SVD)

eqs < # unknowns

There exist at least $n-m$ vectors that makes $\mathbf{Ax}=0$.

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} \quad \begin{matrix} \mathbf{x} \\ n \times 1 \end{matrix} = \begin{matrix} \mathbf{0} \end{matrix}$$

$$\longrightarrow \mathbf{x} = \text{null}(\mathbf{A})$$

$$\begin{matrix} \mathbf{A} \\ m \times n \end{matrix} = \begin{matrix} \mathbf{U} \\ m \times m \end{matrix} \begin{matrix} \mathbf{D} \\ m \times n \end{matrix} \begin{matrix} \mathbf{V}^T \\ n \times n \end{matrix}$$

Column space

Row space

m $n-m$

$$\begin{matrix} \mathbf{A} \\ \mathbf{V}_{:, \text{end}} \end{matrix} = \begin{matrix} \mathbf{0} \end{matrix} \quad \mathbf{V}_{:, \text{end}} = \text{null}(\mathbf{A})$$

For a unique solution, \mathbf{A} should be $m \times (m+1)$

Singular Value Decomposition (SVD)

eqs > # unknowns

There exist no nullspace of \mathbf{A} .

$$\begin{array}{c|c|c} \mathbf{A} & \mathbf{x} & \approx & \mathbf{0} \\ m \times n & n \times 1 & & \\ \hline & m > n & & \end{array}$$

$$\begin{array}{c|c|c} \mathbf{A} & = & \mathbf{U} & \text{Column space} \\ m \times n & & m \times n & \\ \hline & & \mathbf{D} & \text{Row space} \\ & & n \times n & \\ & & \mathbf{V}^T & n \times n \end{array}$$

$$\underset{\mathbf{x}}{\text{minimize}} \|\mathbf{Ax}\|^2 \text{ subject to } \|\mathbf{x}\| = 1$$

Singular Value Decomposition (SVD)

eqs > # unknowns

There exist no nullspace of \mathbf{A} .

$$\begin{array}{c|c|c} \mathbf{A} & \mathbf{x} & \approx \mathbf{0} \\ m \times n & n \times 1 & \\ \hline & m > n & \end{array}$$

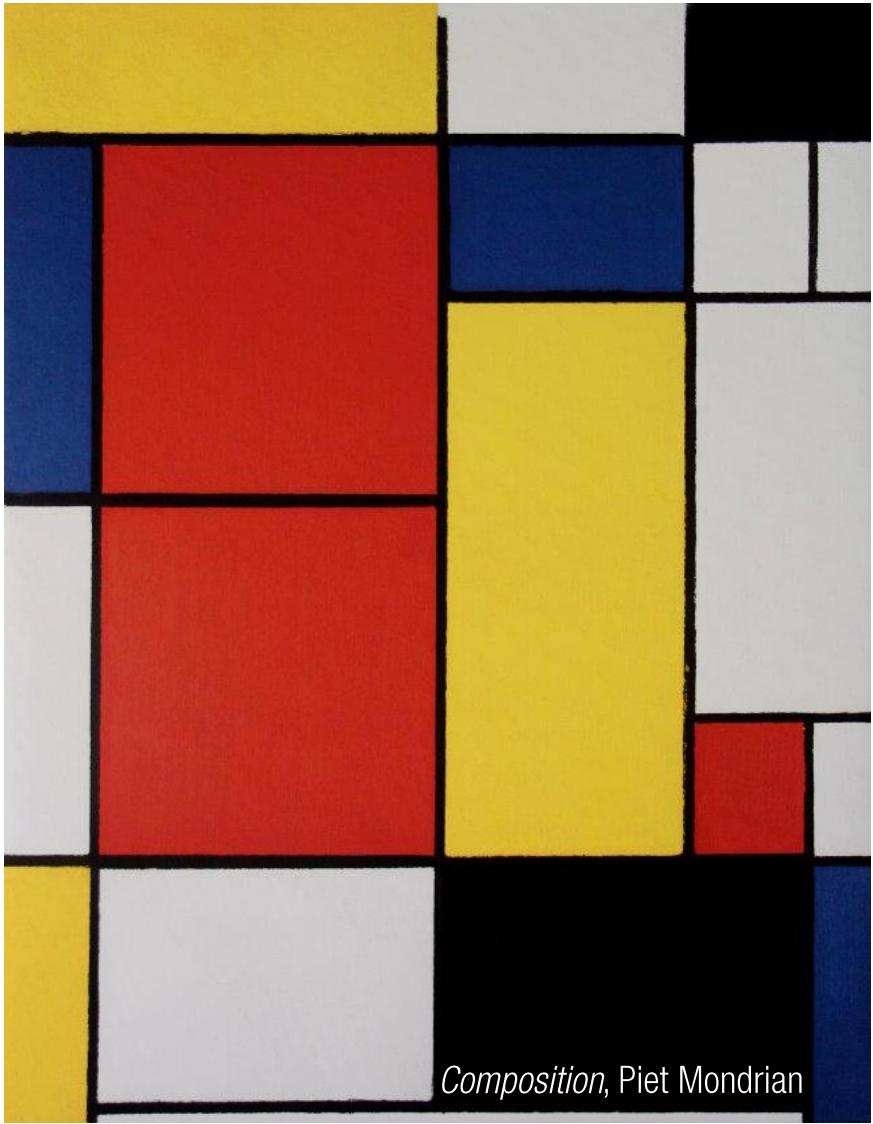
$$\underset{\mathbf{x}}{\text{minimize}} \|\mathbf{Ax}\|^2 \text{ subject to } \|\mathbf{x}\| = 1$$

$$\begin{array}{ccccc} \mathbf{A} & = & \mathbf{U} & \begin{array}{c} \text{Column space} \\ \mathbf{D} \\ \mathbf{n} \times \mathbf{n} \end{array} & \mathbf{V}^T \\ m \times n & & m \times n & & n \times n \\ & & & \leftarrow \text{Last row} & \end{array}$$

Approximated nullspace of \mathbf{A} :

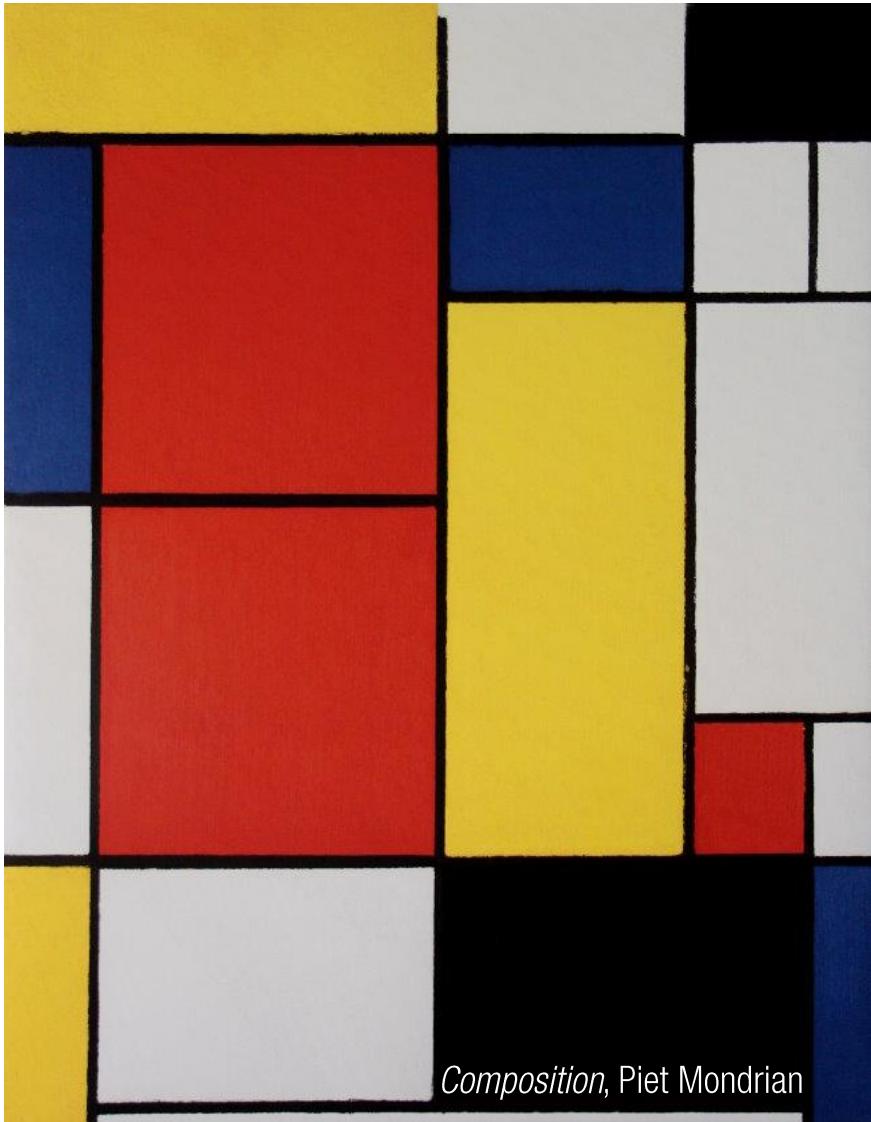
$$\mathbf{V}_{:, \text{end}}$$

Mondrian Painting SVD



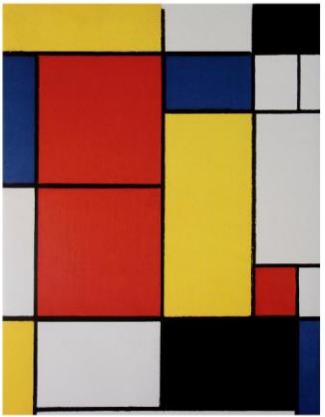
Composition, Piet Mondrian

Mondrian Painting SVD

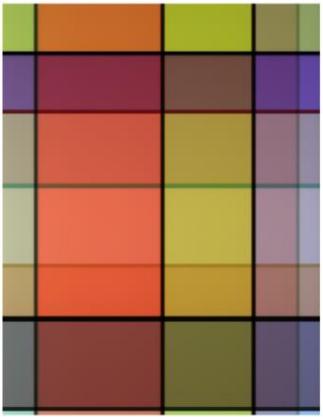


$$= \begin{matrix} U & D & V^T \\ m \times n & n \times n & n \times n \end{matrix}$$

Mondrian Painting SVD Approximation



Ground truth



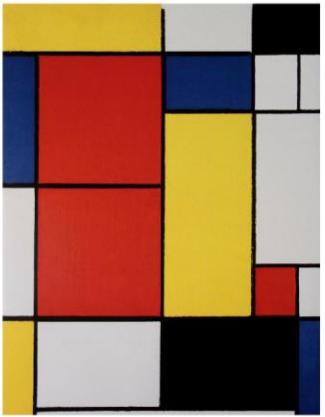
Number of basis: 1

MondrianSVD.m

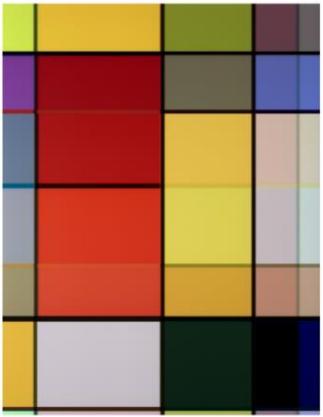


$$A = U \quad D \quad V^T$$

Mondrian Painting SVD Approximation

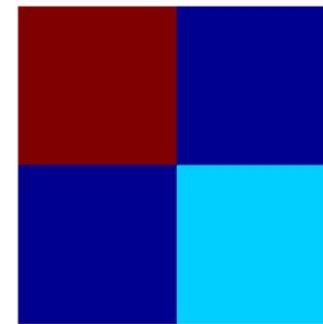


Ground truth



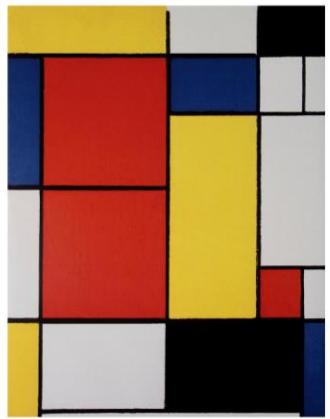
Number of basis: 2

MondrianSVD.m

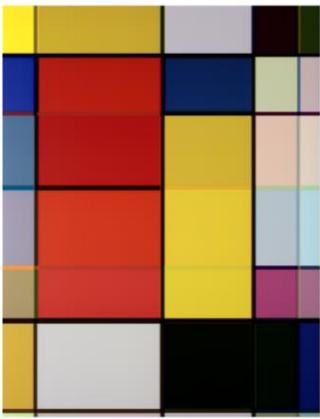


$$A = U D V^T$$

Mondrian Painting SVD Approximation

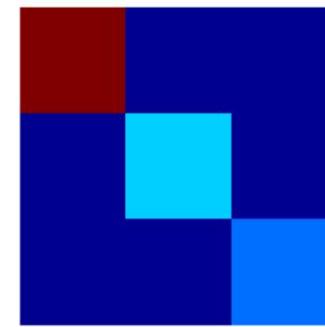


Ground truth



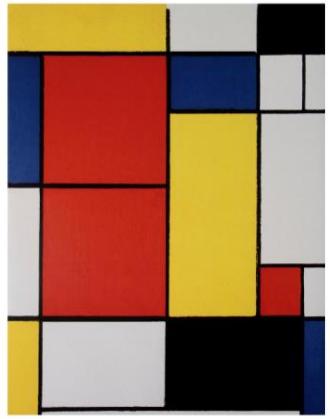
Number of basis: 3

MondrianSVD.m

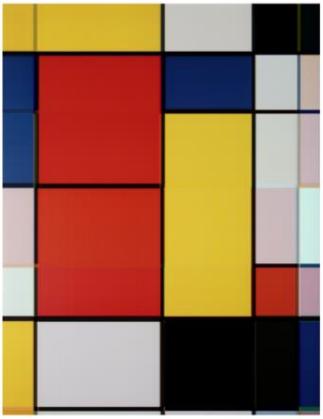


$$A = U D V^T$$

Mondrian Painting SVD Approximation

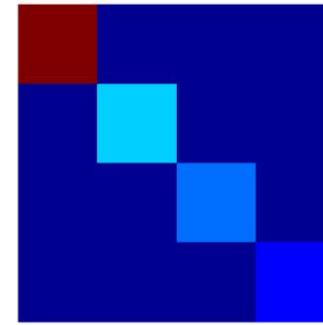


Ground truth



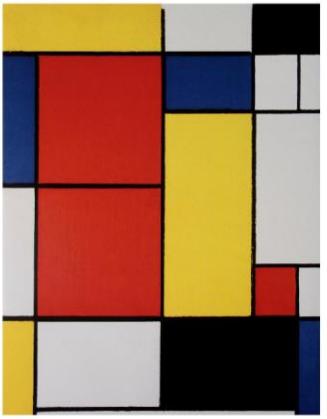
Number of basis: 4

MondrianSVD.m

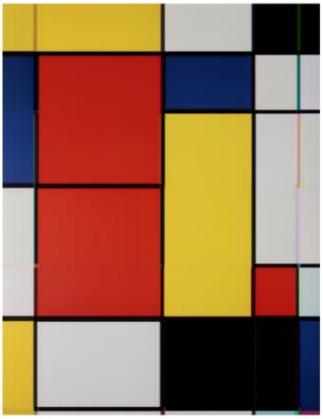


$$A = U \quad D \quad V^T$$

Mondrian Painting SVD Approximation

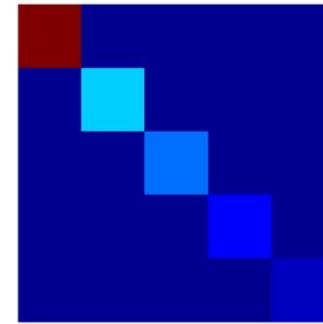


Ground truth



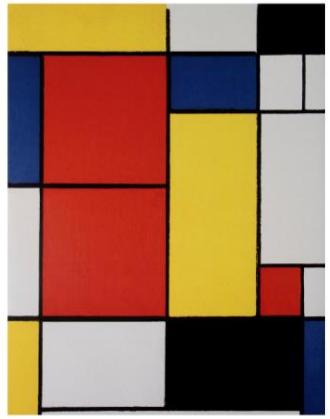
Number of basis: 5

MondrianSVD.m

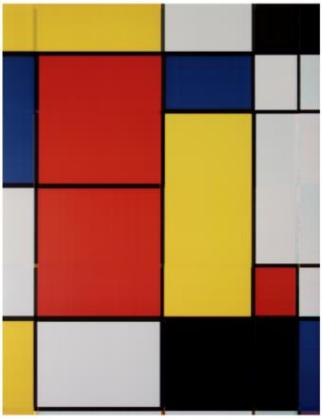


$$A = U D V^T$$

Mondrian Painting SVD Approximation

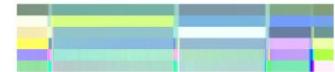
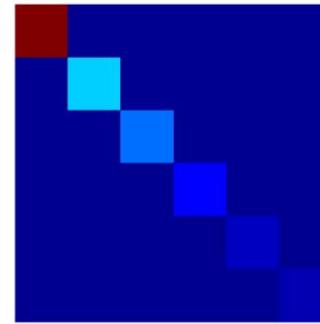
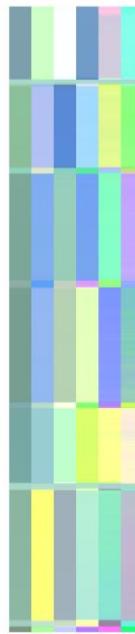


Ground truth



Number of basis: 6

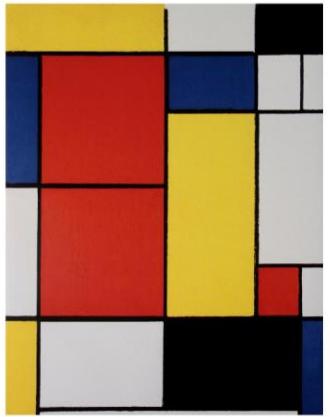
MondrianSVD.m



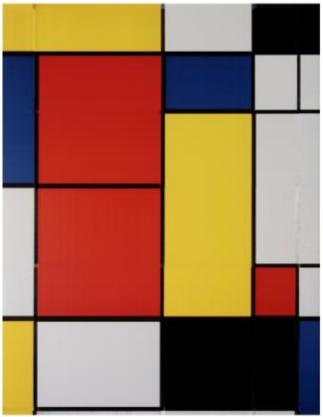
$$A = U \quad D \quad V^T$$

Mondrian Painting SVD Approximation

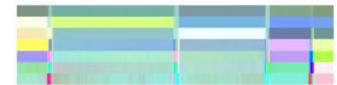
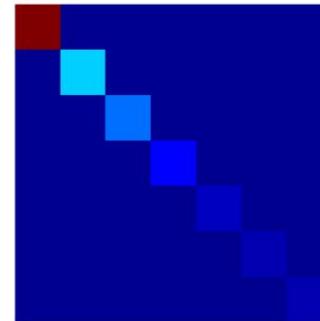
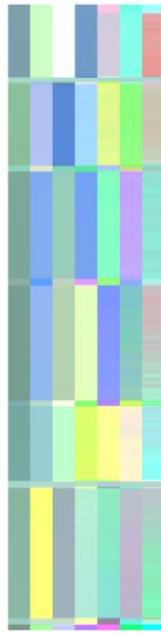
MondrianSVD.m



Ground truth

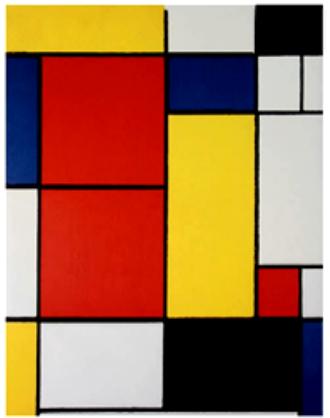


Number of basis: 7

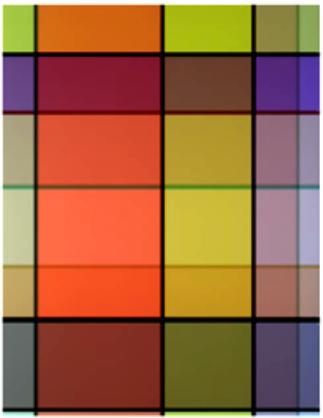


$$A = U \quad D \quad V^T$$

Mondrian Painting SVD Approximation

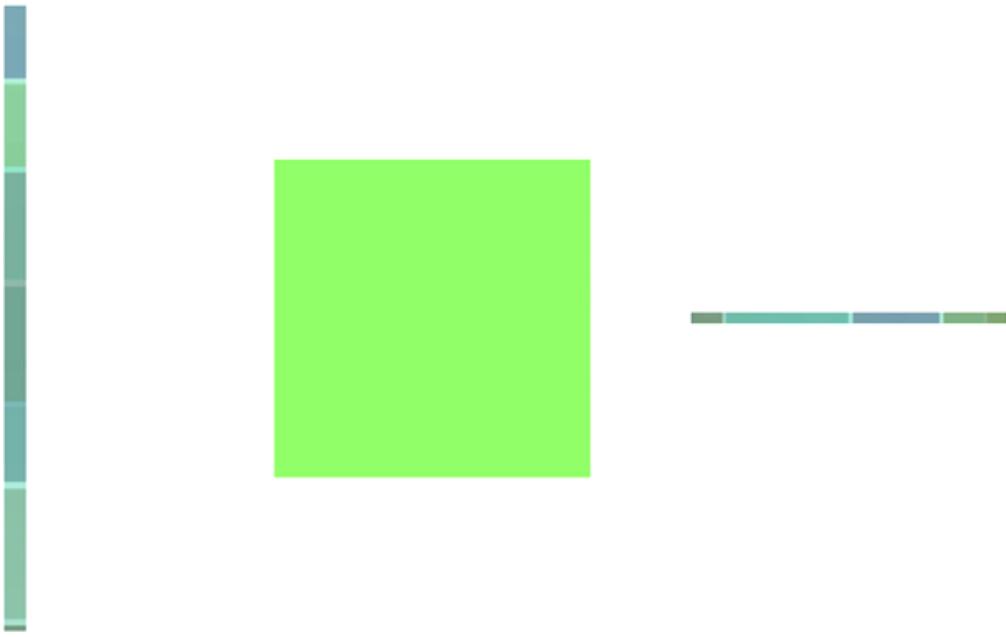


Ground truth



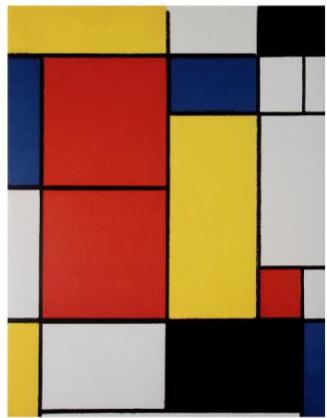
Number of basis: 1

MondrianSVD.m

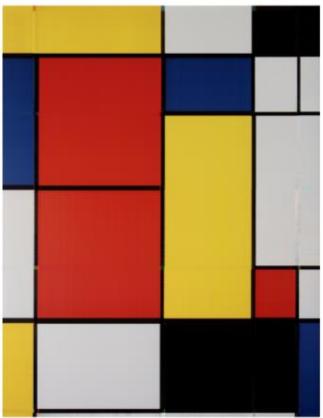


$$A = U D V^T$$

Reconstruction Error



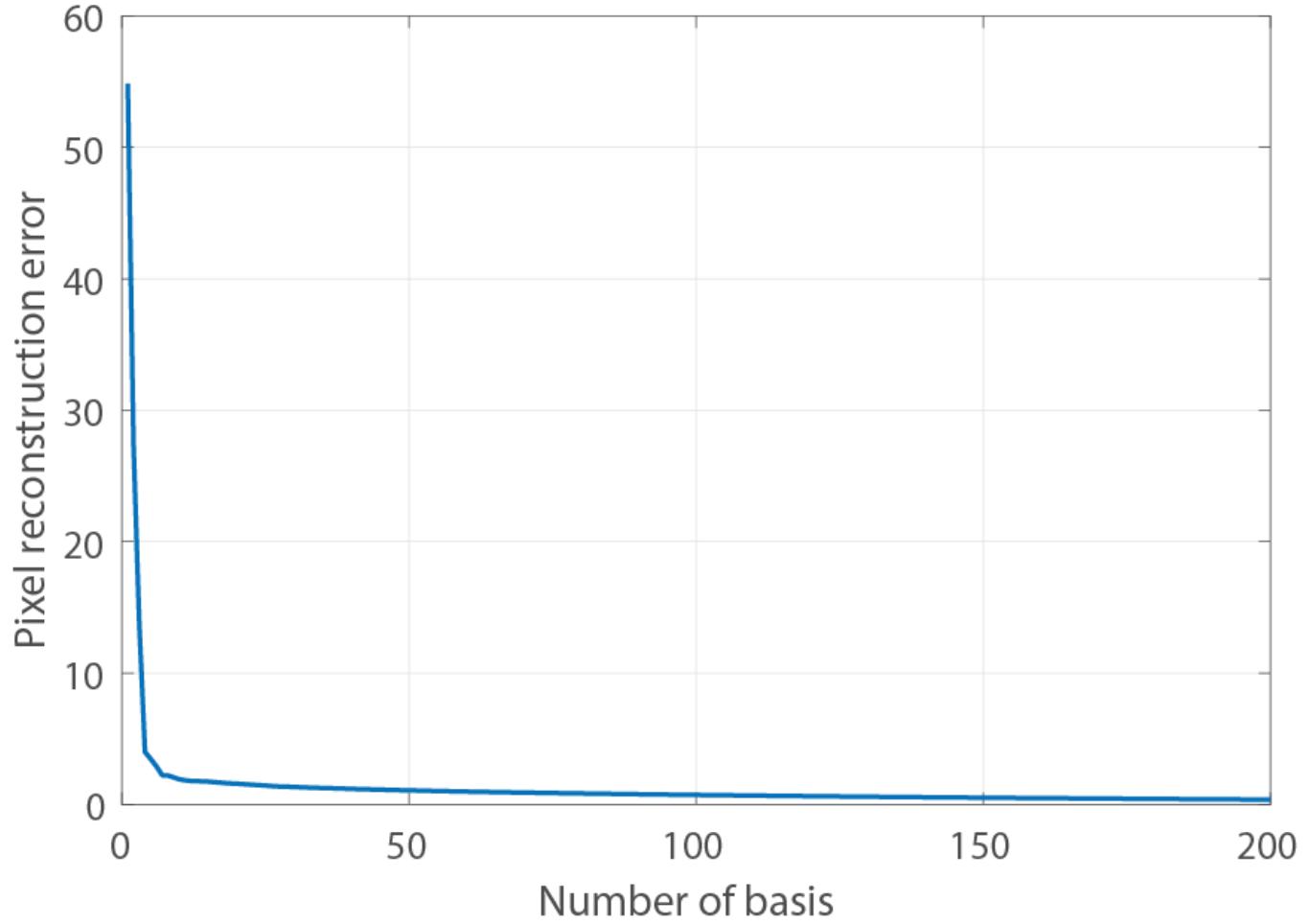
Ground truth



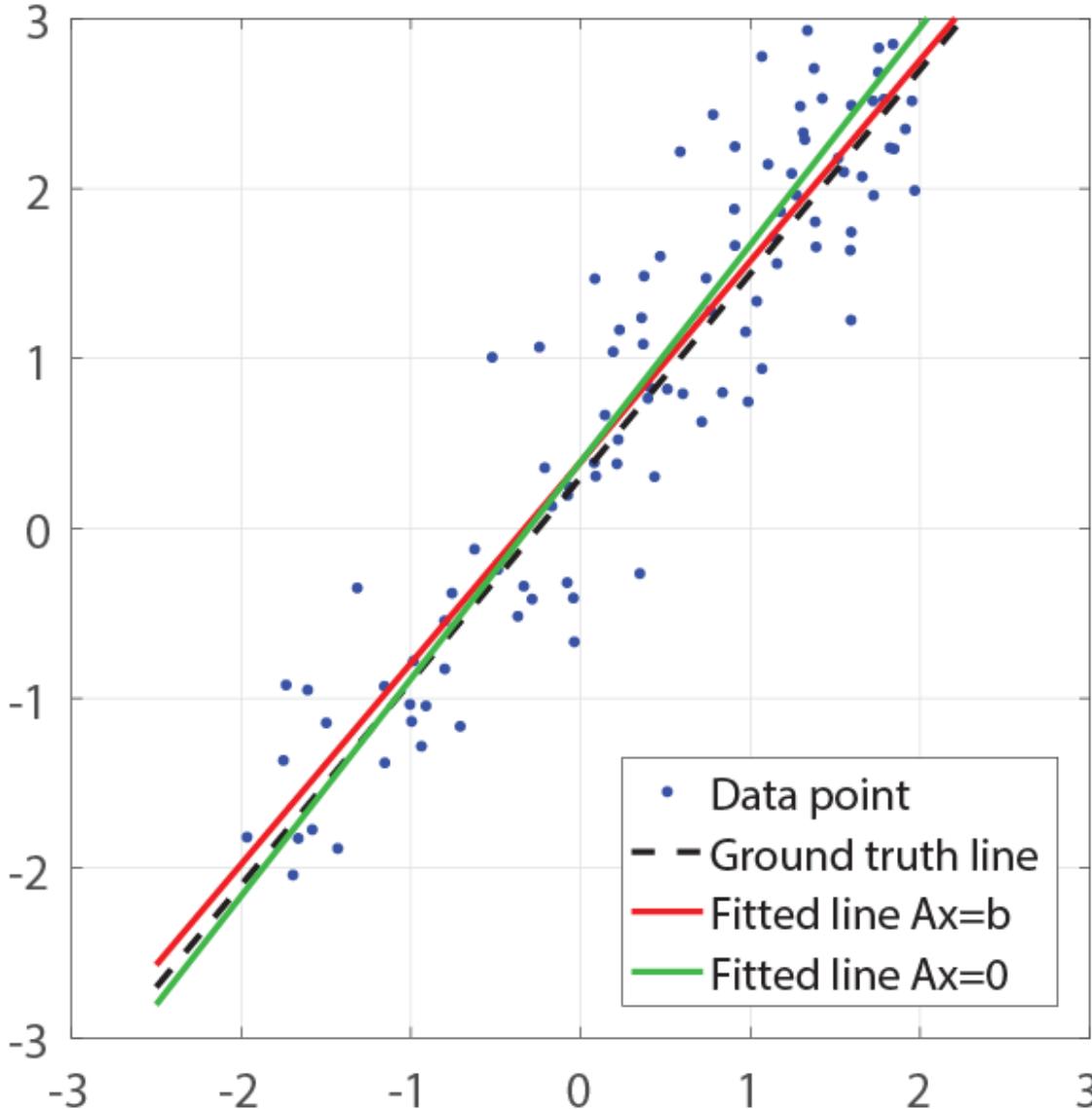
Number of basis: 7

A

MondrianSVD.m



Line Fitting ($Ax=b$)



Given points: $(u_1, v_1), (u_2, v_2), \dots, (u_n, v_n)$

Find the best line: (a, b, c)

$$au_1 + bv_1 + c \approx 0$$

$$au_2 + bv_2 + c \approx 0$$

⋮

$$au_n + bv_n + c \approx 0$$

$$v_1 \approx mu_1 + d$$

$$v_2 \approx mu_2 + d$$

⋮

$$v_n \approx mu_n + d$$

$$\begin{bmatrix} u_1 & v_1 & 1 \\ u_2 & v_2 & 1 \\ \vdots & \vdots & \vdots \\ u_n & v_n & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} \approx \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} u_1 & 1 \\ u_2 & 1 \\ \vdots & \vdots \\ u_n & 1 \end{bmatrix} \begin{bmatrix} m \\ d \end{bmatrix} \approx \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$



How to compute homography?

Homography Computation



$$\lambda \begin{bmatrix} v_x \\ v_y \\ 1 \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} u_x \\ u_y \\ 1 \end{bmatrix}$$

Homography Computation



$$v_x = \frac{h_{11}u_x + h_{12}u_y + h_{13}}{h_{31}u_x + h_{32}u_y + h_{33}}$$

$$v_y = \frac{h_{21}u_x + h_{22}u_y + h_{23}}{h_{31}u_x + h_{32}u_y + h_{33}}$$

$$\begin{aligned} \rightarrow & h_{11}u_x + h_{12}u_y + h_{13} + h_{31}u_x v_x + h_{32}u_y v_x + h_{33}v_x = 0 \\ & h_{21}u_x + h_{22}u_y + h_{23} + h_{31}u_x v_y + h_{32}u_y v_y + h_{33}v_y = 0 \end{aligned}$$

$$\lambda \begin{bmatrix} v_x \\ v_y \\ 1 \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} u_x \\ u_y \\ 1 \end{bmatrix}$$

Unknowns: h_{11}, \dots, h_{33}

Equations: 2 per correspondence

Homography Computation



$$\lambda \begin{bmatrix} v_x \\ v_y \\ 1 \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} u_x \\ u_y \\ 1 \end{bmatrix}$$

$$\begin{aligned}
 v_x &= \frac{h_{11}u_x + h_{12}u_y + h_{13}}{h_{31}u_x + h_{32}u_y + h_{33}} \\
 v_y &= \frac{h_{21}u_x + h_{22}u_y + h_{23}}{h_{31}u_x + h_{32}u_y + h_{33}} \\
 \rightarrow \quad h_{11}u_x + h_{12}u_y + h_{13} - h_{31}u_xv_x - h_{32}u_yv_x - h_{33}v_x &= 0 \\
 h_{21}u_x + h_{22}u_y + h_{23} - h_{31}u_xv_y - h_{32}u_yv_y - h_{33}v_y &= 0
 \end{aligned}$$

$$\rightarrow \begin{bmatrix} u_x & u_y & 1 & -u_xv_x & -u_yv_x & -v_x \\ & & u_x & u_y & 1 & -u_xv_y & -u_yv_y & -v_y \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{12} \\ h_{13} \\ h_{21} \\ h_{22} \\ h_{23} \\ h_{31} \\ h_{32} \\ h_{33} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Homography Computation



$$\lambda \begin{bmatrix} v_x \\ v_y \\ 1 \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} u_x \\ u_y \\ 1 \end{bmatrix}$$

$$v_x = \frac{h_{11}u_x + h_{12}u_y + h_{13}}{h_{31}u_x + h_{32}u_y + h_{33}}$$

$$v_y = \frac{h_{21}u_x + h_{22}u_y + h_{23}}{h_{31}u_x + h_{32}u_y + h_{33}}$$

$$\rightarrow \begin{aligned} h_{11}u_x + h_{12}u_y + h_{13} - h_{31}u_xv_x - h_{32}u_yv_x - h_{33}v_x &= 0 \\ h_{21}u_x + h_{22}u_y + h_{23} - h_{31}u_xv_y - h_{32}u_yv_y - h_{33}v_y &= 0 \end{aligned}$$

$$\rightarrow \begin{bmatrix} u_x & u_y & 1 & -u_xv_x & -u_yv_x & -v_x \\ & & u_x & u_y & -u_xv_y & -u_yv_y & -v_y \end{bmatrix} \mathbf{A} \begin{bmatrix} h_{11} \\ h_{12} \\ h_{13} \\ h_{21} \\ h_{22} \\ h_{23} \\ h_{31} \\ h_{32} \\ h_{33} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

\mathbf{A} 2x9

Homography Computation

How many correspondences are needed?



$$\lambda \begin{bmatrix} v_x \\ v_y \\ 1 \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} u_x \\ u_y \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} u_x & u_y & 1 \\ u_x & u_y & 1 \end{bmatrix} \mathbf{A} \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} -u_x v_x & -u_y v_x & -v_x \\ -u_x v_y & -u_y v_y & -v_y \\ 0 & 0 \end{bmatrix} = \mathbf{0}$$

2x9

Homography Computation

How many correspondences are needed? 4



$$\lambda \begin{bmatrix} v_x \\ v_y \\ 1 \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} u_x \\ u_y \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} u_x & u_y & 1 \\ u_x & u_y & 1 \end{bmatrix} \mathbf{A} \begin{bmatrix} -u_x v_x & -u_y v_x & -v_x \\ -u_x v_y & -u_y v_y & -v_y \end{bmatrix} \mathbf{X} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

8x9

$$\mathbf{x} = \mathbf{V}_{:,end}^T = \text{null}(\mathbf{A})$$

Camera Calibration



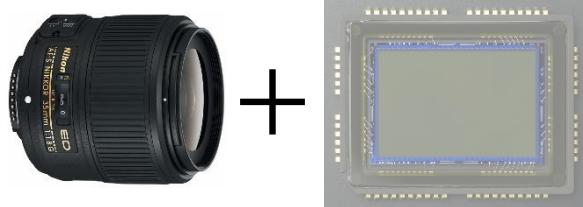
Camera Intrinsic Parameter



Pixel space

Metric space

$$\lambda \begin{bmatrix} u_{\text{img}} \\ v_{\text{img}} \\ 1 \end{bmatrix} = \begin{bmatrix} f_x & p_x \\ f_y & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$



Camera intrinsic parameter
: metric space to pixel space

Physical Focal Point



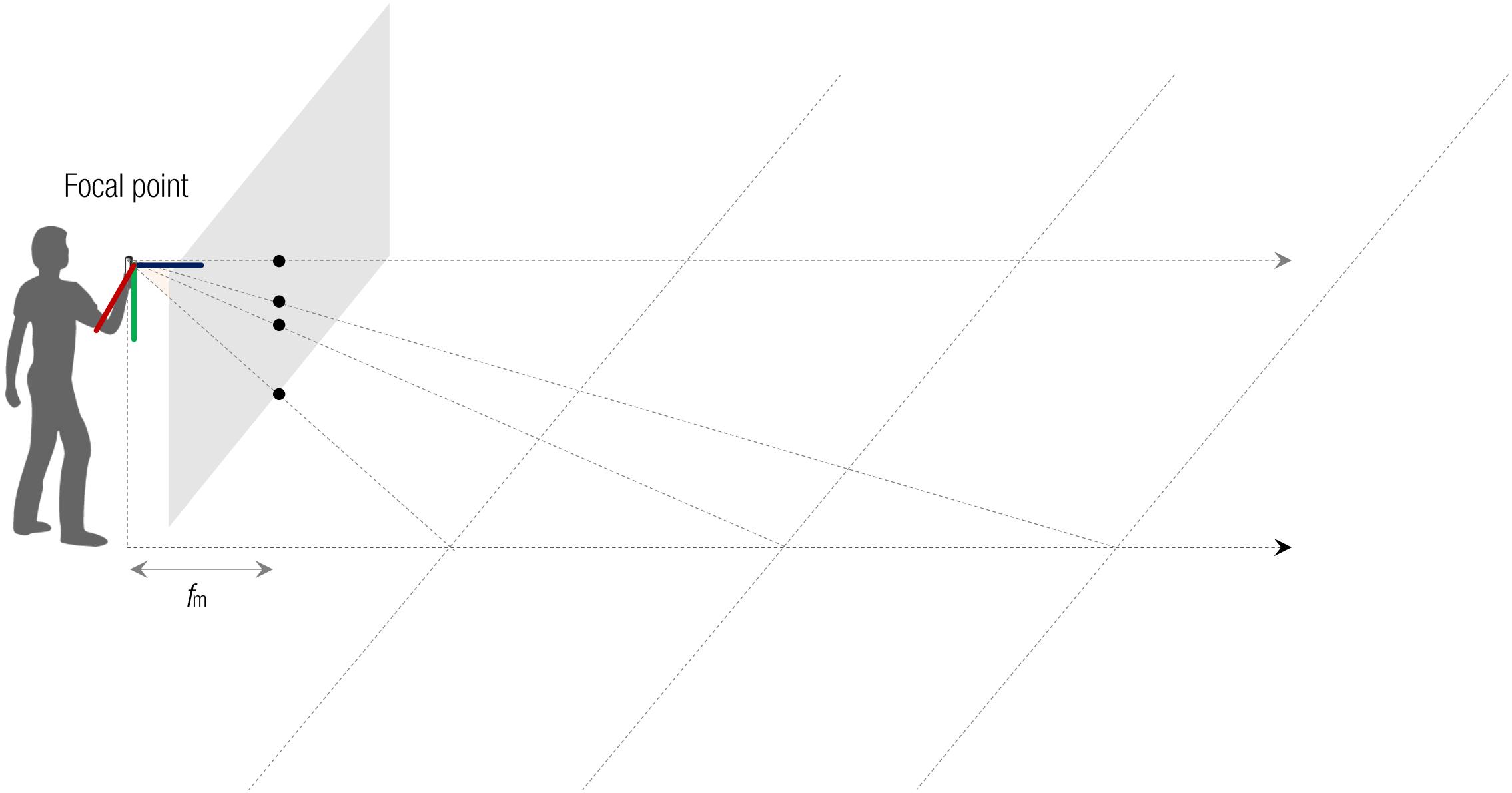
Physical Focal Point



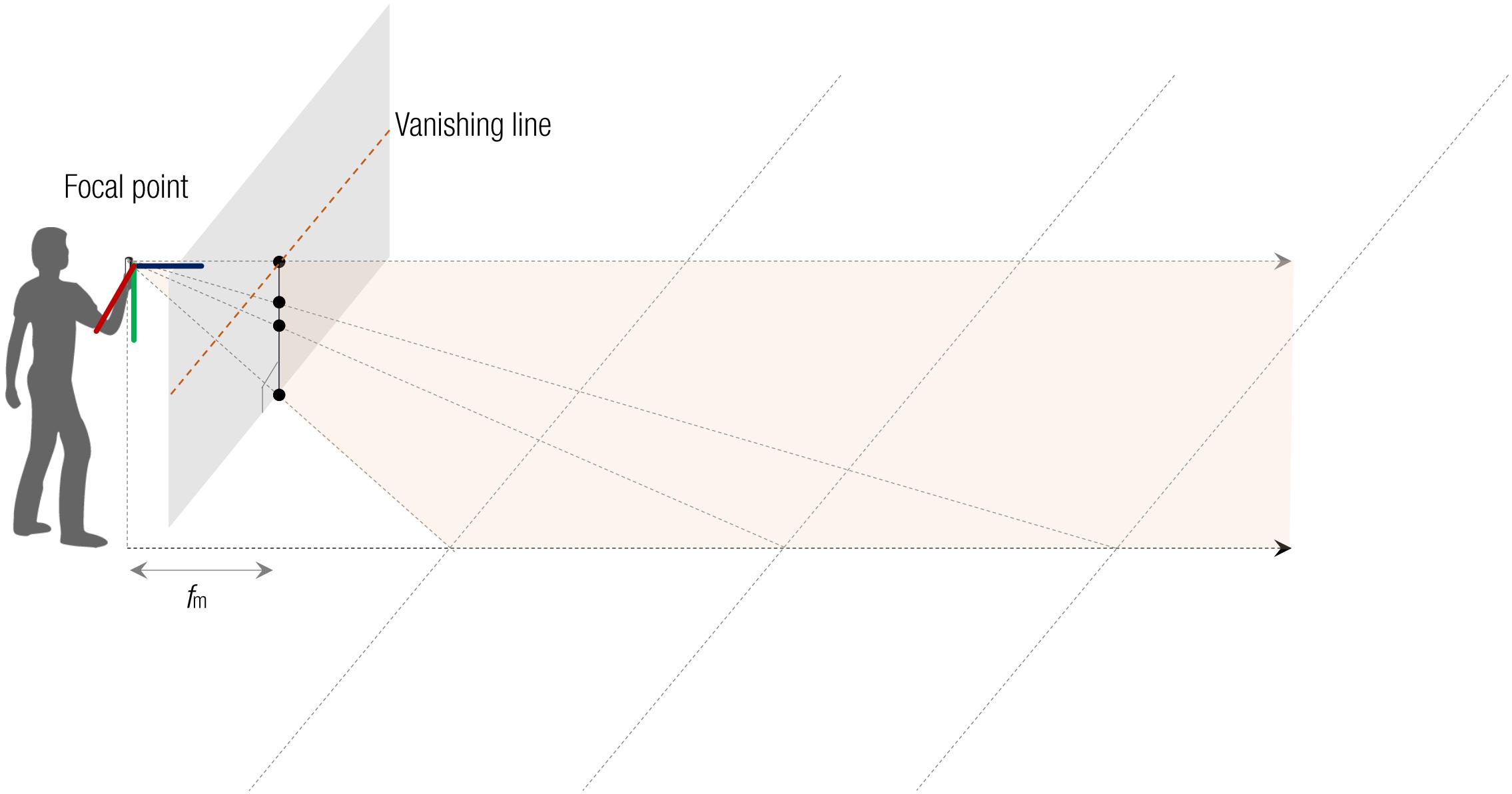
Physical Focal Point



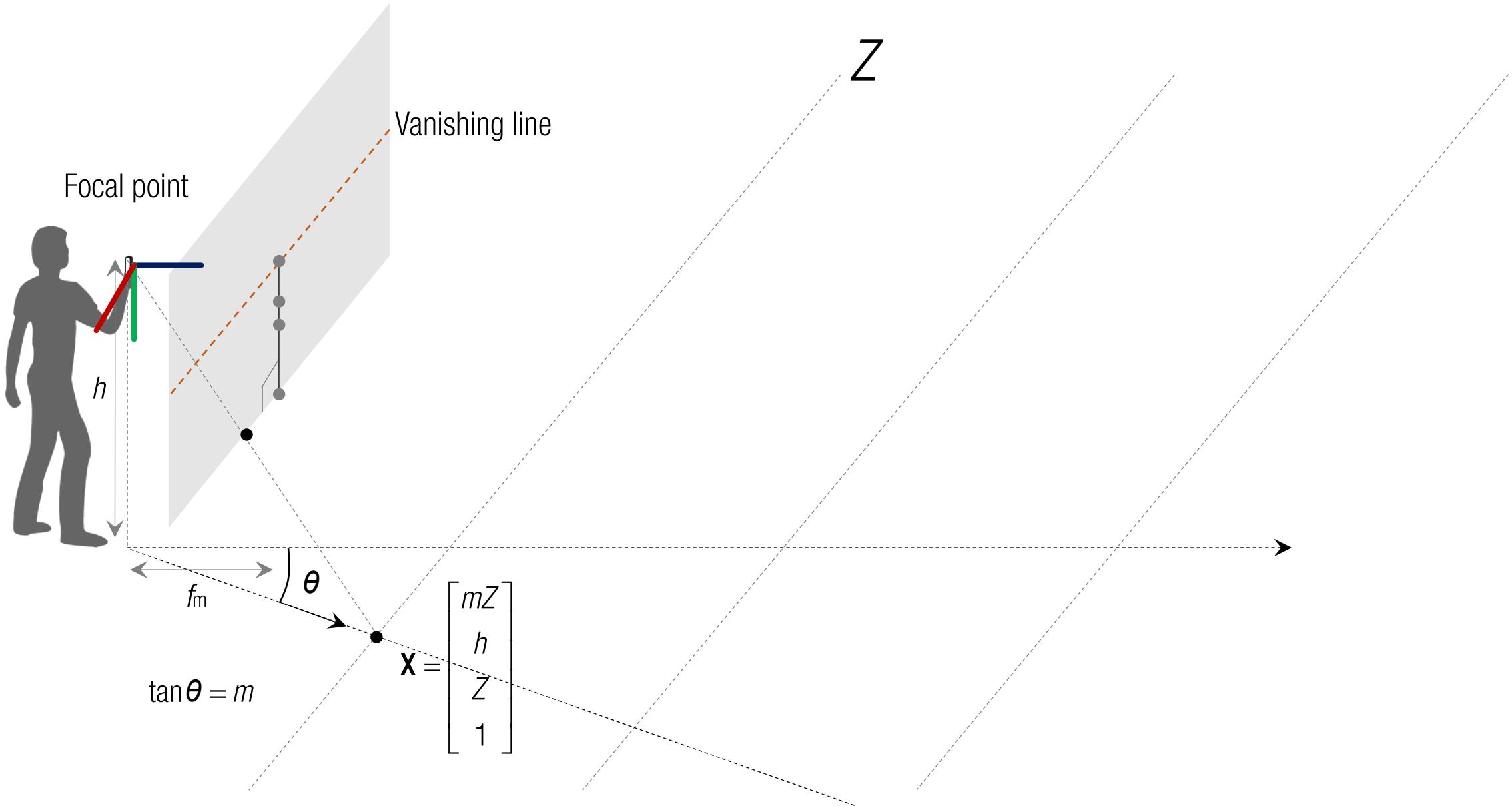
Physical Focal Point



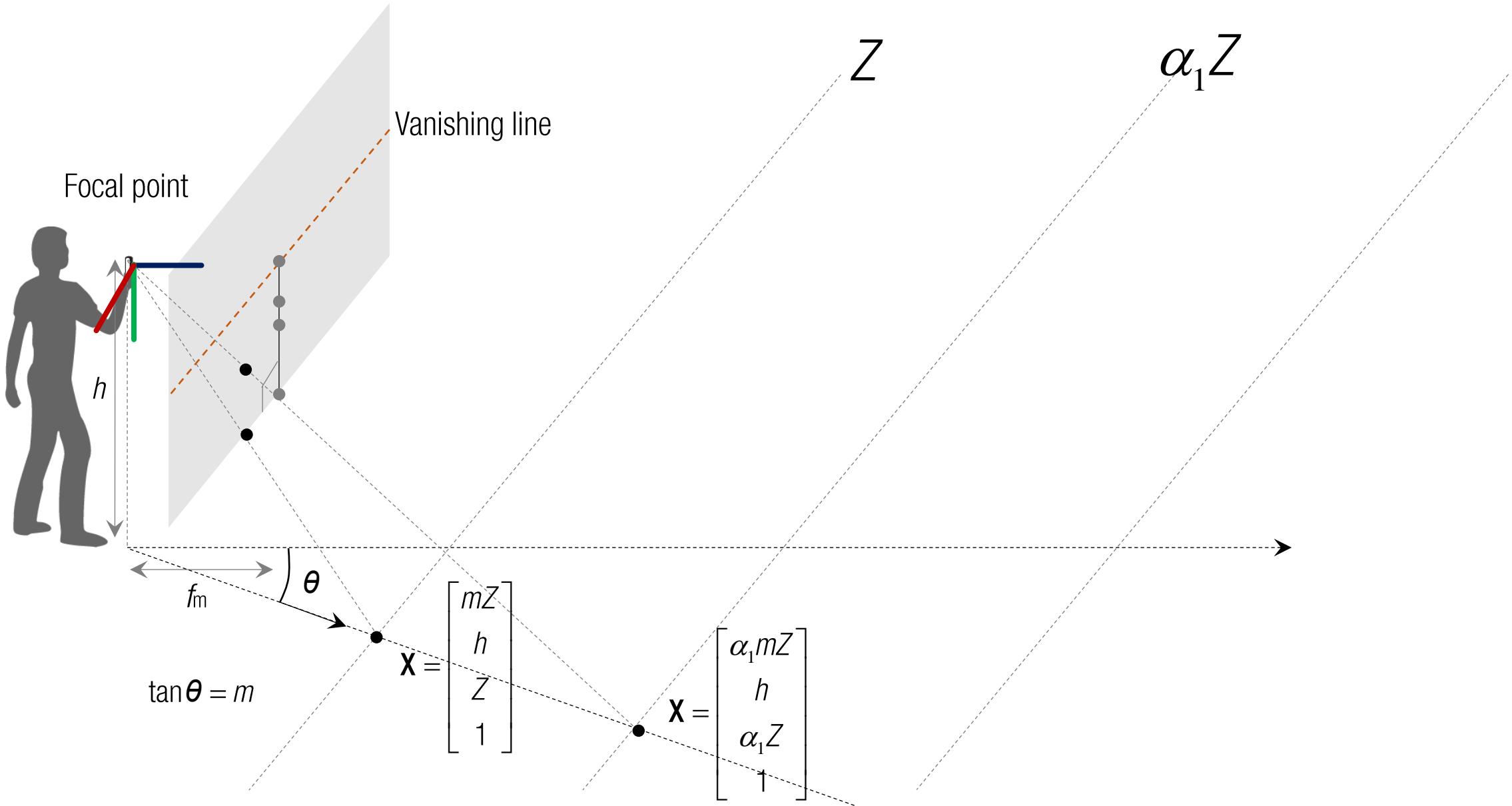
Physical Focal Point



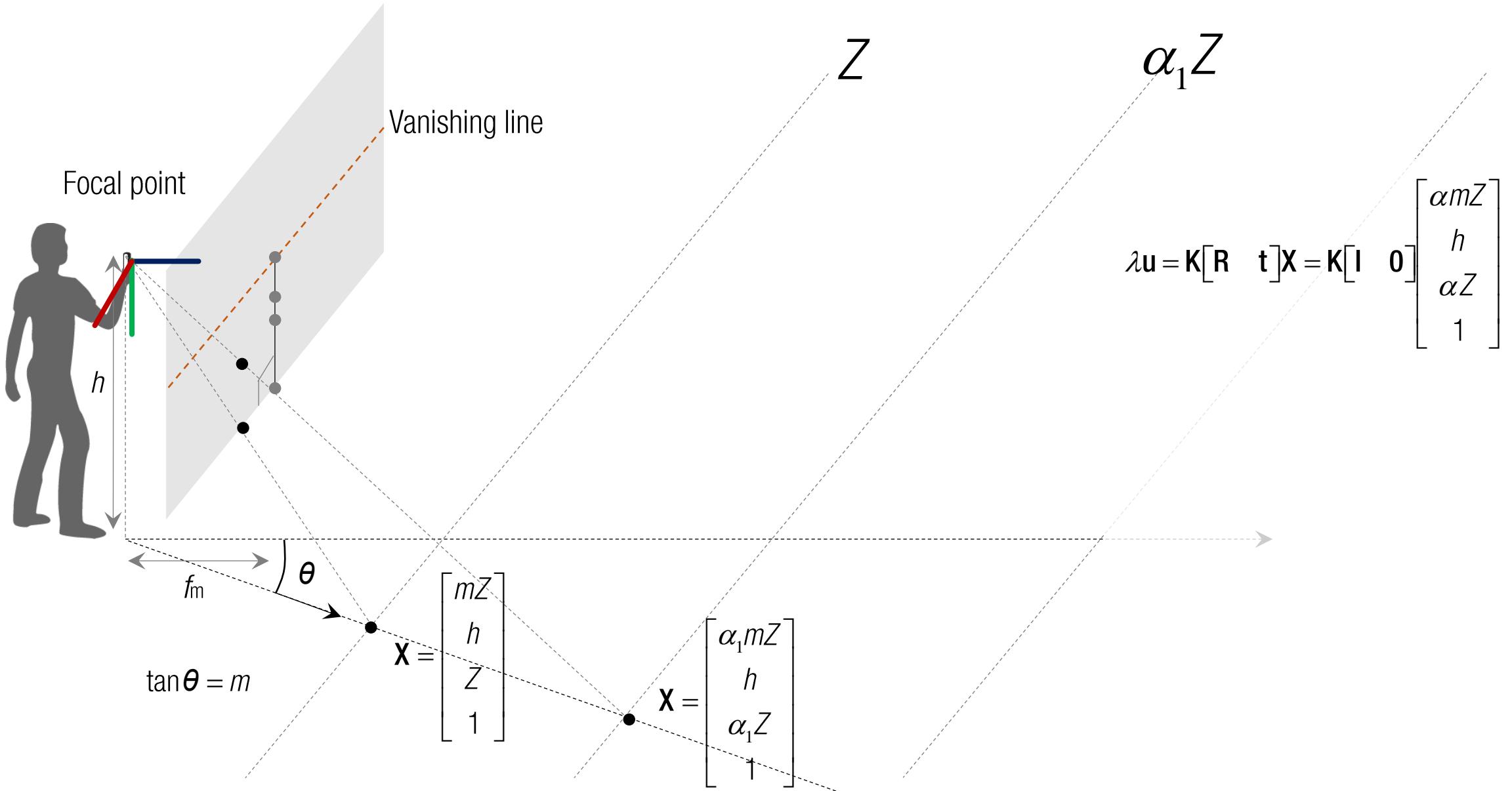
Physical Focal Point



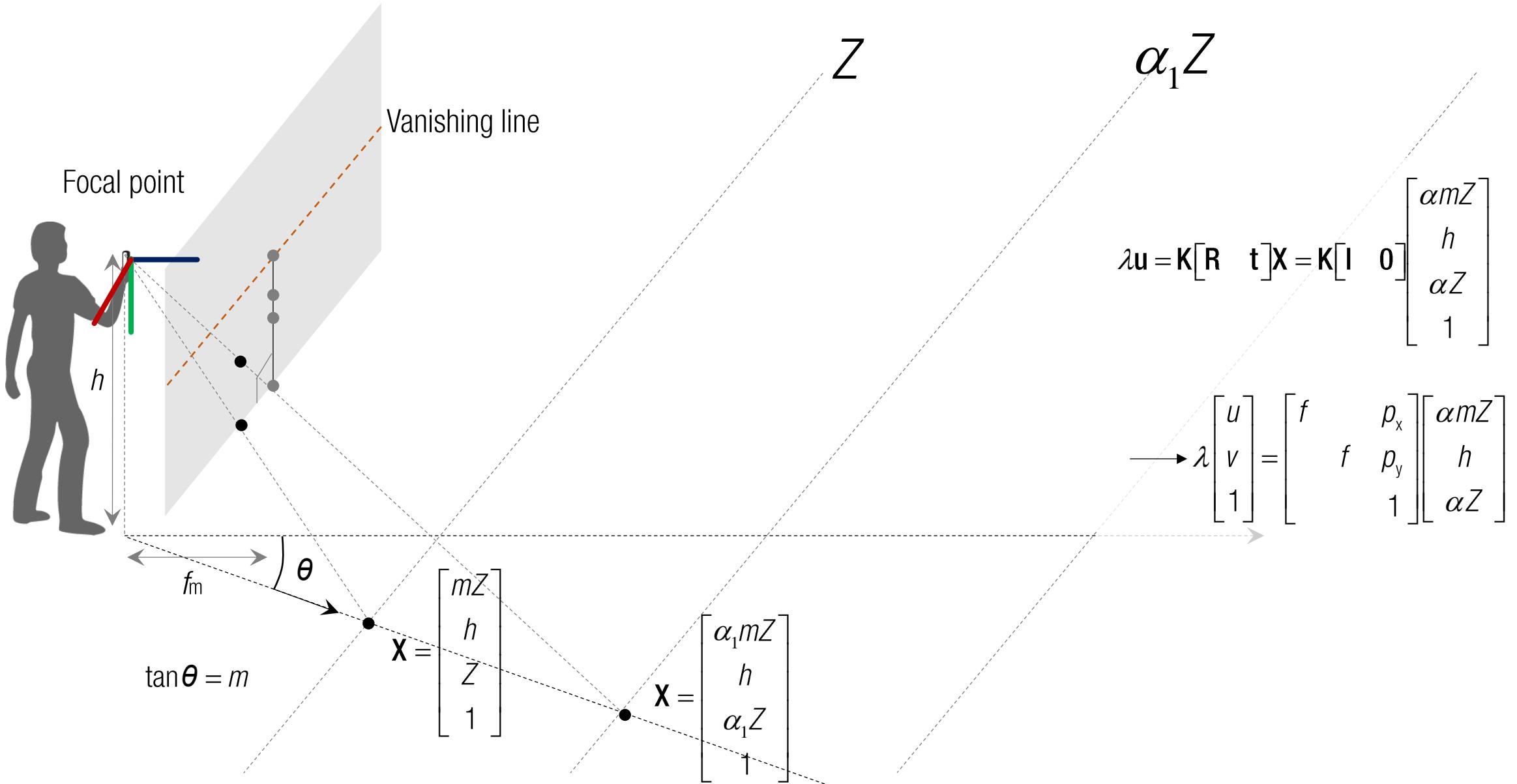
Physical Focal Point



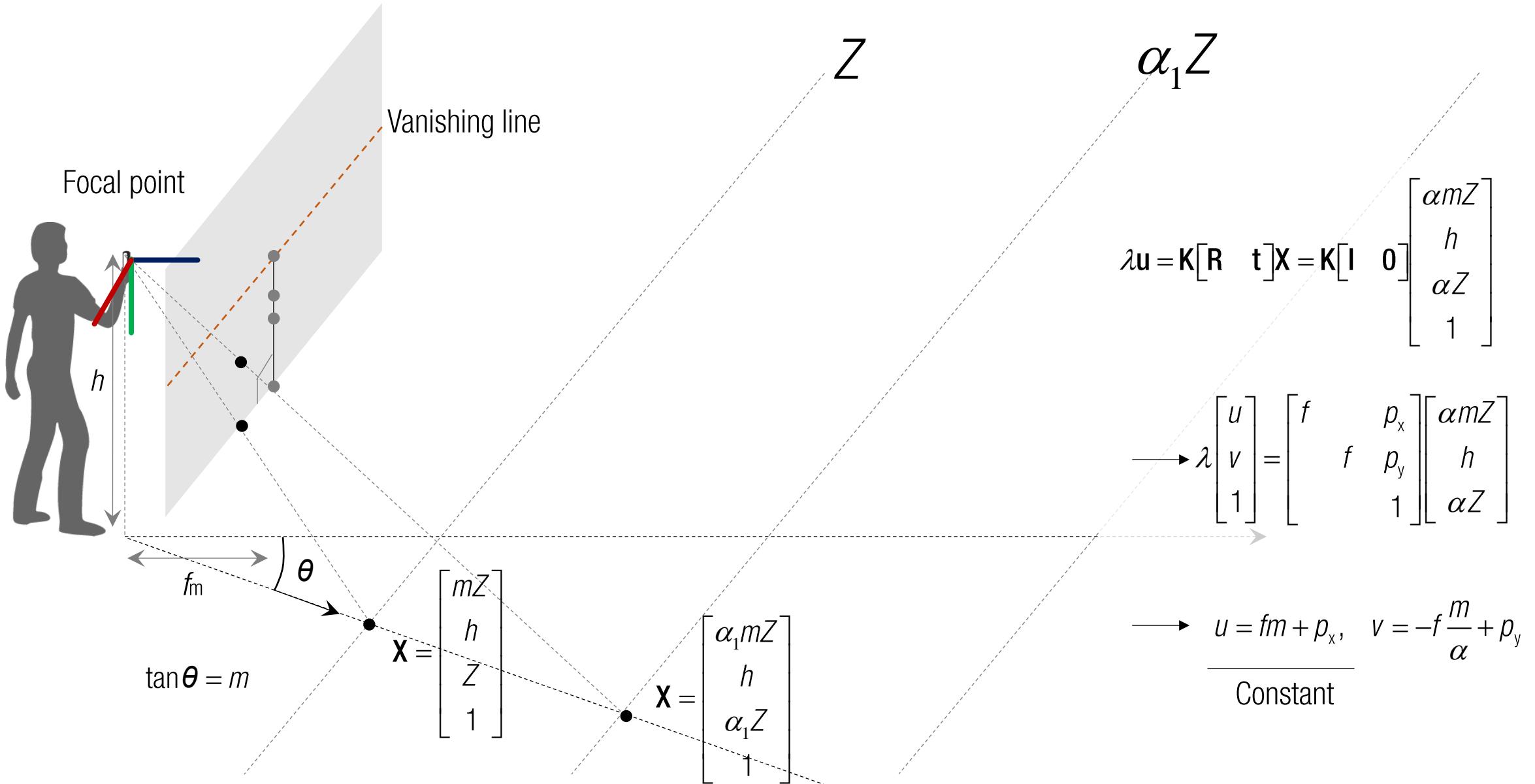
Physical Focal Point



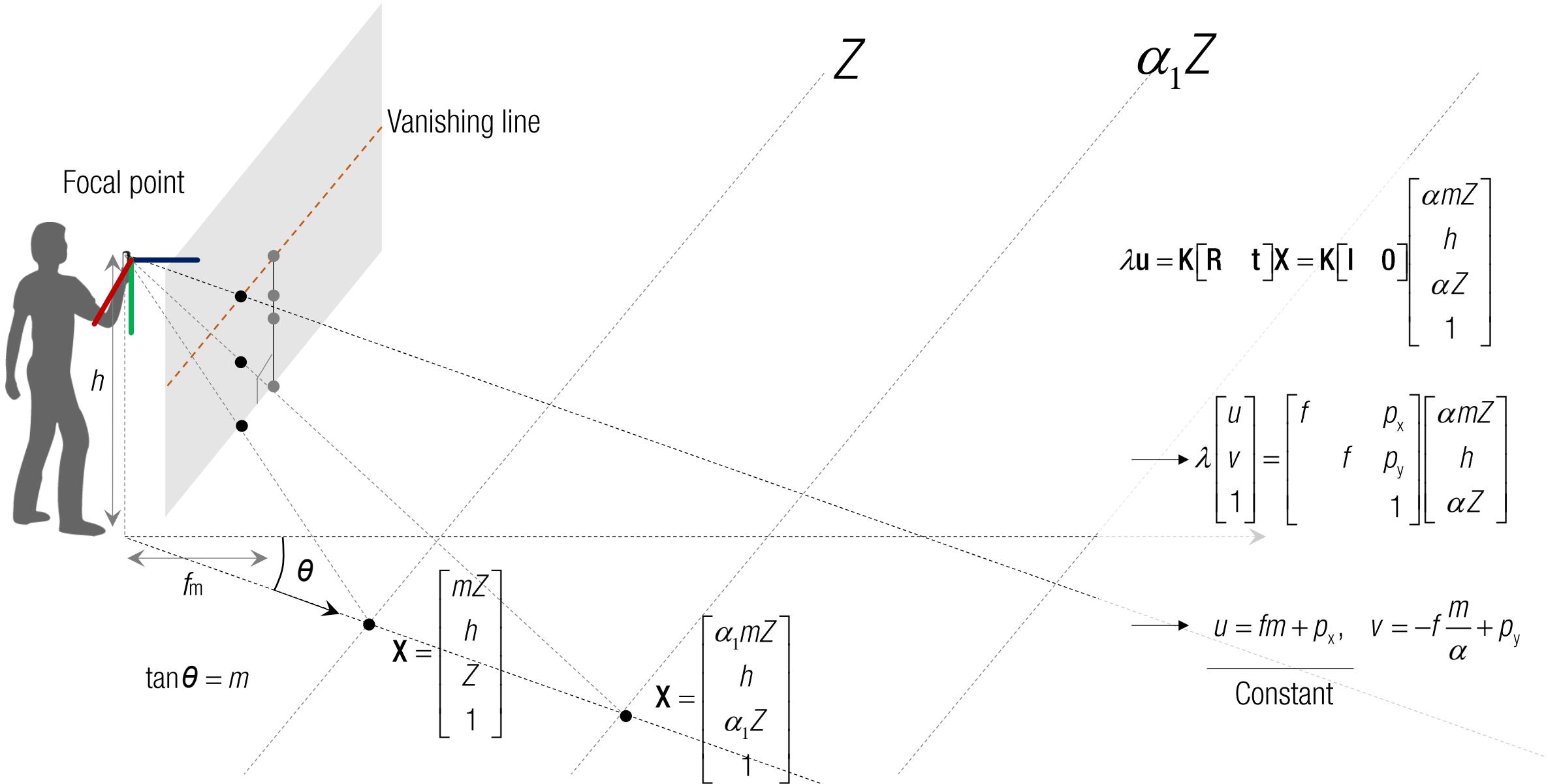
Physical Focal Point



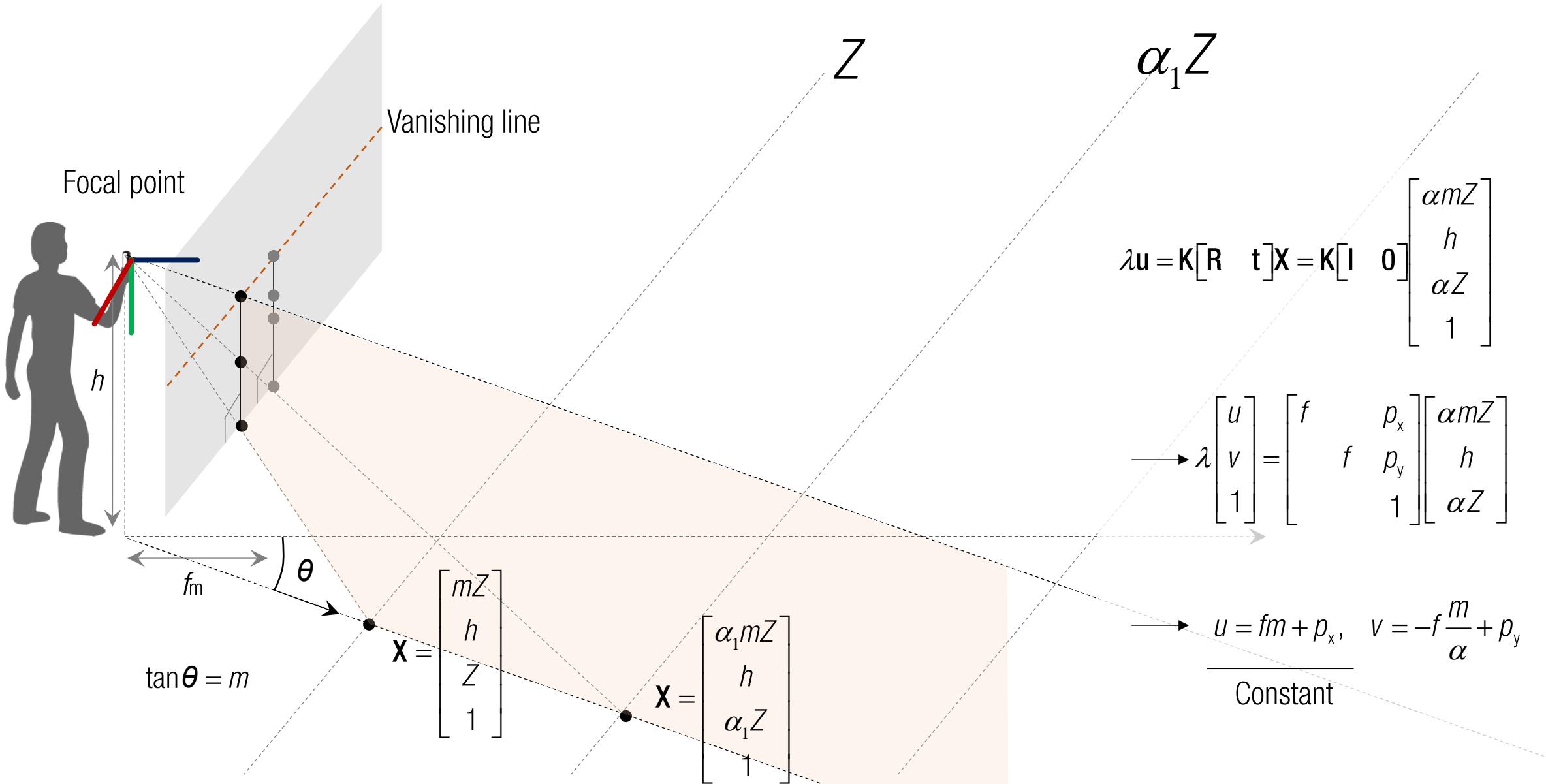
Physical Focal Point



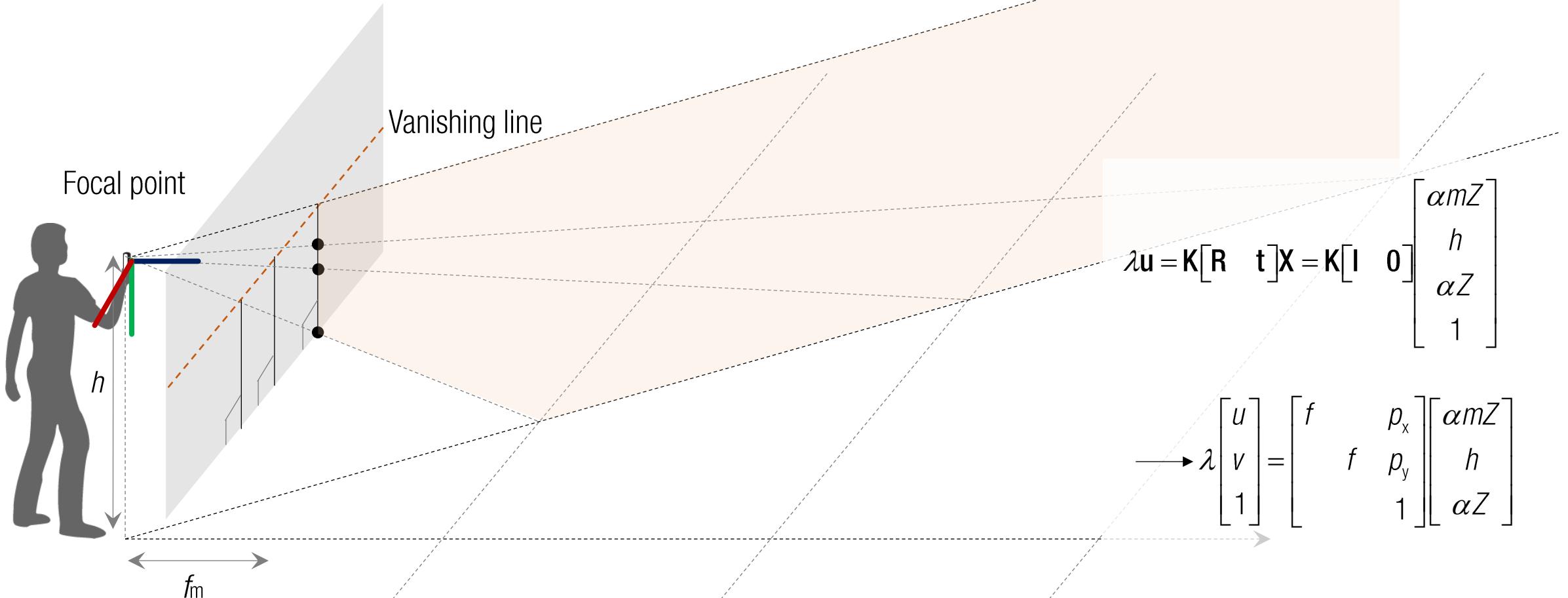
Physical Focal Point



Physical Focal Point



Physical Focal Point



$$\lambda u = K[R \ t]X = K[I \ 0]$$

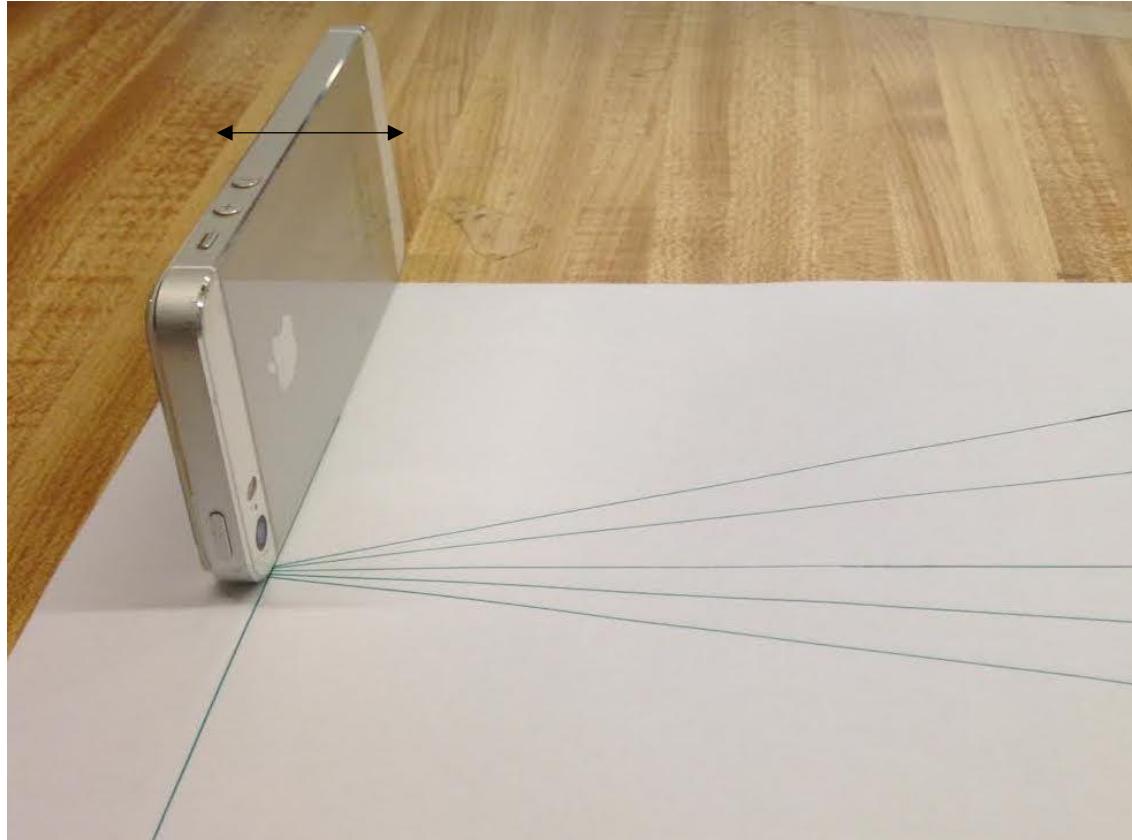
$$\begin{bmatrix} \alpha mZ \\ h \\ \alpha Z \\ 1 \end{bmatrix}$$

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \alpha mZ \\ h \\ \alpha Z \end{bmatrix}$$

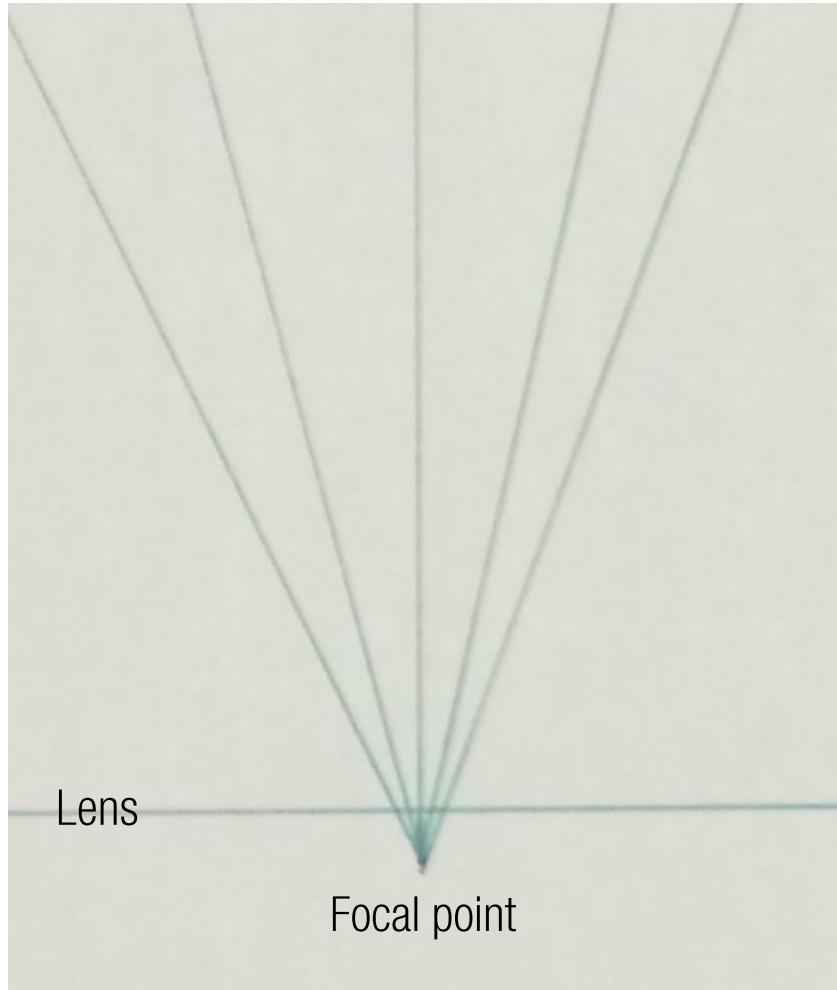
$$u = fm + p_x, \quad v = -f \frac{m}{\alpha} + p_y$$

Constant

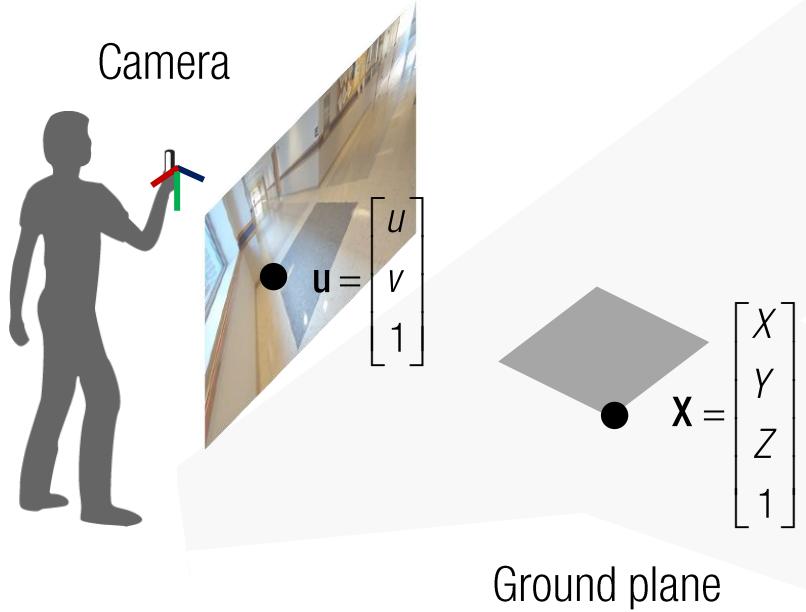
Where am I? (Focal Point)



Where am I? (Focal Point)



Camera Calibration in Pixel Space

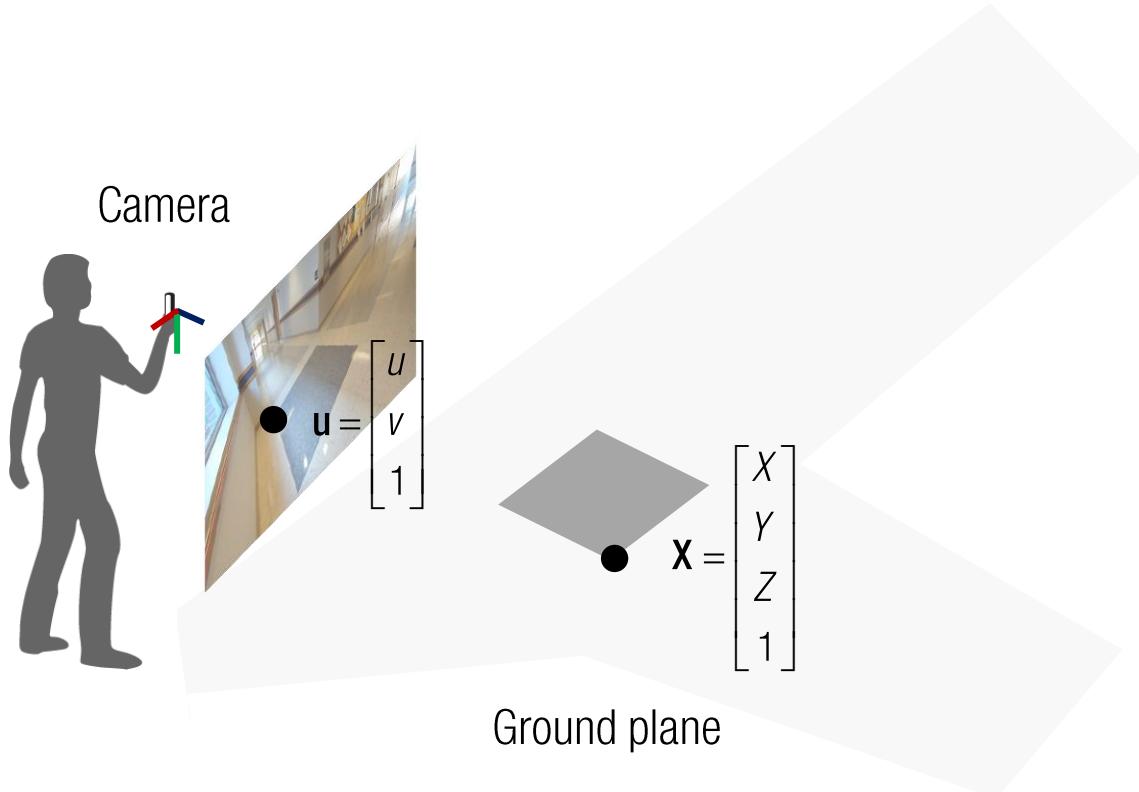


$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} R \\ t \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

of unknowns:

of equations:

Camera Calibration in Pixel Space

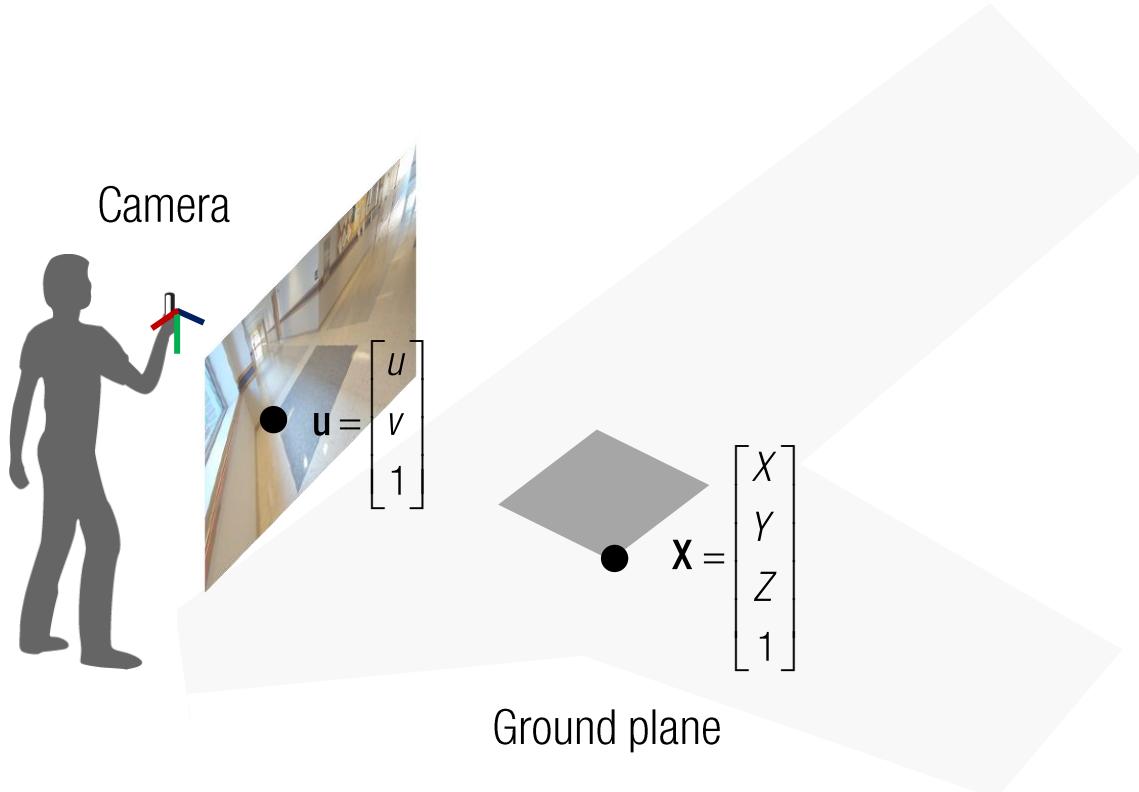


$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

of unknowns: 3 (\mathbf{K}) + 6 (\mathbf{R} and \mathbf{t}) + 3 (\mathbf{X})

of equations:

Camera Calibration in Pixel Space

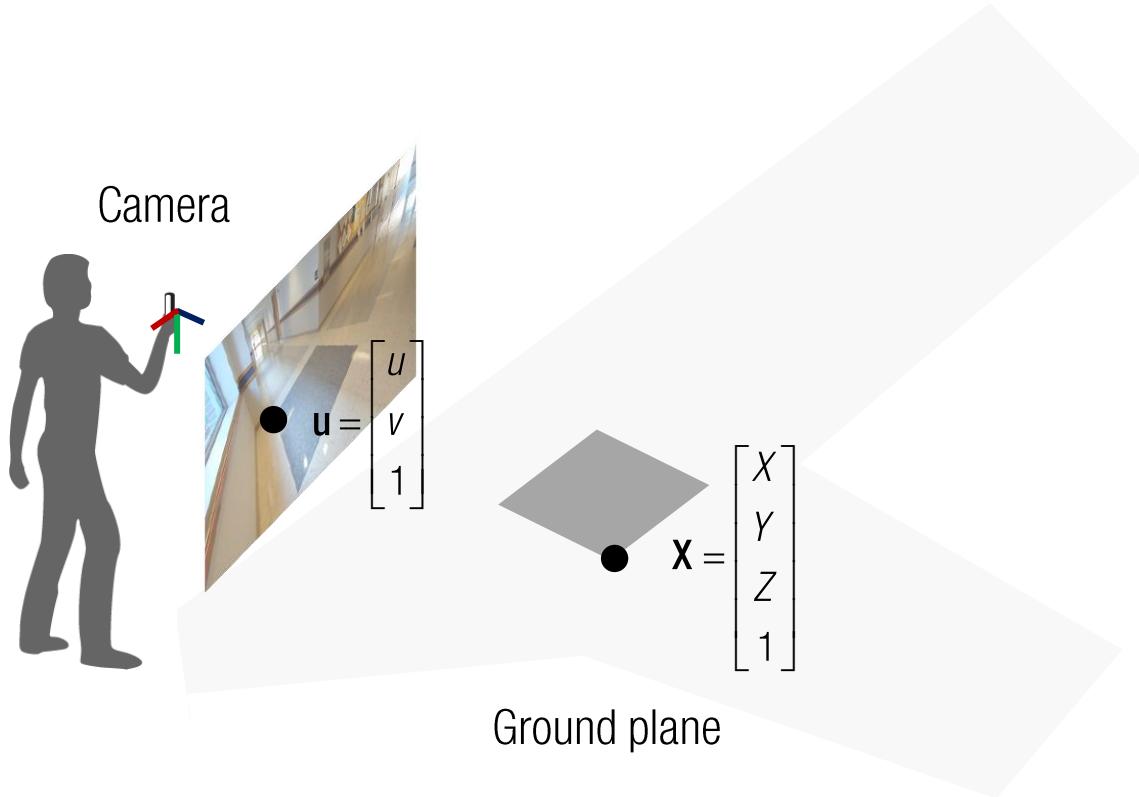


$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

of unknowns: 3 (\mathbf{K}) + 6 (\mathbf{R} and \mathbf{t}) + 3 (\mathbf{X})

of equations: 2

Camera Calibration in Pixel Space



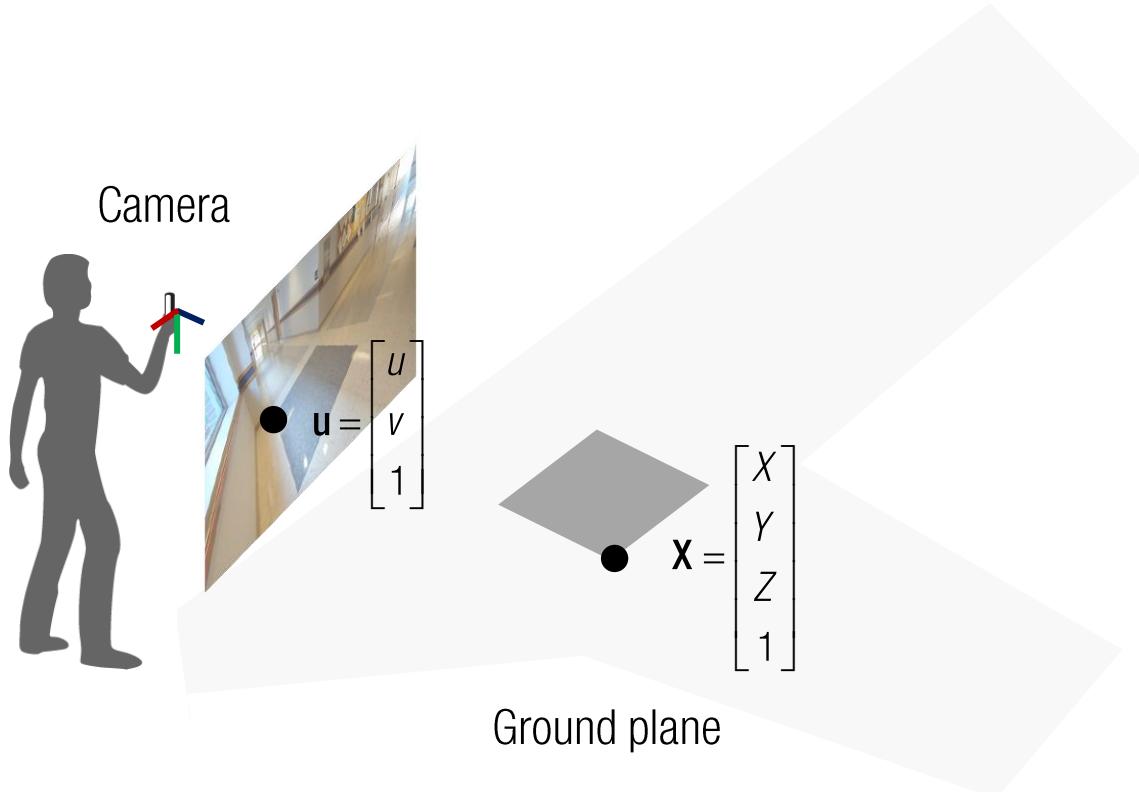
$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

of unknowns: $3(\mathbf{K}) + 6F(\mathbf{R} \text{ and } \mathbf{t}) + 3P(\mathbf{X})$

of equations: $2P$

where F is # of images and P is # of points.

Camera Calibration in Pixel Space



$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

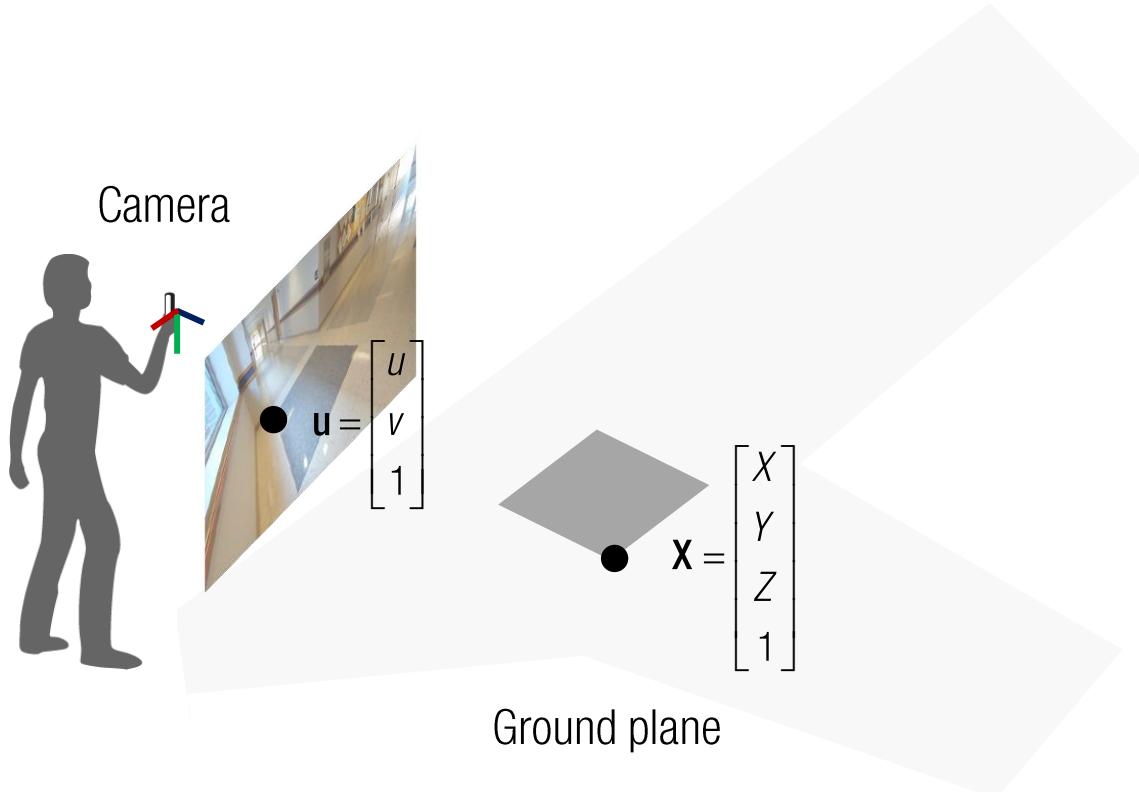
of unknowns: $3(\mathbf{K}) + 6F(\mathbf{R} \text{ and } \mathbf{t}) + 3P(\mathbf{X})$

of equations: $2P$

where F is # of images and P is # of points.

of unknowns $>$ # of equations

Camera Calibration in Pixel Space



$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{R} & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

of unknowns: $3(\mathbf{K}) + 6F(\mathbf{R} \text{ and } \mathbf{t}) + 3P(\mathbf{X})$

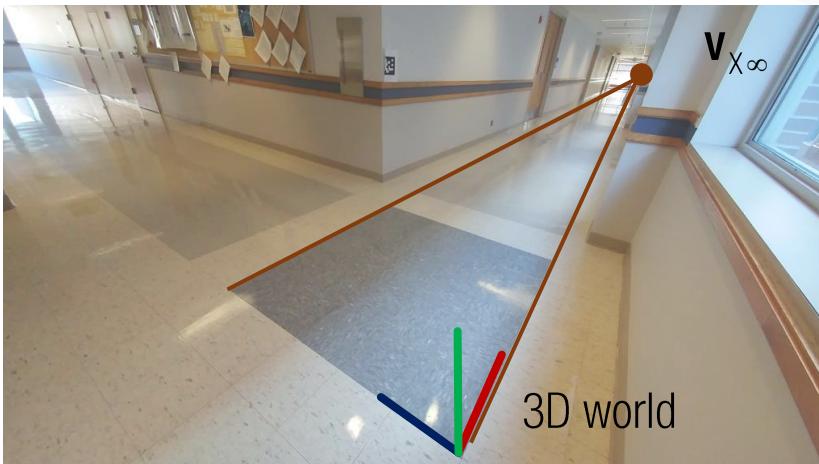
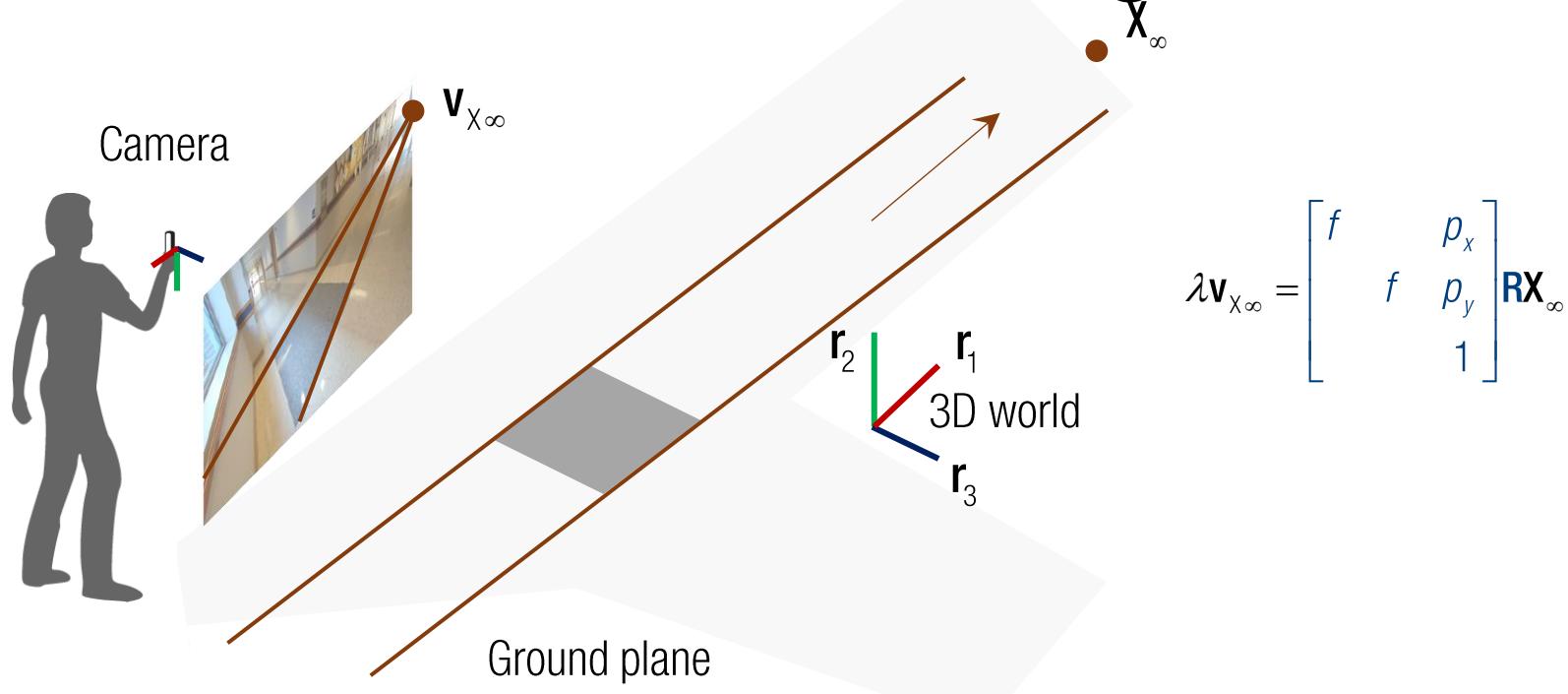
of equations: $2P$

where F is # of images and P is # of points.

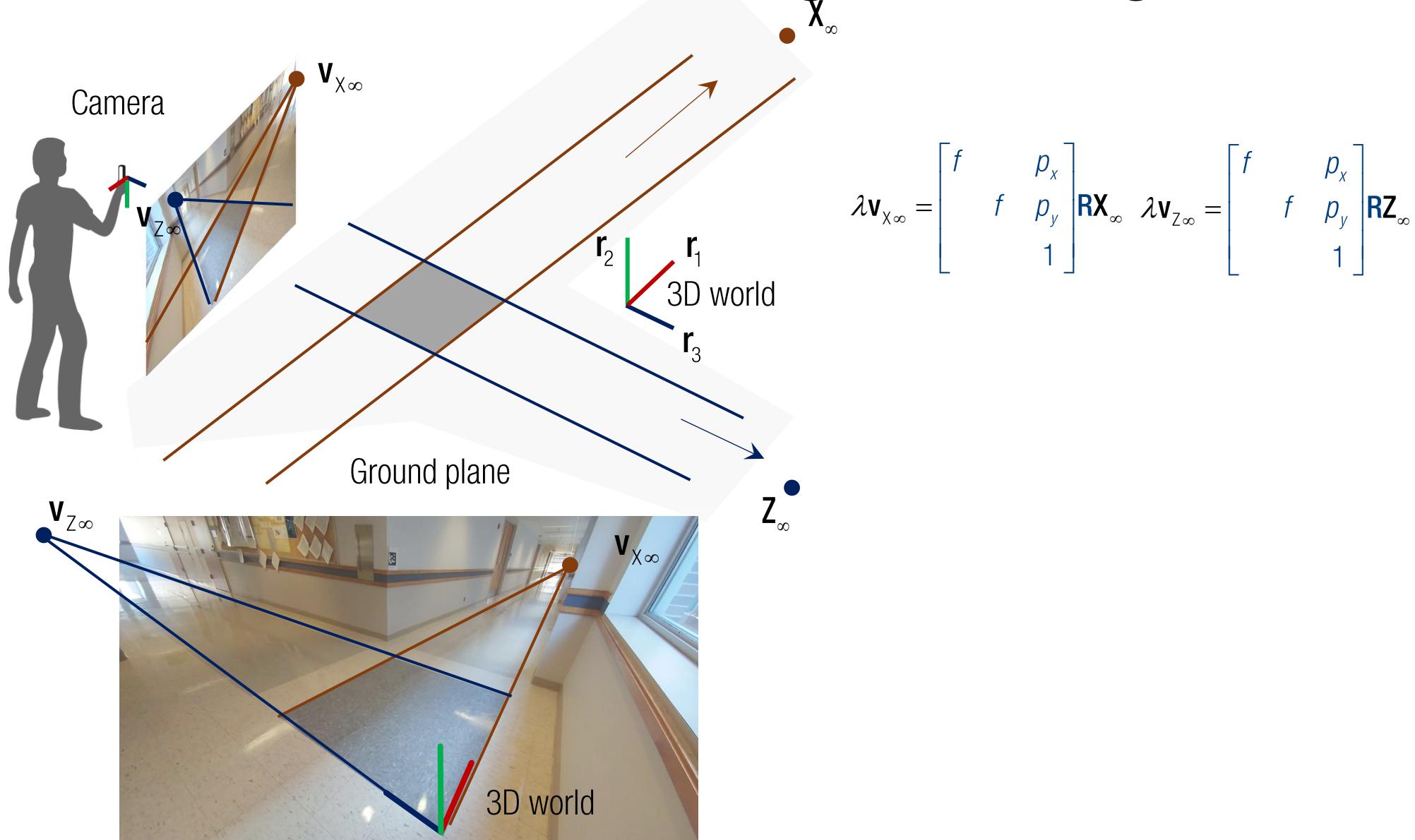
of unknowns $>$ # of equations

What do we know about the scene?

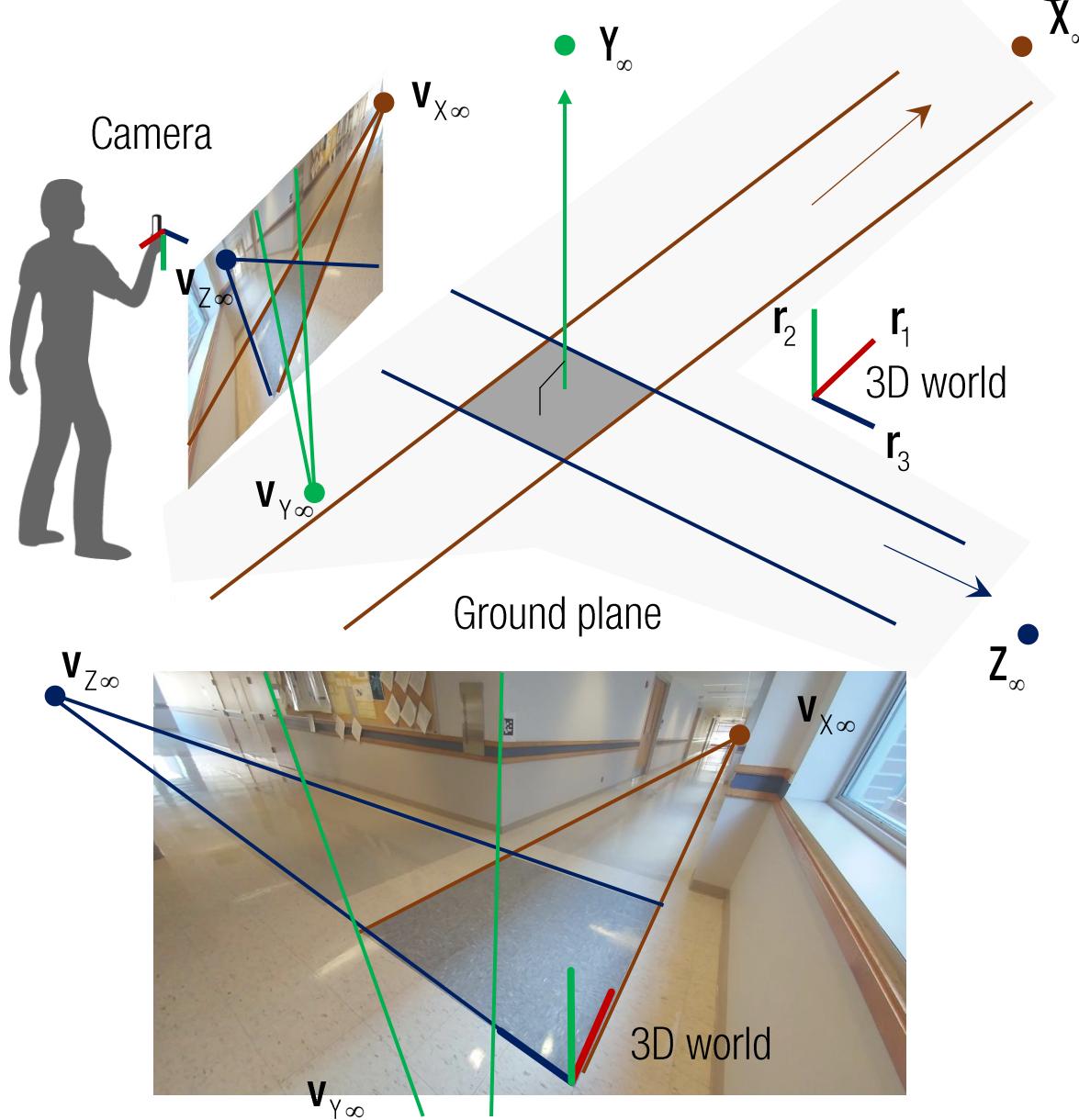
Camera Calibration using Vanishing Points



Camera Calibration using Vanishing Points



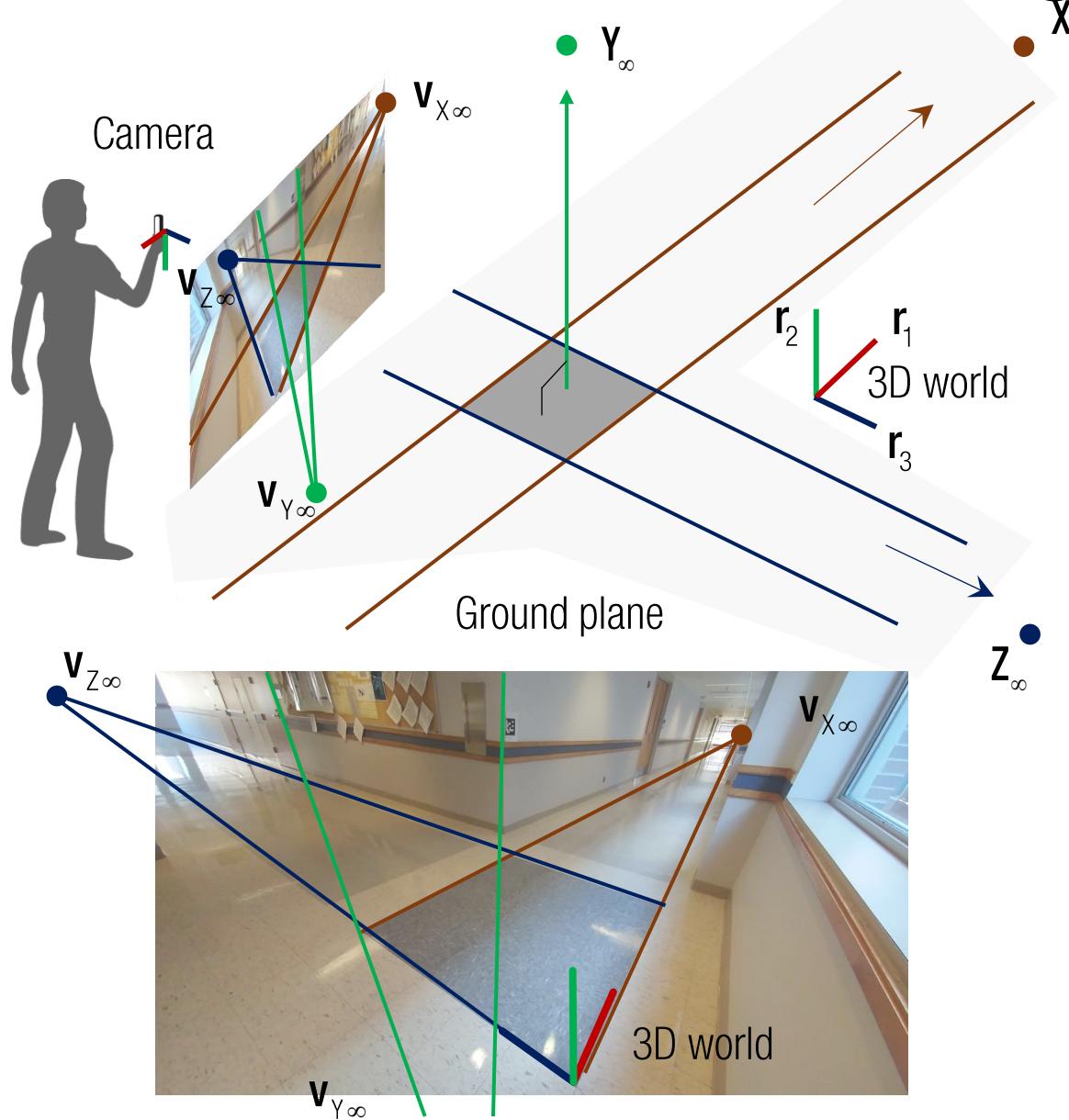
Camera Calibration using Vanishing Points



$$\lambda \mathbf{v}_{X\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \mathbf{R} \mathbf{X}_\infty \quad \lambda \mathbf{v}_{Z\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \mathbf{R} \mathbf{Z}_\infty \quad \lambda \mathbf{v}_{Y\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \mathbf{R} \mathbf{Y}_\infty$$

Note that the camera extrinsic is still unknown (\mathbf{R} and \mathbf{t}).

Camera Calibration using Vanishing Points



$$\lambda v_{X_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R X_\infty \quad \lambda v_{Z_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R Z_\infty \quad \lambda v_{Y_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R Y_\infty$$

Note that the camera extrinsic is still unknown (\mathbf{R} and \mathbf{t}).

Known property of points at infinity:

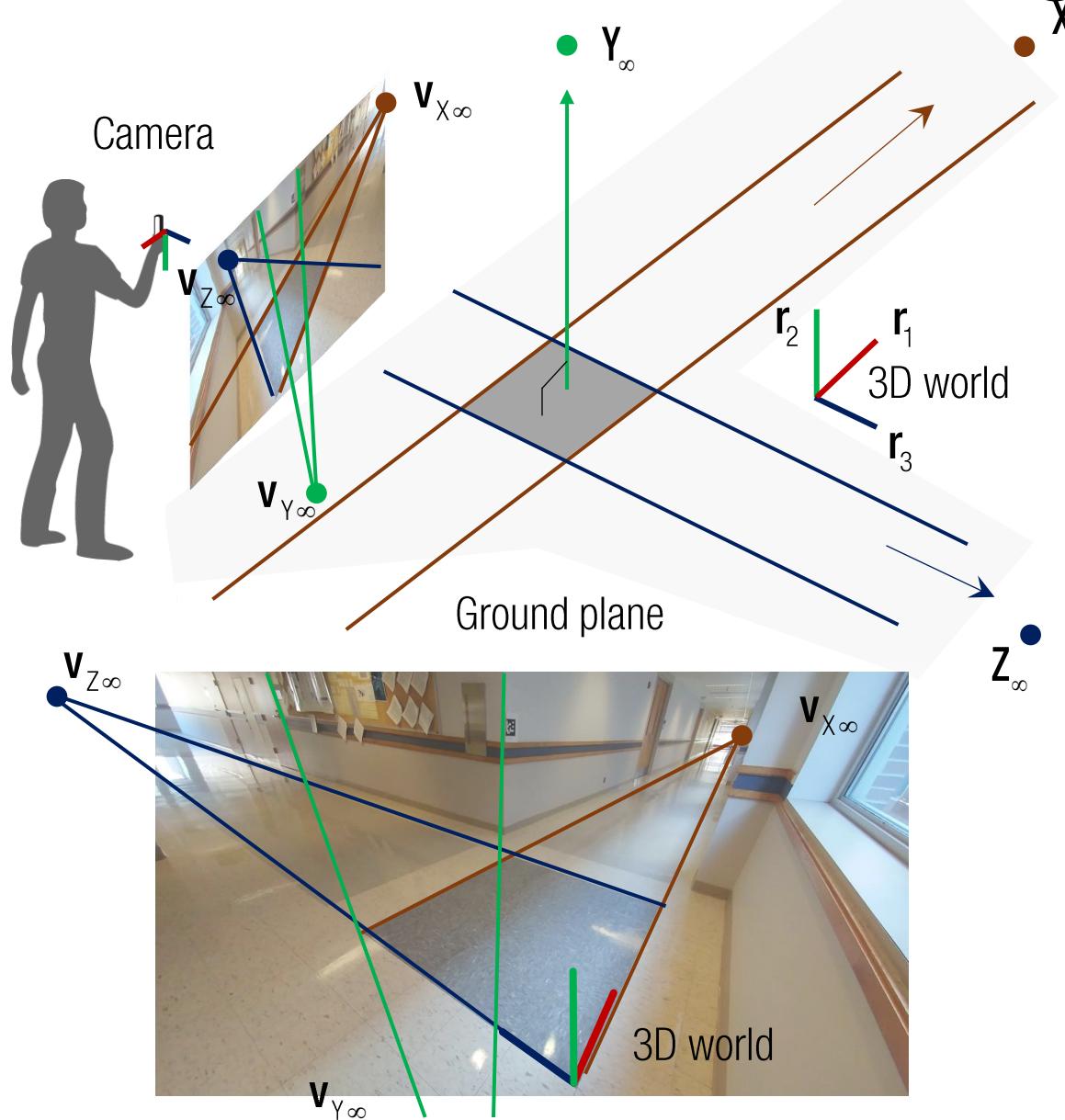
$$(X_\infty)^\top (Y_\infty) = 0$$

$$(Y_\infty)^\top (Z_\infty) = 0$$

$$(Z_\infty)^\top (X_\infty) = 0$$

These axes are perpendicular to each other.

Camera Calibration using Vanishing Points



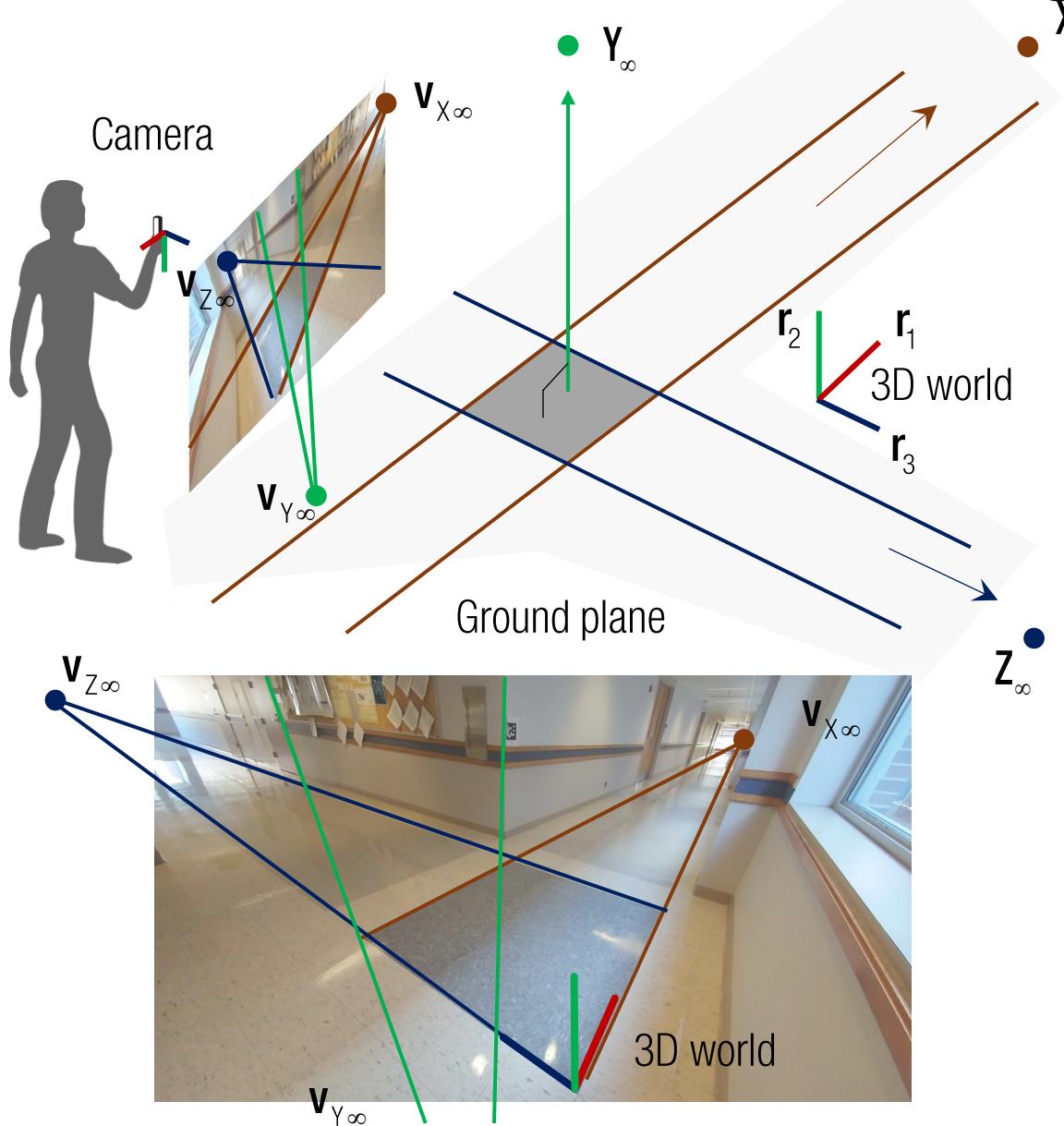
$$\lambda v_{X_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R X_\infty \quad \lambda v_{Z_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R Z_\infty \quad \lambda v_{Y_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R Y_\infty$$

Note that the camera extrinsic is still unknown (\mathbf{R} and \mathbf{t}).

Known property of points at infinity:

$$\begin{array}{ll}
 (X_\infty)^\top (Y_\infty) = 0 & (RX_\infty)^\top (RY_\infty) = 0 \\
 (Y_\infty)^\top (Z_\infty) = 0 & \longleftrightarrow (RY_\infty)^\top (RZ_\infty) = 0 \\
 (Z_\infty)^\top (X_\infty) = 0 & (RZ_\infty)^\top (RX_\infty) = 0
 \end{array}$$

Camera Calibration using Vanishing Points



$$\lambda v_{X_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R X_\infty \quad \lambda v_{Z_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R Z_\infty \quad \lambda v_{Y_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R Y_\infty$$

$$\lambda K^{-1} v_{X_\infty} = R X_\infty \quad \lambda K^{-1} v_{Y_\infty} = R Y_\infty \quad \lambda K^{-1} v_{Z_\infty} = R Z_\infty$$

Note that the camera extrinsic is still unknown (R and t).

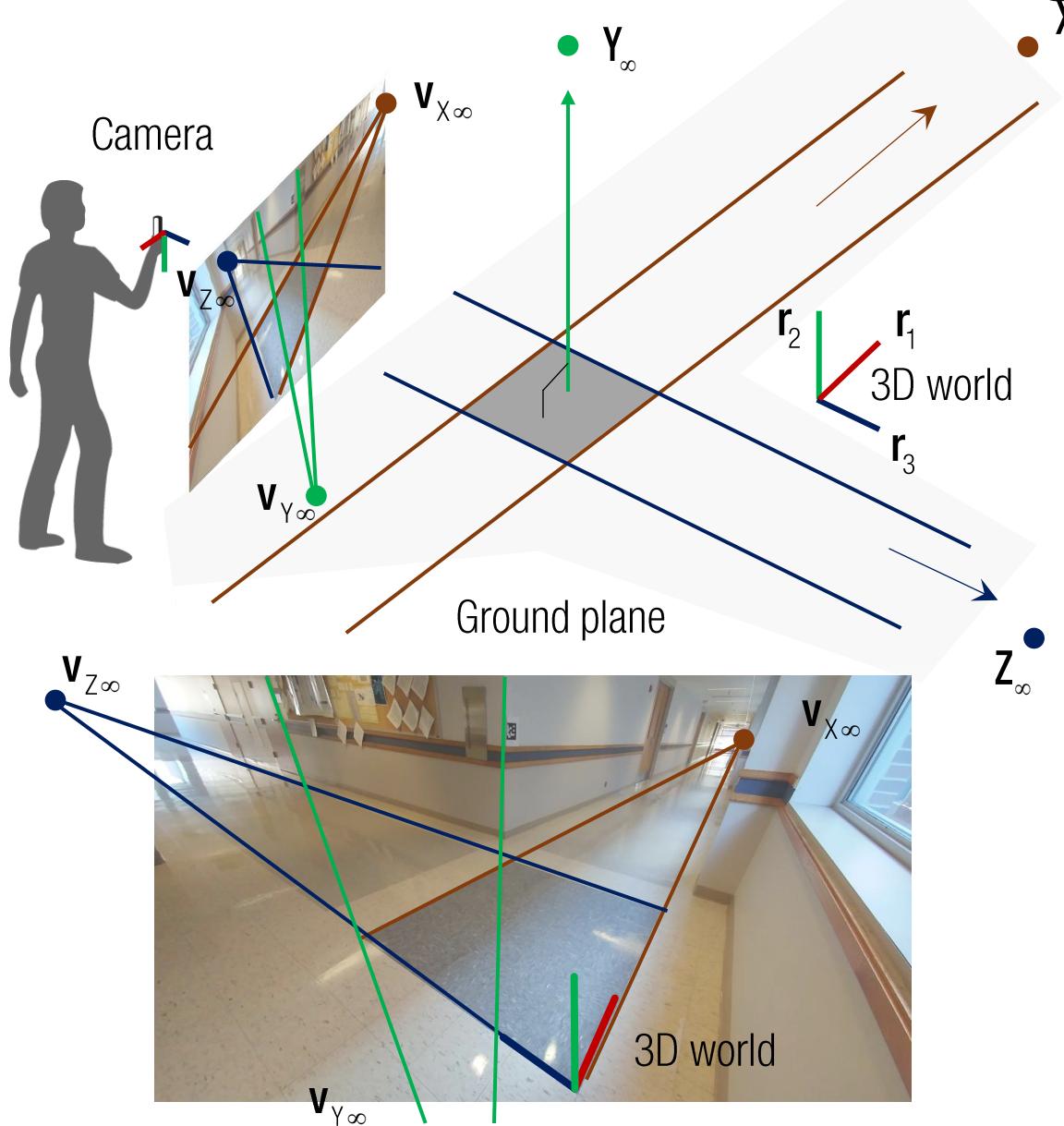
Known property of points at infinity:

$$(X_\infty)^\top (Y_\infty) = 0 \qquad (R X_\infty)^\top (R Y_\infty) = 0$$

$$(Y_\infty)^\top (Z_\infty) = 0 \quad \longleftrightarrow \quad (R Y_\infty)^\top (R Z_\infty) = 0$$

$$(Z_\infty)^\top (X_\infty) = 0 \qquad (R Z_\infty)^\top (R X_\infty) = 0$$

Camera Calibration using Vanishing Points



$$\lambda \mathbf{v}_{X_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \mathbf{R} \mathbf{X}_\infty \quad \lambda \mathbf{v}_{Z_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \mathbf{R} \mathbf{Z}_\infty \quad \lambda \mathbf{v}_{Y_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \mathbf{R} \mathbf{Y}_\infty$$

$$\lambda \mathbf{K}^{-1} \mathbf{v}_{X_\infty} = \mathbf{R} \mathbf{X}_\infty \quad \lambda \mathbf{K}^{-1} \mathbf{v}_{Y_\infty} = \mathbf{R} \mathbf{Y}_\infty \quad \lambda \mathbf{K}^{-1} \mathbf{v}_{Z_\infty} = \mathbf{R} \mathbf{Z}_\infty$$

Note that the camera extrinsic is still unknown (\mathbf{R} and \mathbf{t}).

Known property of points at infinity:

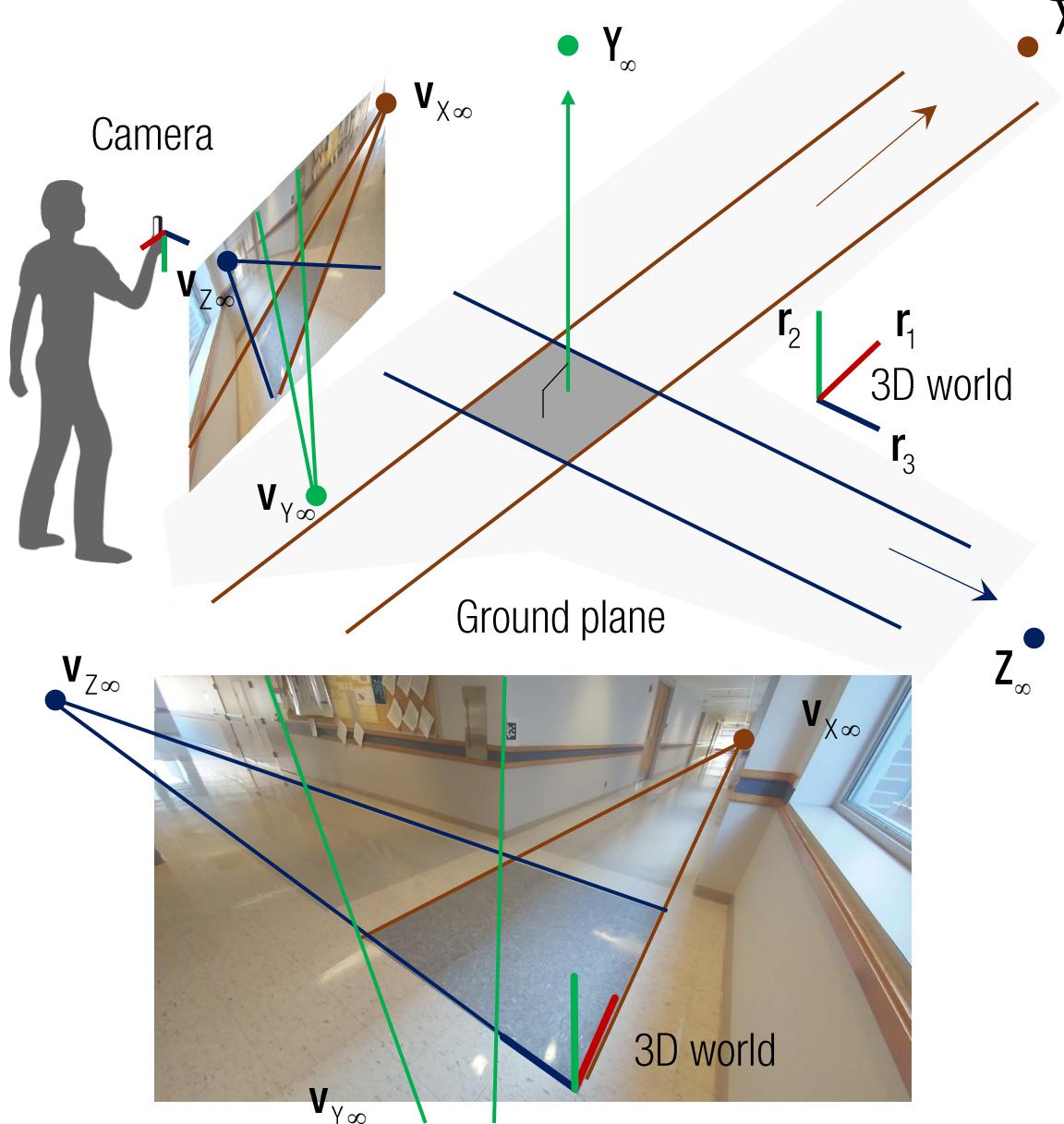
$$(\mathbf{X}_\infty)^\top (\mathbf{Y}_\infty) = 0 \qquad (\mathbf{R} \mathbf{X}_\infty)^\top (\mathbf{R} \mathbf{Y}_\infty) = 0$$

$$(\mathbf{Y}_\infty)^\top (\mathbf{Z}_\infty) = 0 \quad \longleftrightarrow \quad (\mathbf{R} \mathbf{Y}_\infty)^\top (\mathbf{R} \mathbf{Z}_\infty) = 0$$

$$(\mathbf{Z}_\infty)^\top (\mathbf{X}_\infty) = 0 \qquad (\mathbf{R} \mathbf{Z}_\infty)^\top (\mathbf{R} \mathbf{X}_\infty) = 0$$

$$(\mathbf{K}^{-1} \mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1} \mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1} \mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1} \mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1} \mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1} \mathbf{v}_{X_\infty}) = 0$$

Camera Calibration using Vanishing Points



$$\lambda \mathbf{v}_{X_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \mathbf{R} \mathbf{X}_\infty \quad \lambda \mathbf{v}_{Z_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \mathbf{R} \mathbf{Z}_\infty \quad \lambda \mathbf{v}_{Y_\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \mathbf{R} \mathbf{Y}_\infty$$

$$\lambda \mathbf{K}^{-1} \mathbf{v}_{X_\infty} = \mathbf{R} \mathbf{X}_\infty \quad \lambda \mathbf{K}^{-1} \mathbf{v}_{Y_\infty} = \mathbf{R} \mathbf{Y}_\infty \quad \lambda \mathbf{K}^{-1} \mathbf{v}_{Z_\infty} = \mathbf{R} \mathbf{Z}_\infty$$

Note that the camera extrinsic is still unknown (\mathbf{R} and \mathbf{t}).

Known property of points at infinity:

$$(\mathbf{X}_\infty)^\top (\mathbf{Y}_\infty) = 0 \qquad (\mathbf{R} \mathbf{X}_\infty)^\top (\mathbf{R} \mathbf{Y}_\infty) = 0$$

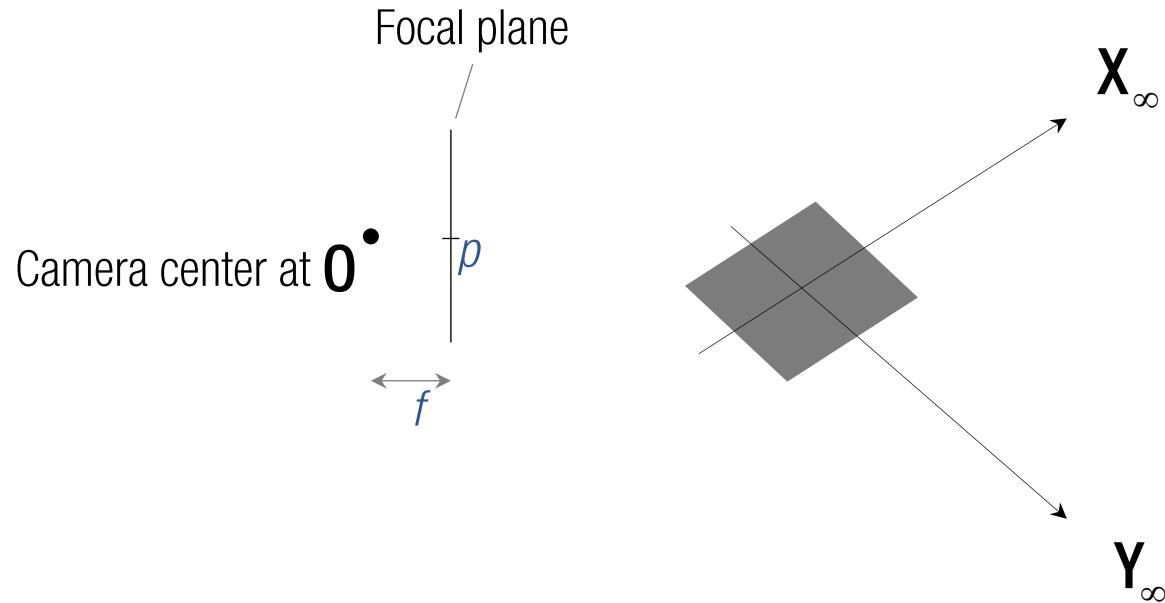
$$(\mathbf{Y}_\infty)^\top (\mathbf{Z}_\infty) = 0 \quad \longleftrightarrow \quad (\mathbf{R} \mathbf{Y}_\infty)^\top (\mathbf{R} \mathbf{Z}_\infty) = 0$$

$$(\mathbf{Z}_\infty)^\top (\mathbf{X}_\infty) = 0 \qquad (\mathbf{R} \mathbf{Z}_\infty)^\top (\mathbf{R} \mathbf{X}_\infty) = 0$$

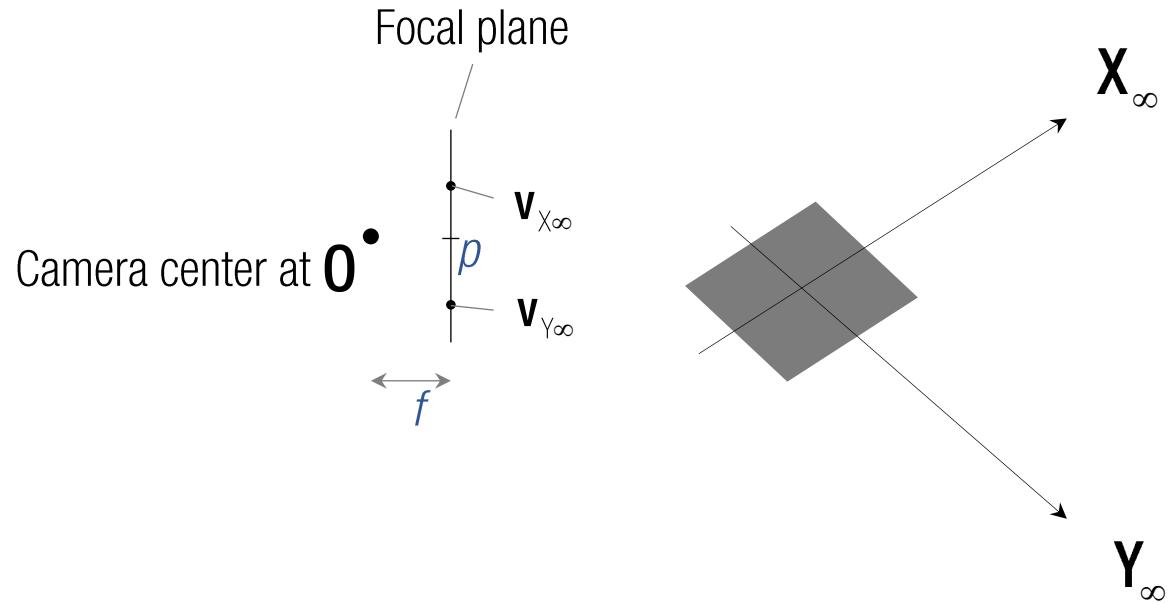
$$(\mathbf{K}^{-1} \mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1} \mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1} \mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1} \mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1} \mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1} \mathbf{v}_{X_\infty}) = 0$$

: 3 unknowns and 3 equations

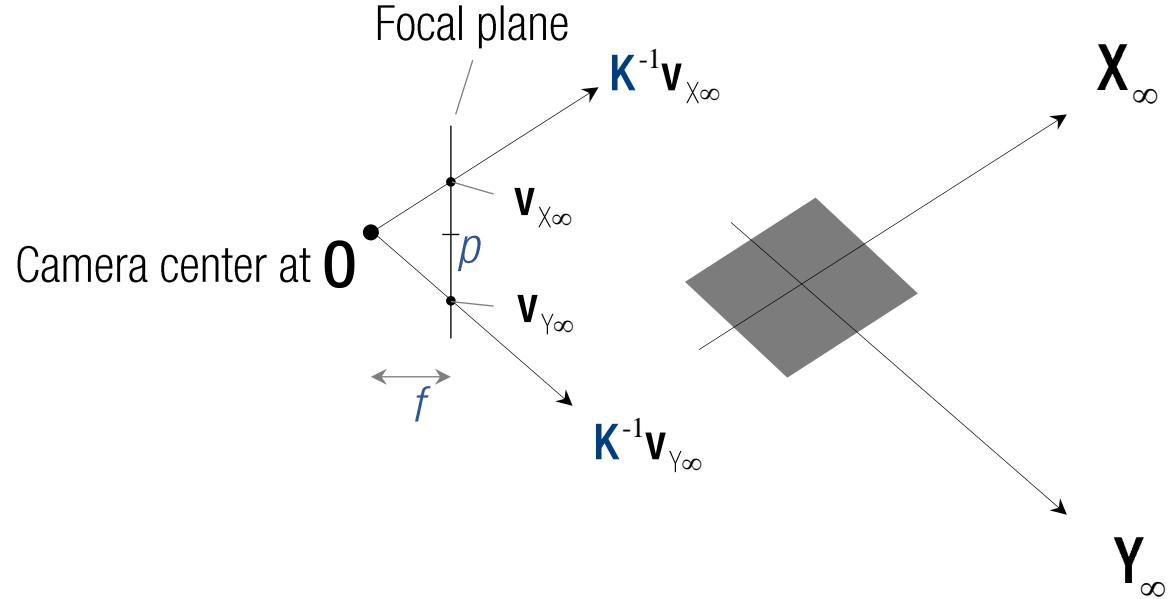
Geometric Interpretation with 1D Camera



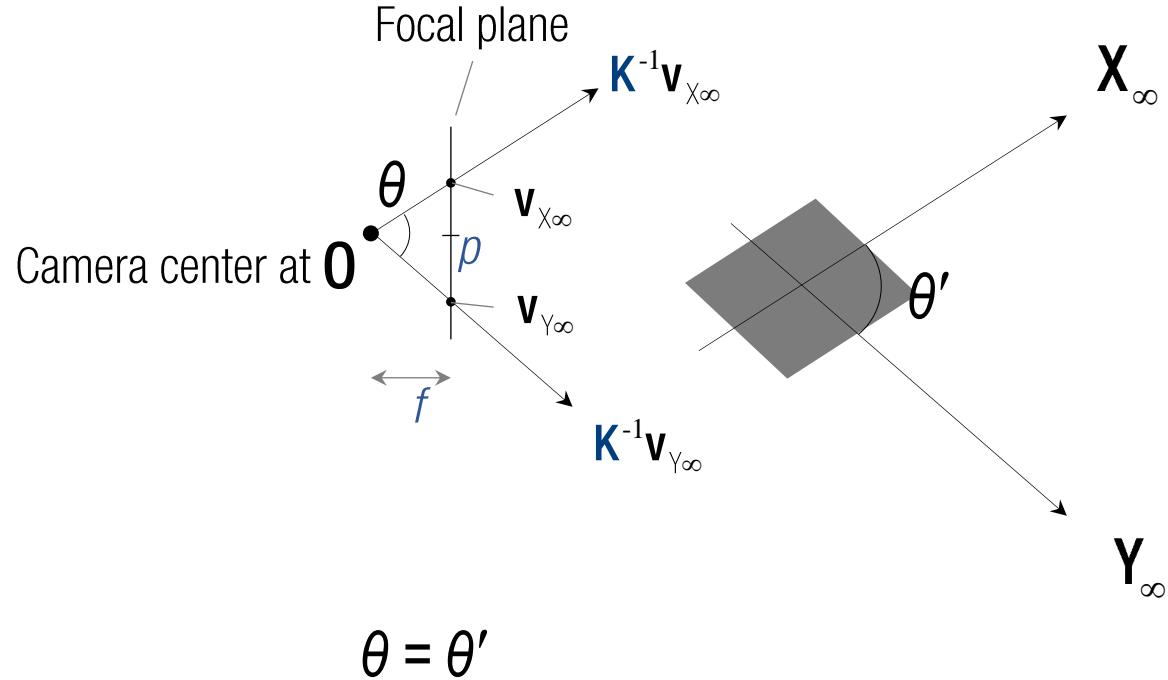
Geometric Interpretation with 1D Camera



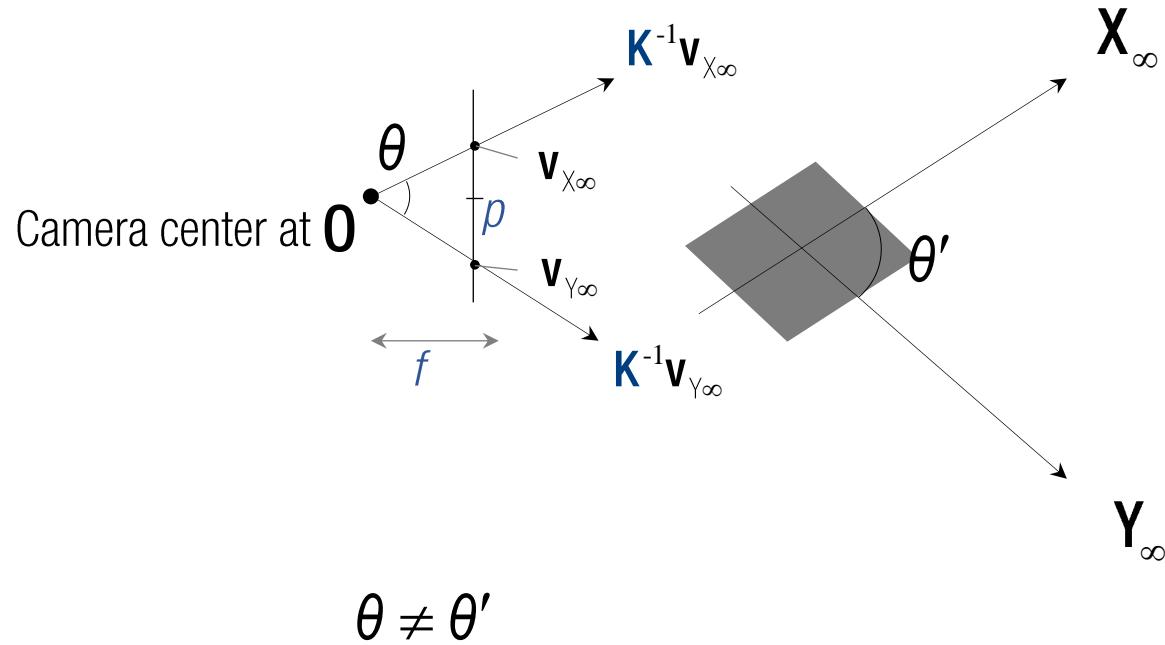
Geometric Interpretation with 1D Camera



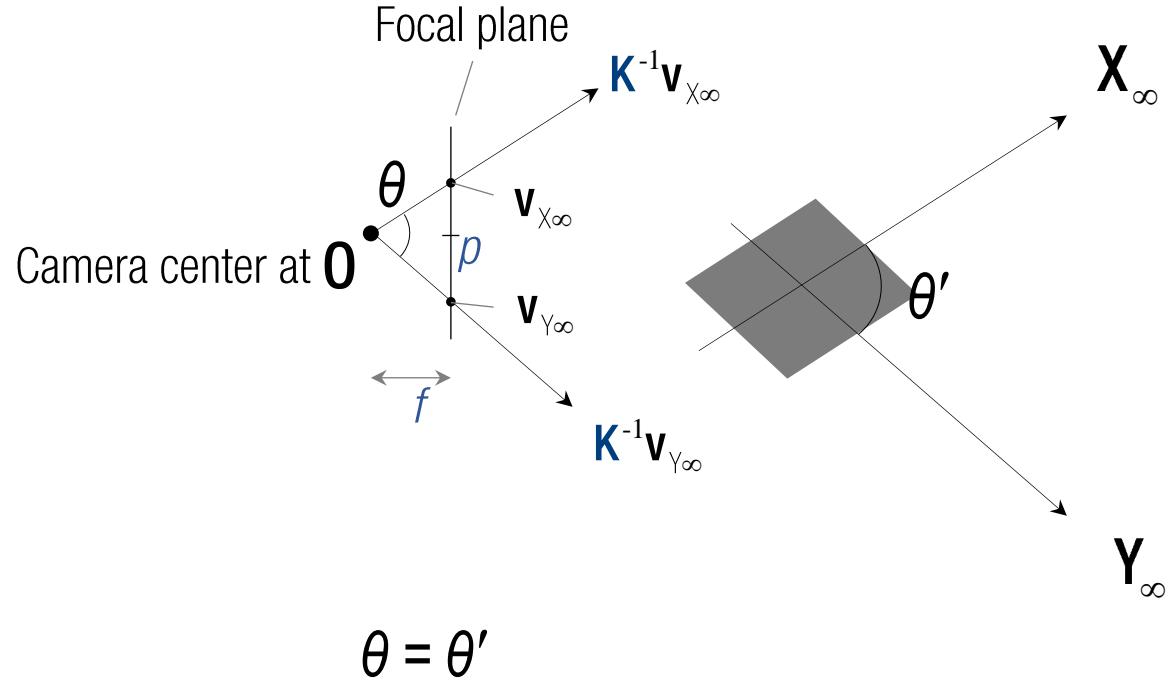
Geometric Interpretation with 1D Camera



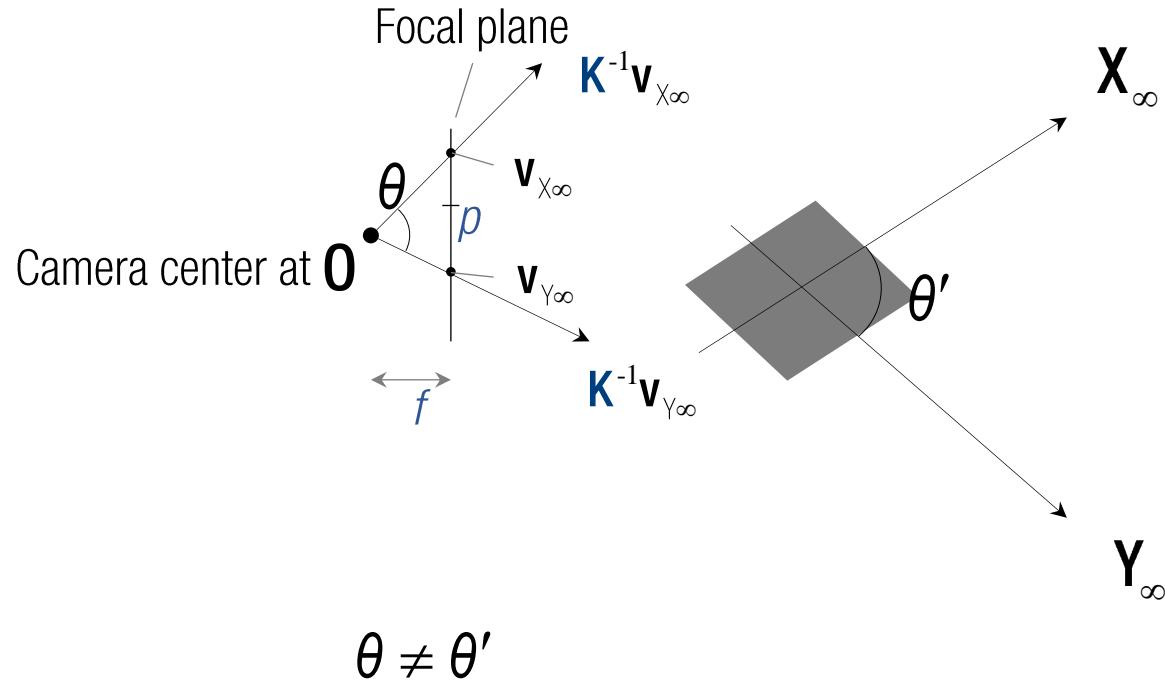
Geometric Interpretation with 1D Camera



Geometric Interpretation with 1D Camera

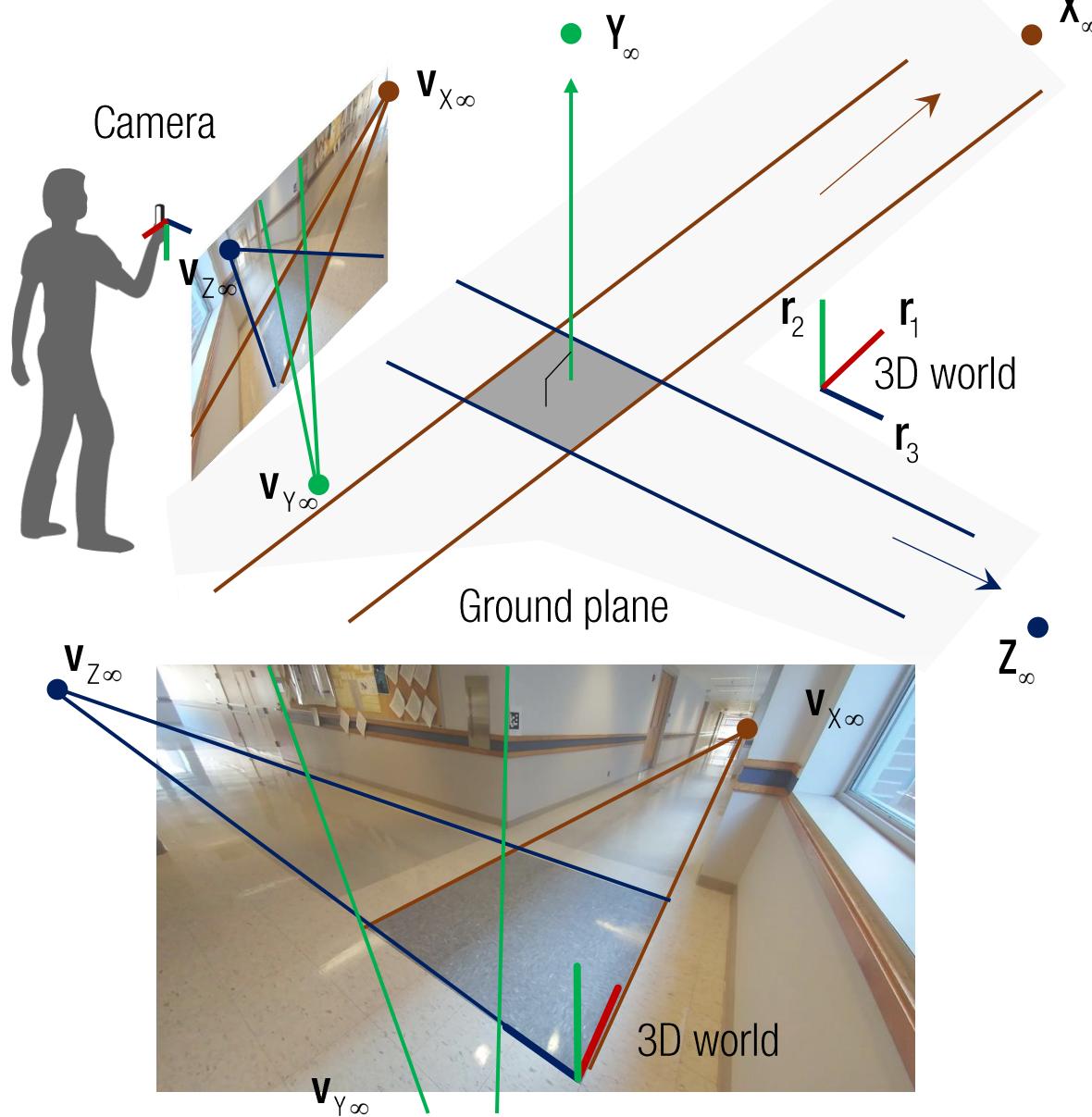


Geometric Interpretation with 1D Camera



Given two vanishing points, the focal length and principal point are uniquely defined.
For the 2D camera case, another vanishing point is needed to uniquely define f , p_x , and p_y .

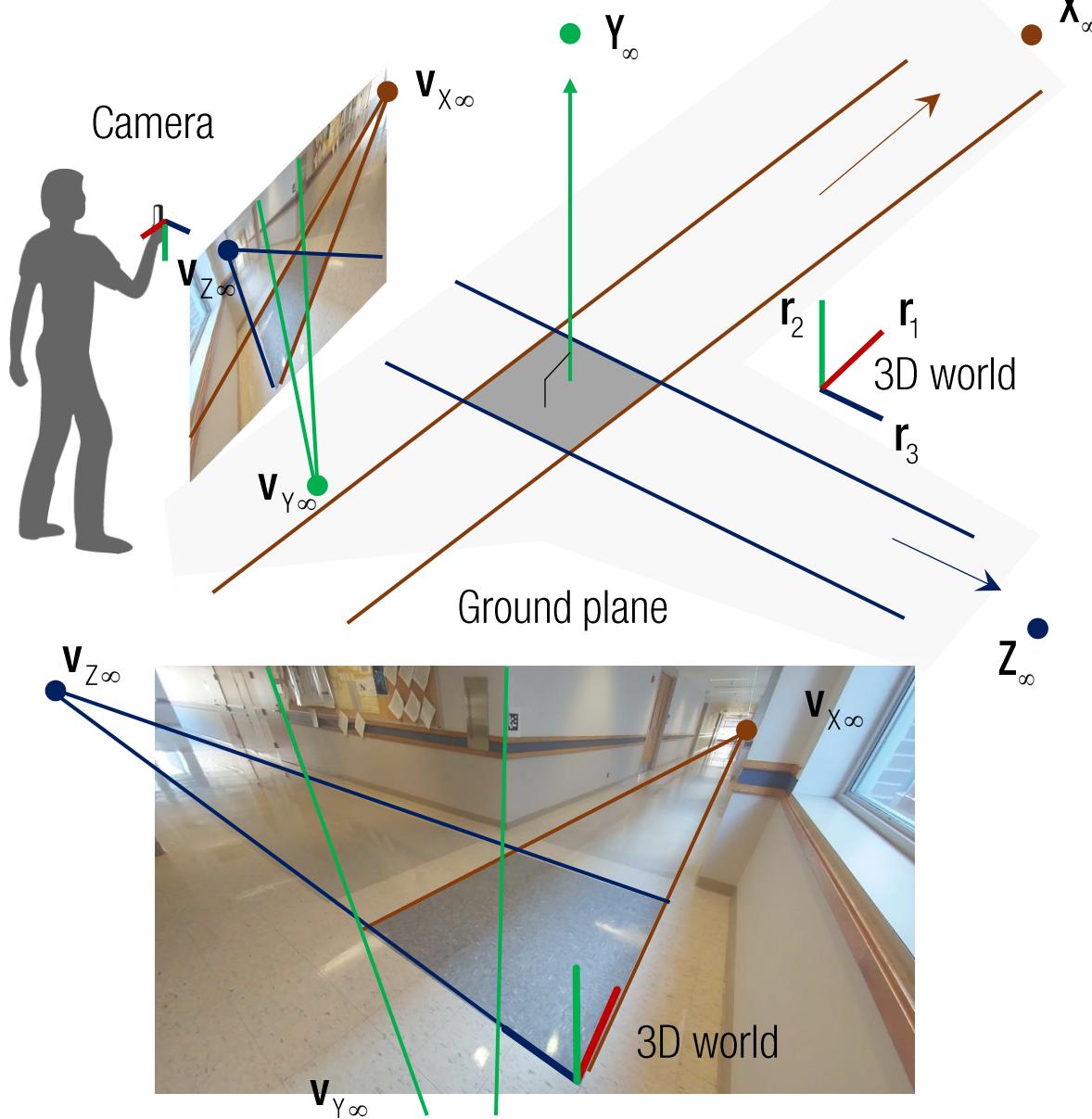
Camera Calibration using Vanishing Points



$$(\mathbf{K}^{-1}\mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X_\infty}) = 0$$

: 3 unknowns and 3 equations

Camera Calibration using Vanishing Points

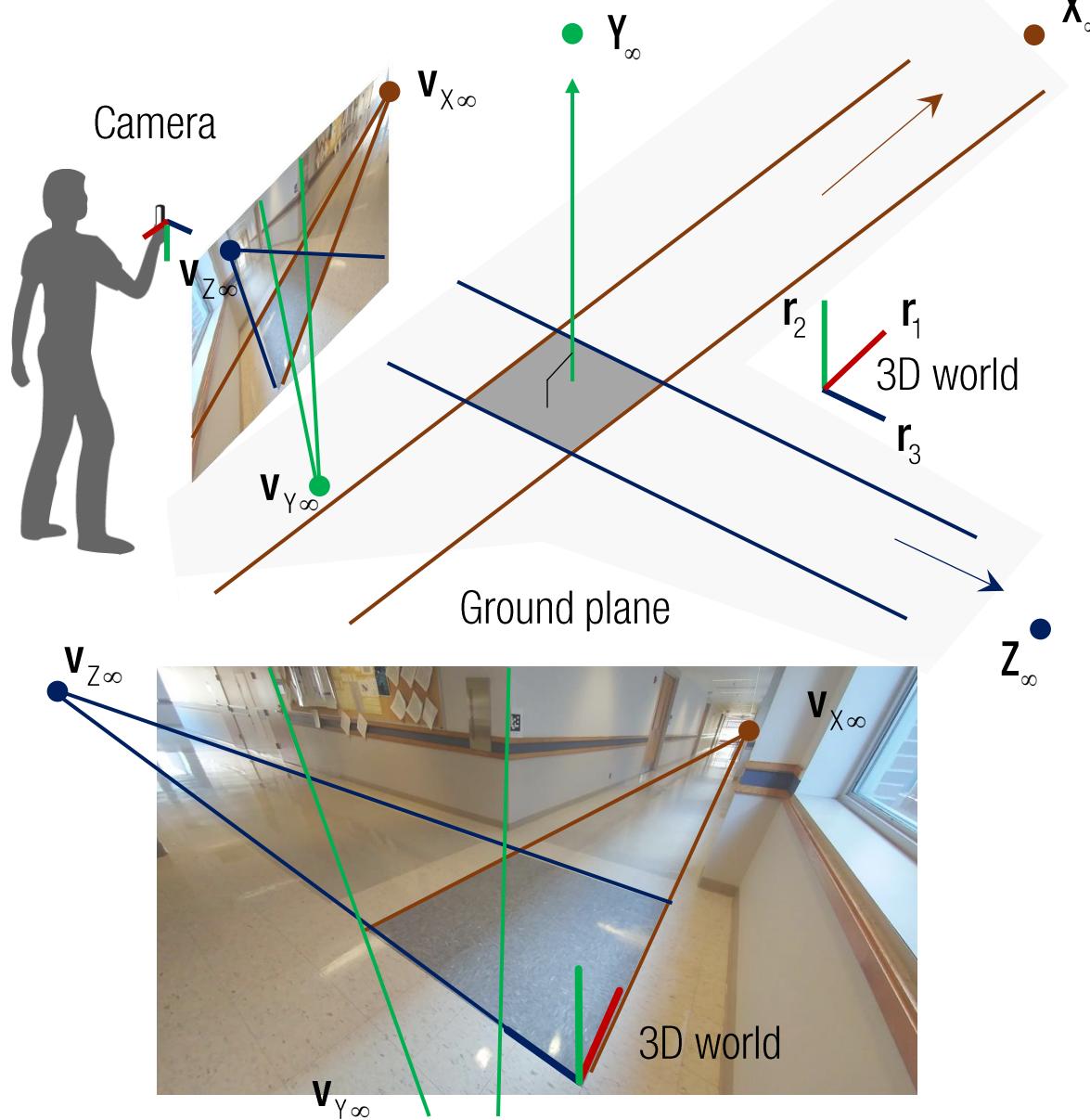


$$(\mathbf{K}^{-1}\mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X_\infty}) = 0$$

: 3 unknowns and 3 equations

$$\rightarrow (\mathbf{K}^{-1}\mathbf{v}_i)^\top (\mathbf{K}^{-1}\mathbf{v}_j) = \mathbf{v}_i^\top \mathbf{K}^{-\top} \mathbf{K}^{-1} \mathbf{v}_j \quad i, j = X, Y, Z, \text{ and } i \neq j$$

Camera Calibration using Vanishing Points



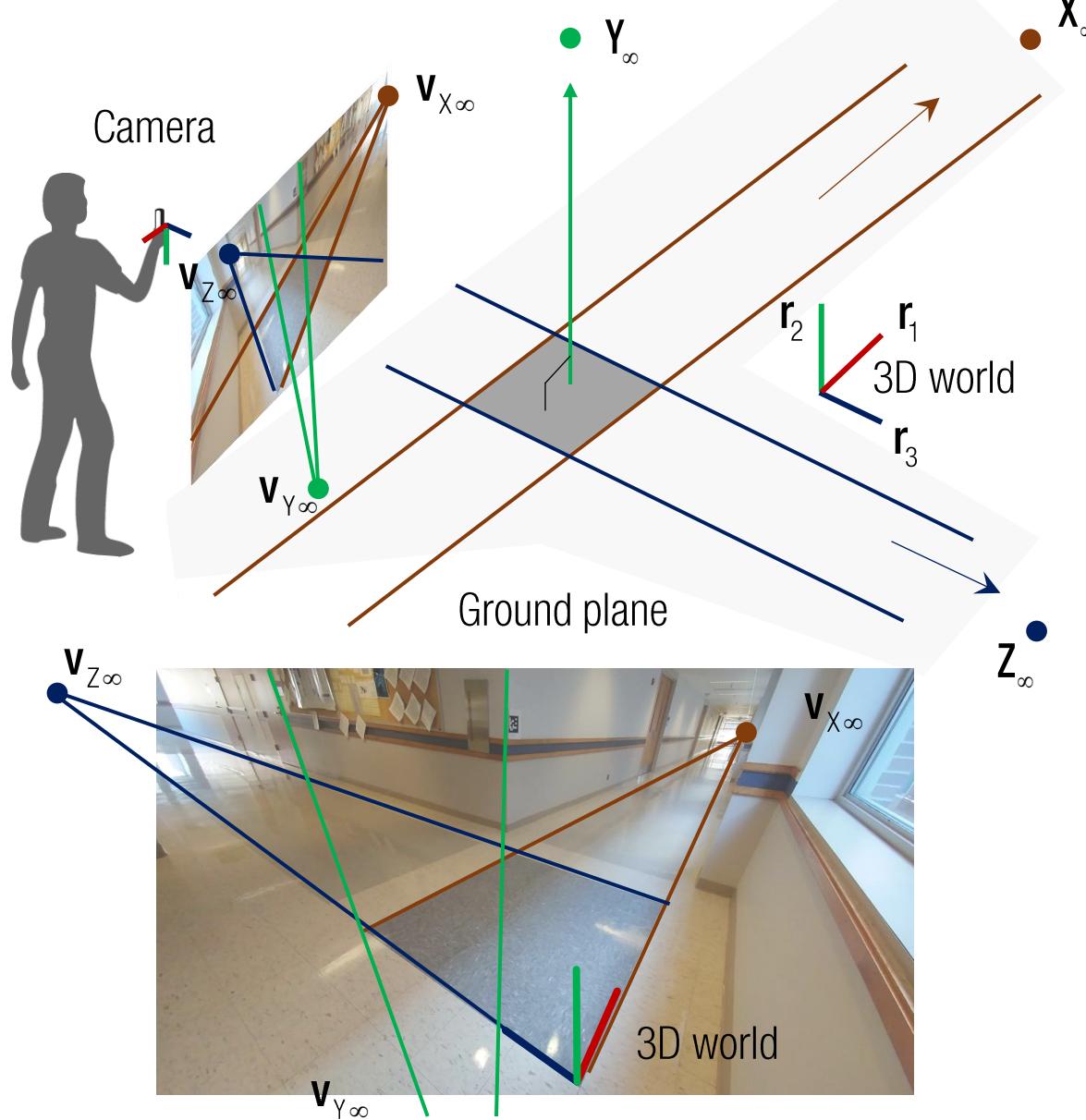
$$(\mathbf{K}^{-1}\mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X_\infty}) = 0$$

: 3 unknowns and 3 equations

$$\rightarrow (\mathbf{K}^{-1}\mathbf{v}_i)^\top (\mathbf{K}^{-1}\mathbf{v}_j) = \mathbf{v}_i^\top \mathbf{K}^{-\top} \mathbf{K}^{-1} \mathbf{v}_j \quad i, j = X, Y, Z, \text{ and } i \neq j$$

$$\mathbf{K}^{-\top} \mathbf{K}^{-1} = \begin{bmatrix} 1/f & & \\ & 1/f & \\ -p_x/f & -p_y/f & 1 \end{bmatrix} \begin{bmatrix} 1/f & -p_x/f \\ 1/f & -p_y/f \\ & 1 \end{bmatrix}$$

Camera Calibration using Vanishing Points



$$(\mathbf{K}^{-1}\mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X_\infty}) = 0$$

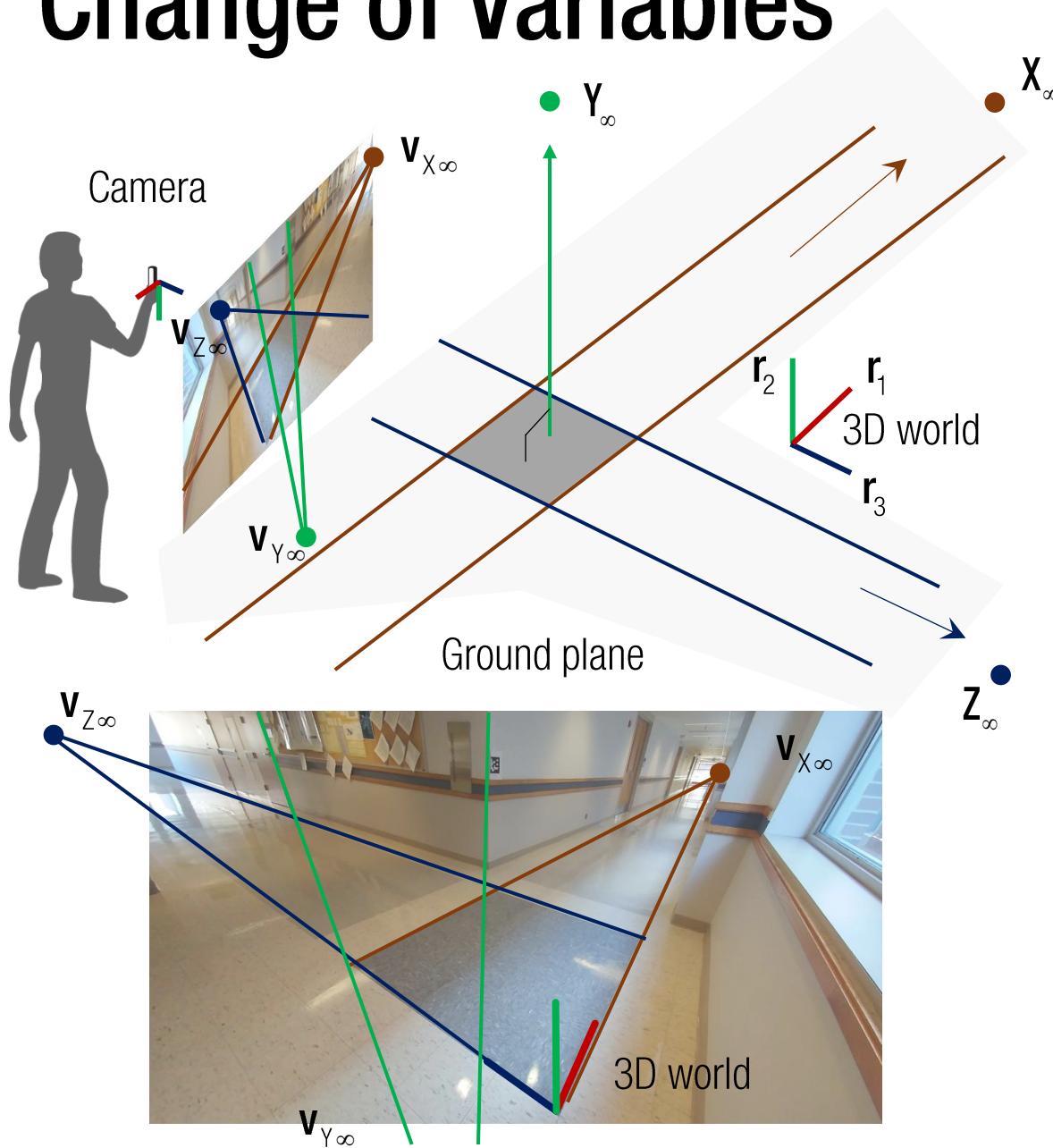
: 3 unknowns and 3 equations

$$\rightarrow (\mathbf{K}^{-1}\mathbf{v}_i)^\top (\mathbf{K}^{-1}\mathbf{v}_j) = \mathbf{v}_i^\top \mathbf{K}^{-\top} \mathbf{K}^{-1} \mathbf{v}_j \quad i, j = X, Y, Z, \text{ and } i \neq j$$

$$\mathbf{K}^{-\top} \mathbf{K}^{-1} = \begin{bmatrix} 1/f & & \\ & 1/f & \\ -p_x/f & -p_y/f & 1 \end{bmatrix} \begin{bmatrix} 1/f & -p_x/f \\ 1/f & -p_y/f \\ & 1 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{f^2} & -\frac{p_x}{f^2} \\ \frac{1}{f^2} & -\frac{p_y}{f^2} \\ -\frac{p_x}{f^2} & -\frac{p_y}{f^2} & \frac{p_x^2 + p_y^2}{f^2} + 1 \end{bmatrix}$$

Change of Variables



$$(\mathbf{K}^{-1}\mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X_\infty}) = 0$$

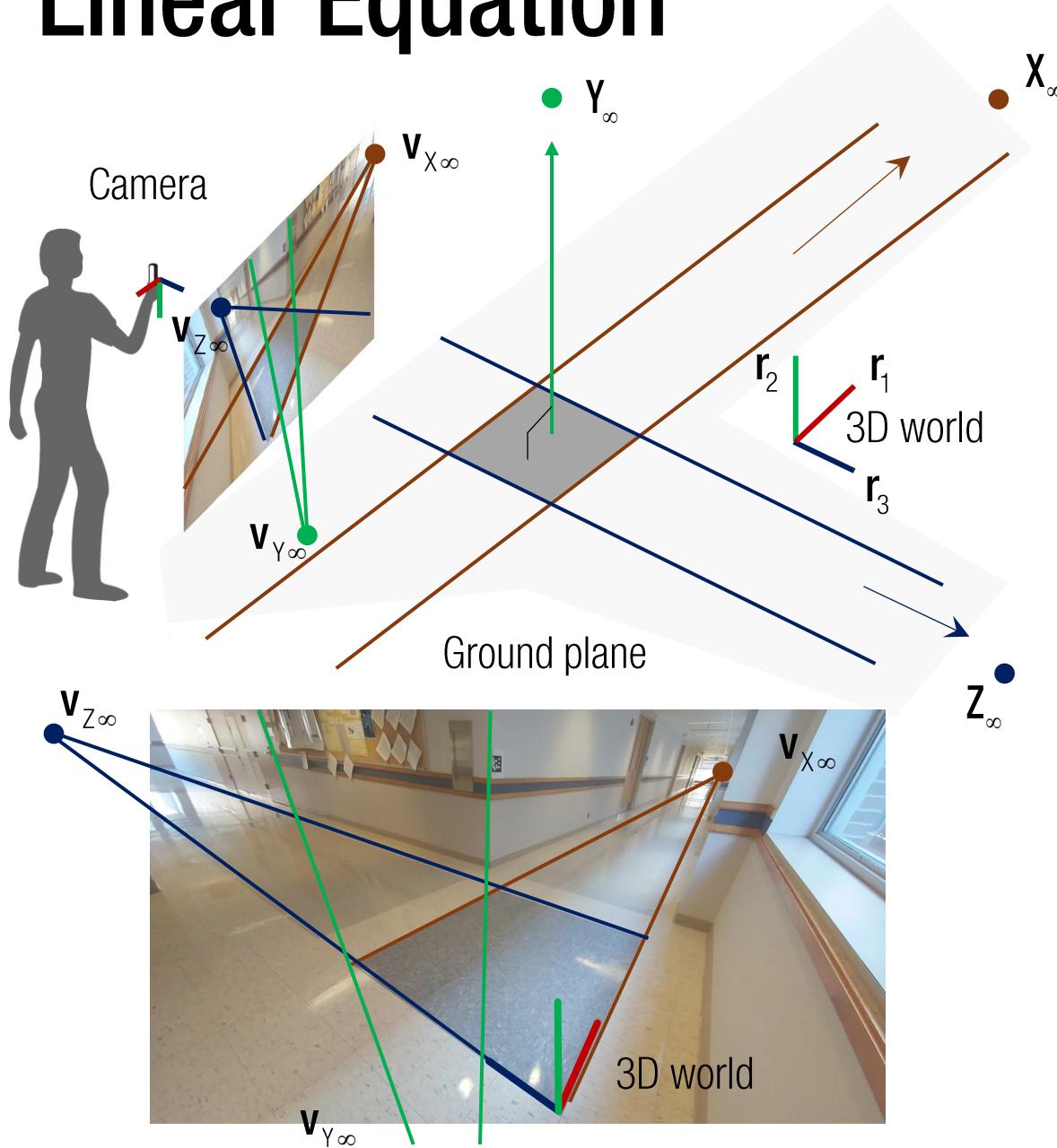
: 3 unknowns and 3 equations

$$\rightarrow (\mathbf{K}^{-1}\mathbf{v}_i)^\top (\mathbf{K}^{-1}\mathbf{v}_j) = \mathbf{v}_i^\top \mathbf{K}^{-\top} \mathbf{K}^{-1} \mathbf{v}_j \quad i, j = X, Y, Z, \text{ and } i \neq j$$

$$\begin{aligned} \mathbf{K}^{-\top} \mathbf{K}^{-1} &= \begin{bmatrix} 1/f & & \\ & 1/f & \\ -p_x/f & -p_y/f & 1 \end{bmatrix} \begin{bmatrix} 1/f & -p_x/f \\ 1/f & -p_y/f \\ & 1 \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{f^2} & -\frac{p_x}{f^2} \\ \frac{1}{f^2} & -\frac{p_y}{f^2} \\ -\frac{p_x}{f^2} & -\frac{p_y}{f^2} \frac{p_x^2}{f^2} + \frac{p_y^2}{f^2} + 1 \end{bmatrix} = \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \end{aligned}$$

$$\text{where } b_1 = \frac{1}{f^2}, \quad b_2 = -\frac{p_x}{f^2}, \quad b_3 = -\frac{p_y}{f^2}, \quad b_4 = \frac{p_x^2}{f^2} + \frac{p_y^2}{f^2} + 1$$

Linear Equation



$$(\mathbf{K}^{-1}\mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X_\infty}) = 0$$

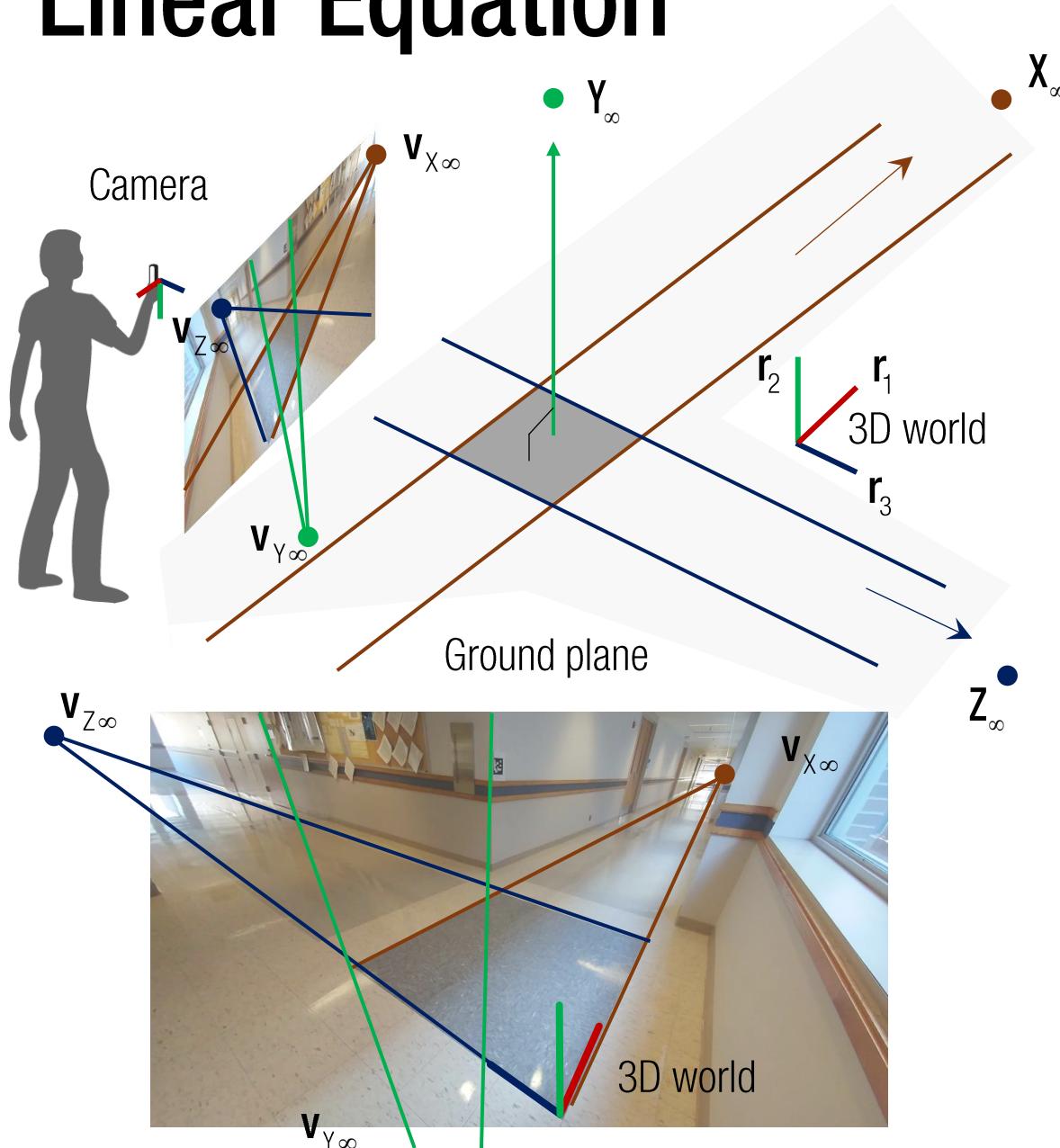
: 3 unknowns and 3 equations

$$\rightarrow (\mathbf{K}^{-1}\mathbf{v}_i)^\top (\mathbf{K}^{-1}\mathbf{v}_j) = \mathbf{v}_i^\top \mathbf{K}^{-\top} \mathbf{K}^{-1} \mathbf{v}_j \quad i, j = X, Y, Z, \text{ and } i \neq j$$

$$\rightarrow \mathbf{v}_i^\top \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \mathbf{v}_j :$$

Linear in b

Linear Equation



$$(\mathbf{K}^{-1}\mathbf{v}_{X\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X\infty}) = 0$$

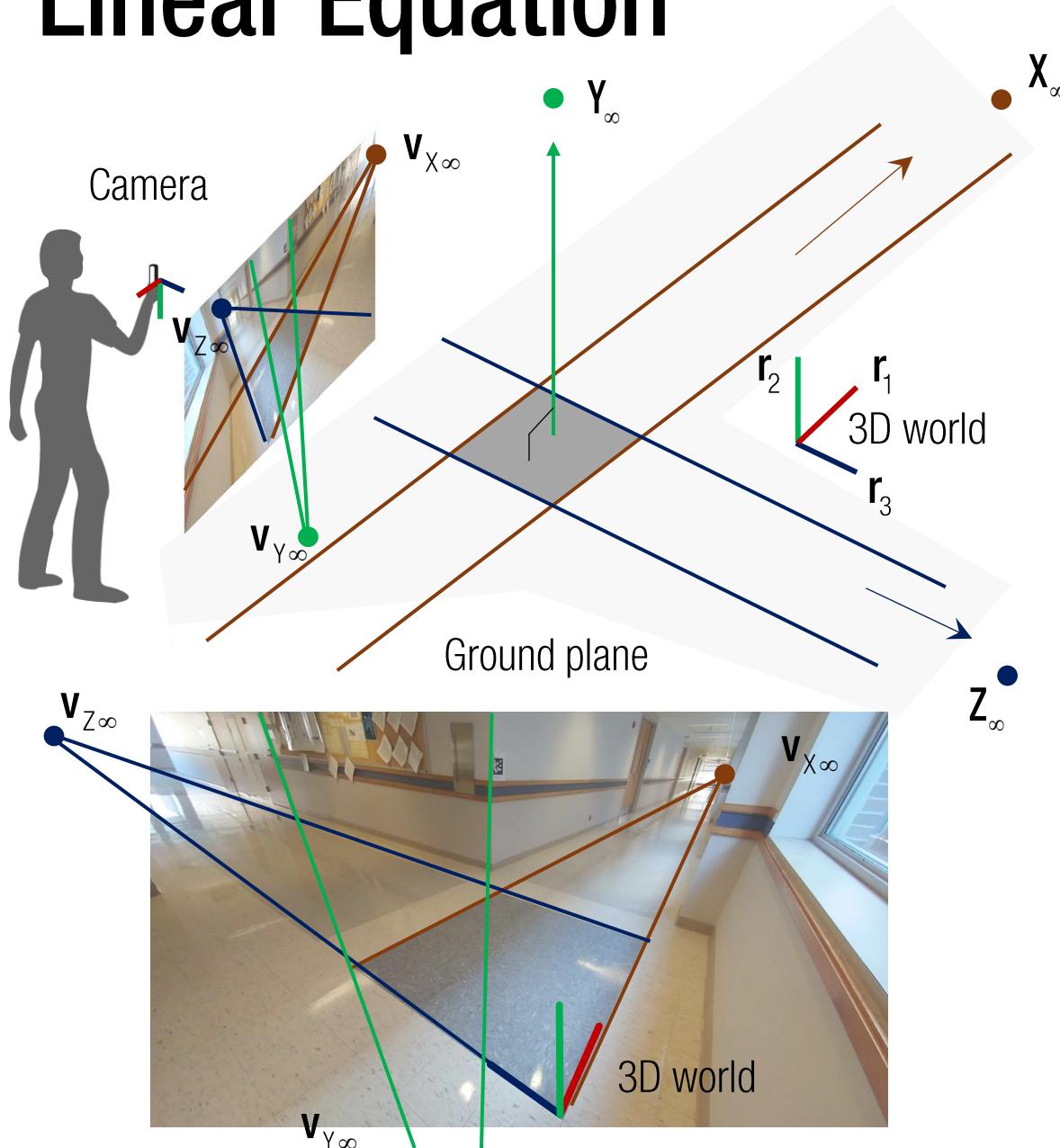
: 3 unknowns and 3 equations

$$\rightarrow (\mathbf{K}^{-1}\mathbf{v}_i)^\top (\mathbf{K}^{-1}\mathbf{v}_j) = \mathbf{v}_i^\top \mathbf{K}^{-\top} \mathbf{K}^{-1} \mathbf{v}_j \quad i, j = X, Y, Z, \text{ and } i \neq j$$

$$\rightarrow \mathbf{v}_i^\top \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \mathbf{v}_j = \begin{bmatrix} u_i \\ v_i \\ 1 \end{bmatrix}^\top \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_4 \end{bmatrix} \begin{bmatrix} u_j \\ v_j \\ 1 \end{bmatrix} = 0$$

Linear in b

Linear Equation



$$(\mathbf{K}^{-1}\mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X_\infty}) = 0$$

: 3 unknowns and 3 equations

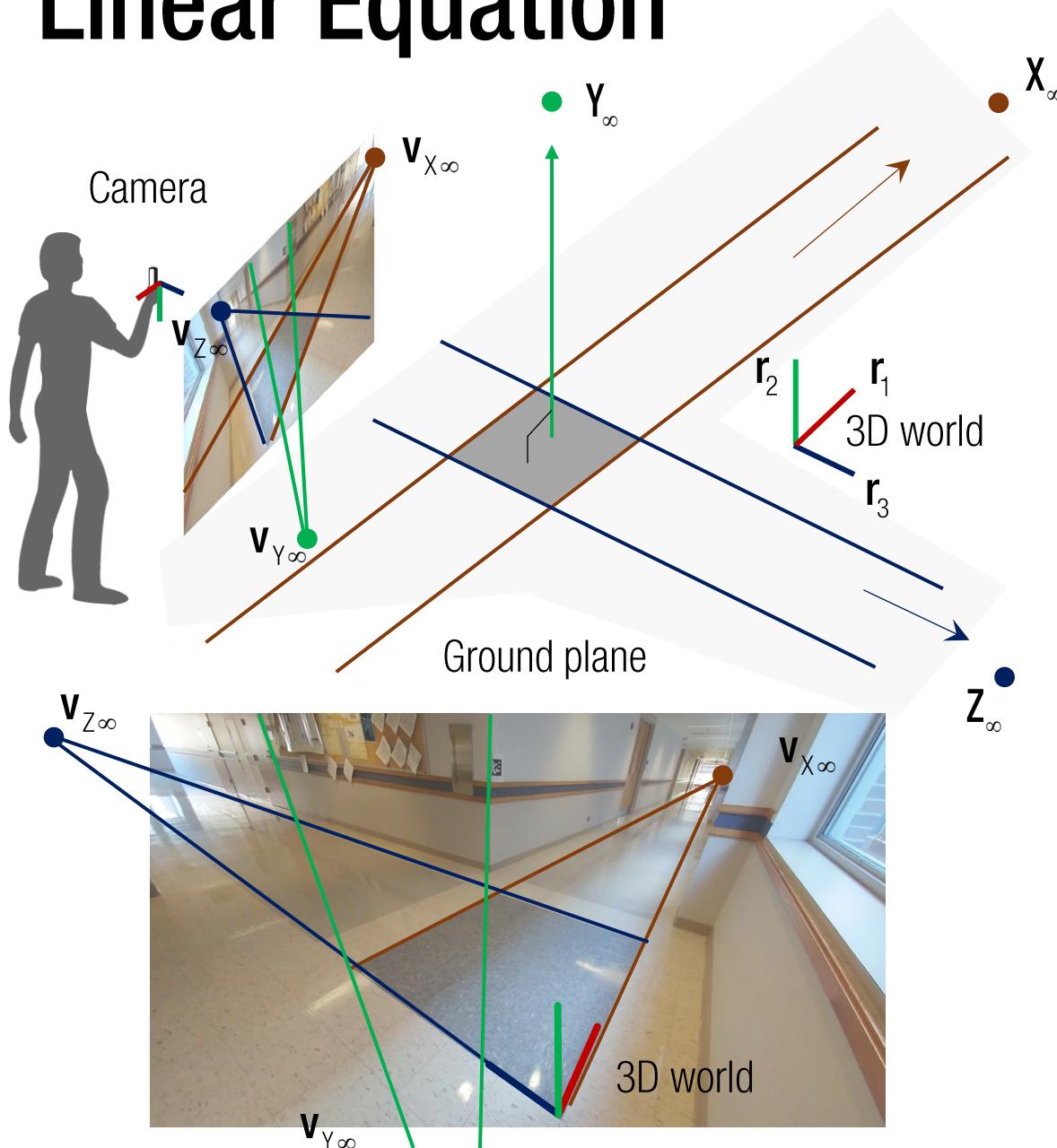
$$\rightarrow (\mathbf{K}^{-1}\mathbf{v}_i)^\top (\mathbf{K}^{-1}\mathbf{v}_j) = \mathbf{v}_i^\top \mathbf{K}^{-\top} \mathbf{K}^{-1} \mathbf{v}_j \quad i, j = X, Y, Z, \text{ and } i \neq j$$

$$\rightarrow \mathbf{v}_i^\top \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \mathbf{v}_j = \begin{bmatrix} u_i \\ v_i \\ 1 \end{bmatrix}^\top \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_4 \end{bmatrix} \begin{bmatrix} u_j \\ v_j \\ 1 \end{bmatrix} = 0$$

Linear in b

$$\rightarrow \begin{bmatrix} u_i u_j + v_i v_j & u_i + u_j & v_i + v_j & 1 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = 0$$

Linear Equation



$$(\mathbf{K}^{-1}\mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X_\infty}) = 0$$

: 3 unknowns and 3 equations

$$\rightarrow (\mathbf{K}^{-1}\mathbf{v}_i)^\top (\mathbf{K}^{-1}\mathbf{v}_j) = \mathbf{v}_i^\top \mathbf{K}^{-\top} \mathbf{K}^{-1} \mathbf{v}_j \quad i, j = X, Y, Z, \text{ and } i \neq j$$

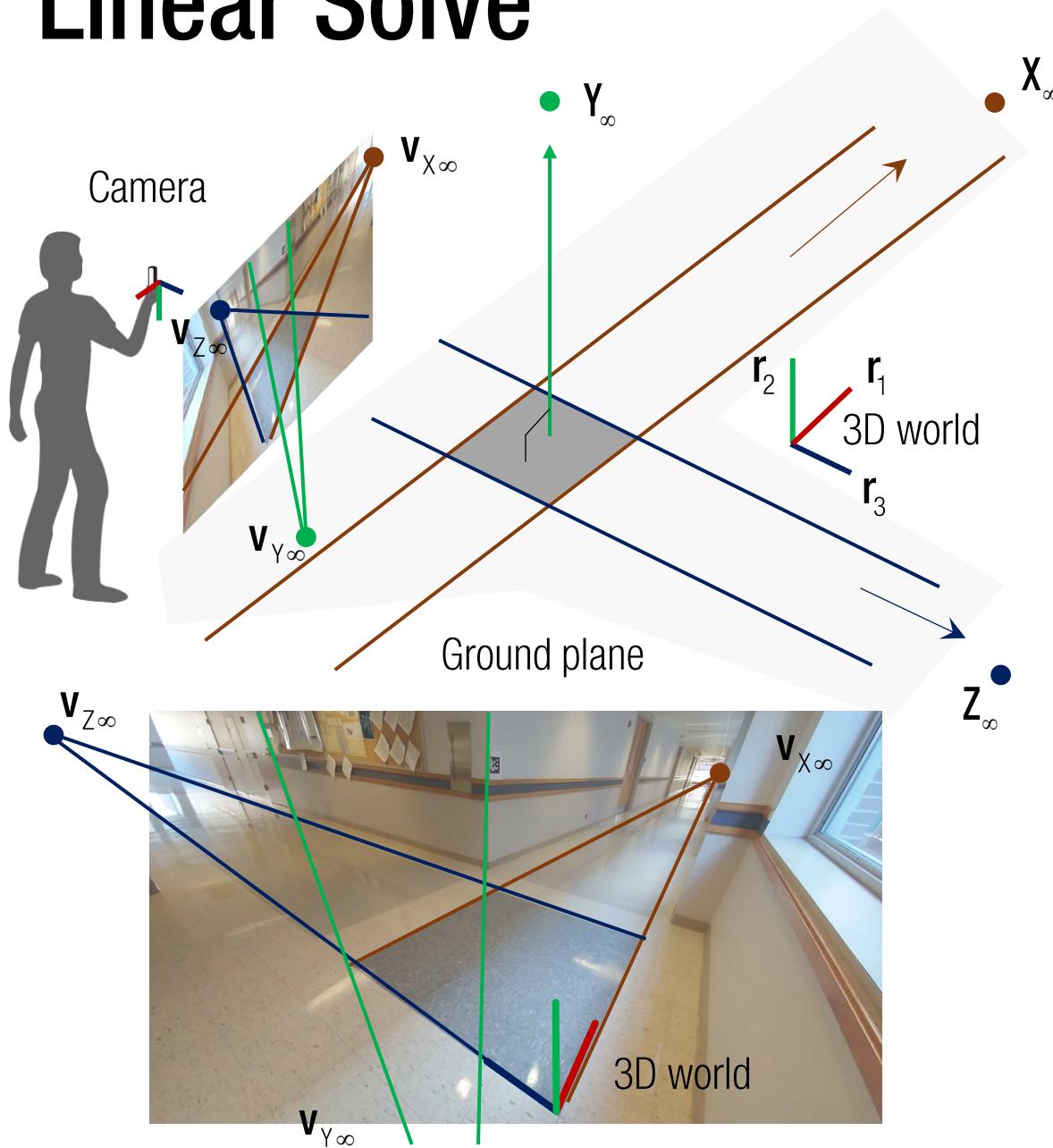
$$\rightarrow \mathbf{v}_i^\top \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \mathbf{v}_j = \begin{bmatrix} u_i \\ v_i \\ 1 \end{bmatrix}^\top \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_4 \end{bmatrix} \begin{bmatrix} u_j \\ v_j \\ 1 \end{bmatrix} = 0$$

Linear in b

$$\rightarrow \begin{bmatrix} u_1u_2 + v_1v_2 & u_1 + u_2 & v_1 + v_2 & 1 \\ u_3u_2 + v_3v_2 & u_3 + u_2 & v_3 + v_2 & 1 \\ u_1u_3 + v_1v_3 & u_1 + u_3 & v_1 + v_3 & 1 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = 0$$

3x4

Linear Solve



$$(\mathbf{K}^{-1}\mathbf{v}_{X_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z_\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X_\infty}) = 0$$

: 3 unknowns and 3 equations

$$\rightarrow (\mathbf{K}^{-1}\mathbf{v}_i)^\top (\mathbf{K}^{-1}\mathbf{v}_j) = \mathbf{v}_i^\top \mathbf{K}^{-\top} \mathbf{K}^{-1} \mathbf{v}_j \quad i, j = X, Y, Z, \text{ and } i \neq j$$

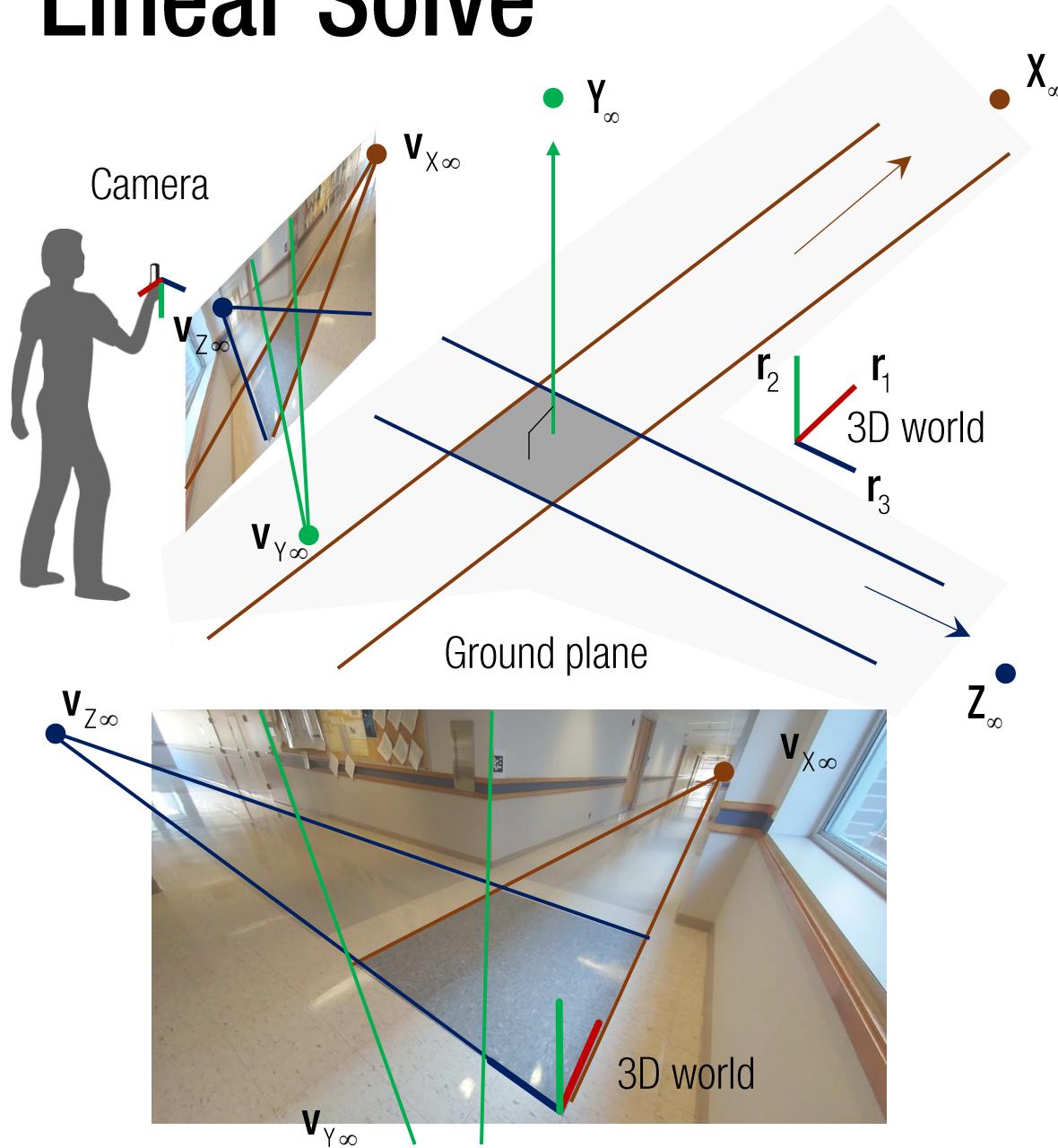
$$\rightarrow \mathbf{v}_i^\top \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \mathbf{v}_j = \begin{bmatrix} u_i \\ v_i \\ 1 \end{bmatrix}^\top \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_4 \end{bmatrix} \begin{bmatrix} u_j \\ v_j \\ 1 \end{bmatrix} = 0$$

Linear in b

$$\rightarrow \begin{bmatrix} u_1 u_2 + v_1 v_2 & u_1 + u_2 & v_1 + v_2 & 1 \\ u_3 u_2 + v_3 v_2 & u_3 + u_2 & v_3 + v_2 & 1 \\ u_1 u_3 + v_1 v_3 & u_1 + u_3 & v_1 + v_3 & 1 \end{bmatrix} \mathbf{A} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = \mathbf{0}$$

3x4

Linear Solve



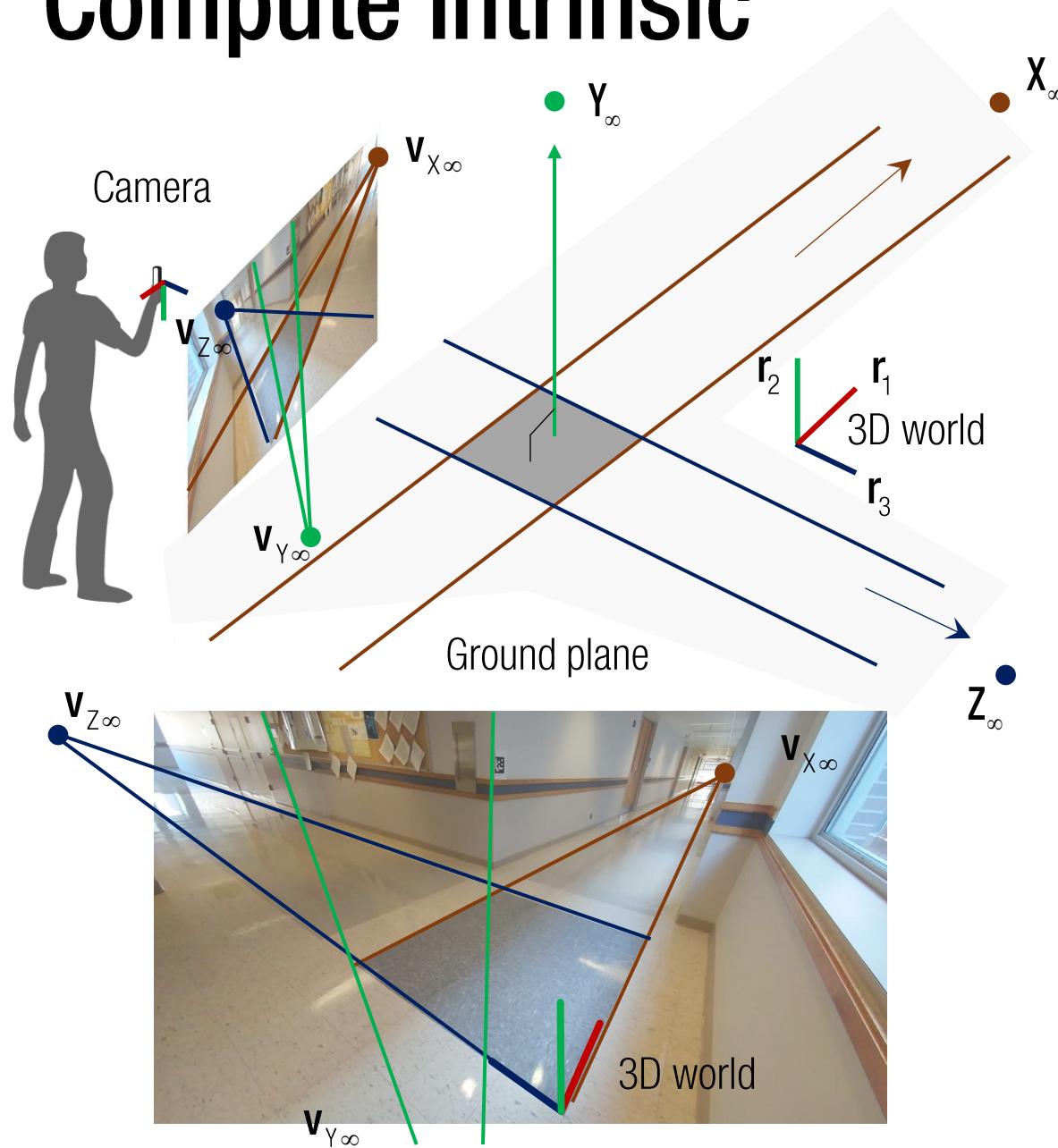
$$(\mathbf{K}^{-1}v_{X_\infty})^T (\mathbf{K}^{-1}v_{Y_\infty}) = (\mathbf{K}^{-1}v_{Y_\infty})^T (\mathbf{K}^{-1}v_{Z_\infty}) = (\mathbf{K}^{-1}v_{Z_\infty})^T (\mathbf{K}^{-1}v_{X_\infty}) = 0$$

: 3 unknowns and 3 equations

$$\rightarrow \mathbf{x} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}$$

$$\text{where } b_1 = \frac{1}{f^2}, \quad b_2 = -\frac{p_x}{f^2}, \quad b_3 = -\frac{p_y}{f^2}, \quad b_4 = \frac{p_x^2}{f^2} + \frac{p_y^2}{f^2} + 1$$

Compute Intrinsic



$$(\mathbf{K}^{-1}\mathbf{v}_{X\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Y\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Y\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{Z\infty}) = (\mathbf{K}^{-1}\mathbf{v}_{Z\infty})^\top (\mathbf{K}^{-1}\mathbf{v}_{X\infty}) = 0$$

: 3 unknowns and 3 equations

$$\rightarrow \mathbf{x} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}$$

$$\text{where } b_1 = \frac{1}{f^2}, \quad b_2 = -\frac{p_x}{f^2}, \quad b_3 = -\frac{p_y}{f^2}, \quad b_4 = \frac{p_x^2}{f^2} + \frac{p_y^2}{f^2} + 1$$

$$\rightarrow \quad p_x = -\frac{b_2}{b_1}, \quad p_y = -\frac{b_3}{b_1}, \quad f = \sqrt{\frac{b_4}{b_1} - (p_x^2 + p_y^2)}$$

Camera Calibration

```
function CameraCalibration
```

```
m11 = [2145;2120;1];m12 = [2566;1191;1];
```

```
m13 = [1804;935;1];m14 = [1050;1320;1];
```

```
z11 = [1772; 364; 1];z12 = [1778; 823; 1];
```

```
z21 = [2564; 31; 1];z22 = [2439; 551; 1];
```

```
m21 = m11;m22 = m14;m23 = m12;m24 = m13;
```

```
|l11 = GetLineFromTwoPoints(m11,m12);
```

```
|l12 = GetLineFromTwoPoints(m13,m14);
```

```
|l21 = GetLineFromTwoPoints(m21,m22);
```

```
|l22 = GetLineFromTwoPoints(m23,m24);
```

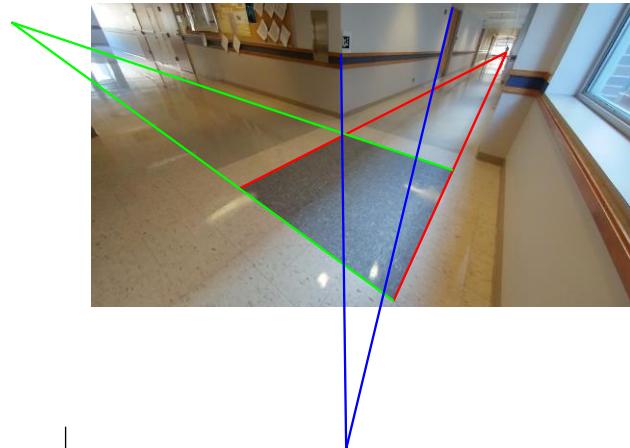
```
|l31 = GetLineFromTwoPoints(z11,z12);
```

```
|l32 = GetLineFromTwoPoints(z21,z22);
```

```
x = GetPointFromTwoLines(l11,l12);
```

```
y = GetPointFromTwoLines(l21,l22);
```

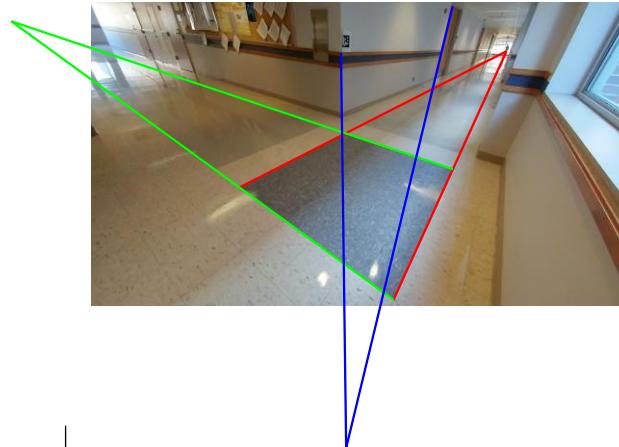
```
z = GetPointFromTwoLines(l31,l32);
```



Vanishing points

Camera Calibration

```
function CameraCalibration  
  
m11 = [2145;2120;1];m12 = [2566;1191;1];  
m13 = [1804;935;1];m14 = [1050;1320;1];  
  
z11 = [1772; 364; 1];z12 = [1778; 823; 1];  
z21 = [2564; 31; 1];z22 = [2439; 551; 1];  
  
m21 = m11;m22 = m14;m23 = m12;m24 = m13;  
  
l11 = GetLineFromTwoPoints(m11,m12);  
l12 = GetLineFromTwoPoints(m13,m14);  
  
l21 = GetLineFromTwoPoints(m21,m22);  
l22 = GetLineFromTwoPoints(m23,m24);  
  
l31 = GetLineFromTwoPoints(z11,z12);  
l32 = GetLineFromTwoPoints(z21,z22);  
  
x = GetPointFromTwoLines(l11,l12);  
y = GetPointFromTwoLines(l21,l22);  
z = GetPointFromTwoLines(l31,l32);
```



$$\mathbf{A} = \begin{bmatrix} u_1u_2 + v_1v_2 & u_1 + u_2 & v_1 + v_2 & 1 \\ u_3u_2 + v_3v_2 & u_3 + u_2 & v_3 + v_2 & 1 \\ u_1u_3 + v_1v_3 & u_1 + u_3 & v_1 + v_3 & 1 \end{bmatrix}$$

```
A = [x(1)*y(1)+x(2)*y(2) x(1)+y(1) x(2)+y(2) 1;  
z(1)*y(1)+z(2)*y(2) z(1)+y(1) z(2)+y(2) 1;  
x(1)*z(1)+x(2)*z(2) x(1)+z(1) x(2)+z(2) 1];
```

```
[u d v] = svd(A);  
x = v(:,end);
```

```
px = -x(2)/x(1);  
py = -x(3)/x(1);  
f = sqrt(x(4)/x(1)-px^2-py^2);
```

```
K = [f 0 px;  
0 f py;  
0 0 1]
```

K =

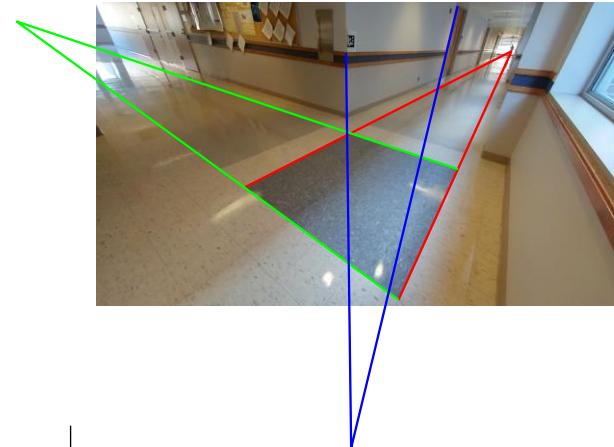
1317.2	0	1931.8
0	1317.2	1146.1
0	0	1

$f = 1500$
 $px = \text{size}(im,2)/2 = 1920$
 $py = \text{size}(im,1)/2 = 1080$

Linear solve
using SVD

Camera Calibration

```
function CameraCalibration  
  
m11 = [2145;2120;1];m12 = [2566;1191;1];  
m13 = [1804;935;1];m14 = [1050;1320;1];  
  
z11 = [1772; 364; 1];z12 = [1778; 823; 1];  
z21 = [2564; 31; 1];z22 = [2439; 551; 1];  
  
m21 = m11;m22 = m14;m23 = m12;m24 = m13;  
  
l11 = GetLineFromTwoPoints(m11,m12);  
l12 = GetLineFromTwoPoints(m13,m14);  
  
l21 = GetLineFromTwoPoints(m21,m22);  
l22 = GetLineFromTwoPoints(m23,m24);  
  
l31 = GetLineFromTwoPoints(z11,z12);  
l32 = GetLineFromTwoPoints(z21,z22);  
  
x = GetPointFromTwoLines(l11,l12);  
y = GetPointFromTwoLines(l21,l22);  
z = GetPointFromTwoLines(l31,l32);
```



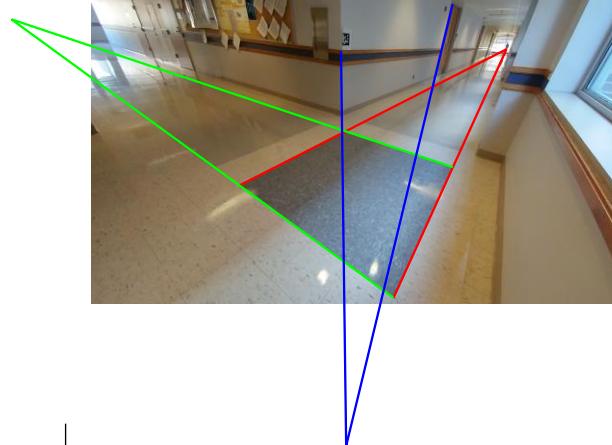
Vanishing points

$$\mathbf{A} = \begin{bmatrix} u_1u_2 + v_1v_2 & u_1 + u_2 & v_1 + v_2 & 1 \\ u_3u_2 + v_3v_2 & u_3 + u_2 & v_3 + v_2 & 1 \\ u_1u_3 + v_1v_3 & u_1 + u_3 & v_1 + v_3 & 1 \end{bmatrix}$$

$$\begin{aligned} \mathbf{A} = & [x(1)*y(1)+x(2)*y(2) \quad x(1)+y(1) \quad x(2)+y(2) \quad 1; \\ & z(1)*y(1)+z(2)*y(2) \quad z(1)+y(1) \quad z(2)+y(2) \quad 1; \\ & x(1)*z(1)+x(2)*z(2) \quad x(1)+z(1) \quad x(2)+z(2) \quad 1]; \end{aligned}$$

Camera Calibration

```
function CameraCalibration  
  
m11 = [2145;2120;1];m12 = [2566;1191;1];  
m13 = [1804;935;1];m14 = [1050;1320;1];  
  
z11 = [1772; 364; 1];z12 = [1778; 823; 1];  
z21 = [2564; 31; 1];z22 = [2439; 551; 1];  
  
m21 = m11;m22 = m14;m23 = m12;m24 = m13;  
  
l11 = GetLineFromTwoPoints(m11,m12);  
l12 = GetLineFromTwoPoints(m13,m14);  
  
l21 = GetLineFromTwoPoints(m21,m22);  
l22 = GetLineFromTwoPoints(m23,m24);  
  
l31 = GetLineFromTwoPoints(z11,z12);  
l32 = GetLineFromTwoPoints(z21,z22);  
  
x = GetPointFromTwoLines(l11,l12);  
y = GetPointFromTwoLines(l21,l22);  
z = GetPointFromTwoLines(l31,l32);
```



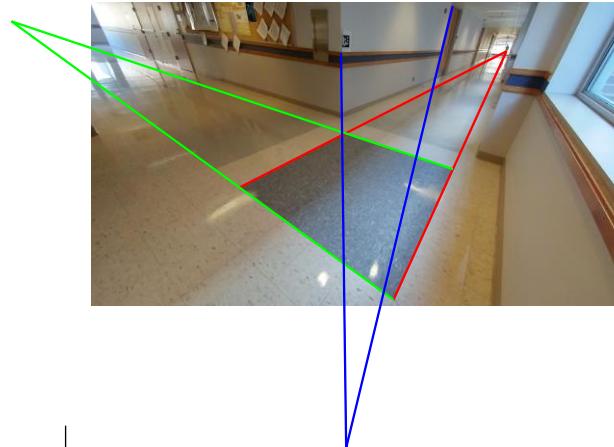
$$\mathbf{A} = \begin{bmatrix} u_1u_2 + v_1v_2 & u_1 + u_2 & v_1 + v_2 & 1 \\ u_3u_2 + v_3v_2 & u_3 + u_2 & v_3 + v_2 & 1 \\ u_1u_3 + v_1v_3 & u_1 + u_3 & v_1 + v_3 & 1 \end{bmatrix}$$

```
A = [x(1)*y(1)+x(2)*y(2) x(1)+y(1) x(2)+y(2) 1;  
z(1)*y(1)+z(2)*y(2) z(1)+y(1) z(2)+y(2) 1;  
x(1)*z(1)+x(2)*z(2) x(1)+z(1) x(2)+z(2) 1];  
  
[u d v] = svd(A);  
x = v(:,end);
```

Linear solve
using SVD

Camera Calibration

```
function CameraCalibration  
  
m11 = [2145;2120;1];m12 = [2566;1191;1];  
m13 = [1804;935;1];m14 = [1050;1320;1];  
  
z11 = [1772; 364; 1];z12 = [1778; 823; 1];  
z21 = [2564; 31; 1];z22 = [2439; 551; 1];  
  
m21 = m11;m22 = m14;m23 = m12;m24 = m13;  
  
l11 = GetLineFromTwoPoints(m11,m12);  
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l21 = GetLineFromTwoPoints(m21,m22);  
l22 = GetLineFromTwoPoints(m23,m24);  
  
l31 = GetLineFromTwoPoints(z11,z12);  
l32 = GetLineFromTwoPoints(z21,z22);  
  
x = GetPointFromTwoLines(l11,l12);  
y = GetPointFromTwoLines(l21,l22);  
z = GetPointFromTwoLines(l31,l32);
```



$$\mathbf{A} = \begin{bmatrix} u_1u_2 + v_1v_2 & u_1 + u_2 & v_1 + v_2 & 1 \\ u_3u_2 + v_3v_2 & u_3 + u_2 & v_3 + v_2 & 1 \\ u_1u_3 + v_1v_3 & u_1 + u_3 & v_1 + v_3 & 1 \end{bmatrix}$$

```
A = [x(1)*y(1)+x(2)*y(2) x(1)+y(1) x(2)+y(2) 1;  
z(1)*y(1)+z(2)*y(2) z(1)+y(1) z(2)+y(2) 1;  
x(1)*z(1)+x(2)*z(2) x(1)+z(1) x(2)+z(2) 1];
```

```
[u d v] = svd(A);  
x = v(:,end);
```

```
px = -x(2)/x(1);  
py = -x(3)/x(1);  
f = sqrt(x(4)/x(1)-px^2-py^2);
```

```
K = [f 0 px;  
0 f py;  
0 0 1]
```

Linear solve
using SVD

$$p_x = -\frac{b_2}{b_1}, \quad p_y = -\frac{b_3}{b_1}, \quad f = \sqrt{\frac{b_4}{b_1} - (p_x^2 + p_y^2)}$$

Camera Calibration

```

function CameraCalibration

m11 = [2145;2120;1];m12 = [2566;1191;1];
m13 = [1804;935;1];m14 = [1050;1320;1];

z11 = [1772; 364; 1];z12 = [1778; 823; 1];
z21 = [2564; 31; 1];z22 = [2439; 551; 1];

m21 = m11;m22 = m14;m23 = m12;m24 = m13;

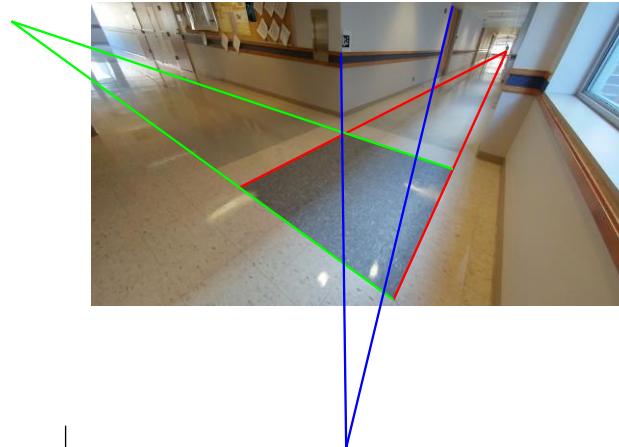
l11 = GetLineFromTwoPoints(m11,m12);
l12 = GetLineFromTwoPoints(m13,m14);

l21 = GetLineFromTwoPoints(m21,m22);
l22 = GetLineFromTwoPoints(m23,m24);

l31 = GetLineFromTwoPoints(z11,z12);
l32 = GetLineFromTwoPoints(z21,z22);

x = GetPointFromTwoLines(l11,l12);
y = GetPointFromTwoLines(l21,l22);
z = GetPointFromTwoLines(l31,l32);

```



CameraCalibration.m

$$\mathbf{A} = \begin{bmatrix} u_1u_2 + v_1v_2 & u_1 + u_2 & v_1 + v_2 & 1 \\ u_3u_2 + v_3v_2 & u_3 + u_2 & v_3 + v_2 & 1 \\ u_1u_3 + v_1v_3 & u_1 + u_3 & v_1 + v_3 & 1 \end{bmatrix}$$

$$\begin{aligned} \mathbf{A} = & [x(1)*y(1)+x(2)*y(2) \quad x(1)+y(1) \quad x(2)+y(2) \quad 1; \\ & z(1)*y(1)+z(2)*y(2) \quad z(1)+y(1) \quad z(2)+y(2) \quad 1; \\ & x(1)*z(1)+x(2)*z(2) \quad x(1)+z(1) \quad x(2)+z(2) \quad 1]; \end{aligned}$$

$$\begin{aligned} [\mathbf{u} \ \mathbf{d} \ \mathbf{v}] &= \text{svd}(\mathbf{A}); \\ \mathbf{x} &= \mathbf{v}(:,\text{end}); \end{aligned}$$

$$\begin{aligned} px &= -x(2)/x(1); \\ py &= -x(3)/x(1); \\ f &= \sqrt{x(4)/x(1)-px^2-py^2}; \end{aligned}$$

$$\mathbf{K} = \begin{bmatrix} f & 0 & px; \\ 0 & f & py; \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{K} =$$

$$\begin{array}{ccc} 1317.2 & 0 & 1931.8 \\ 0 & 1317.2 & 1146.1 \\ 0 & 0 & 1 \end{array}$$

$$f = 1224$$

$$px = \text{size(im,2)}/2 = 1920$$

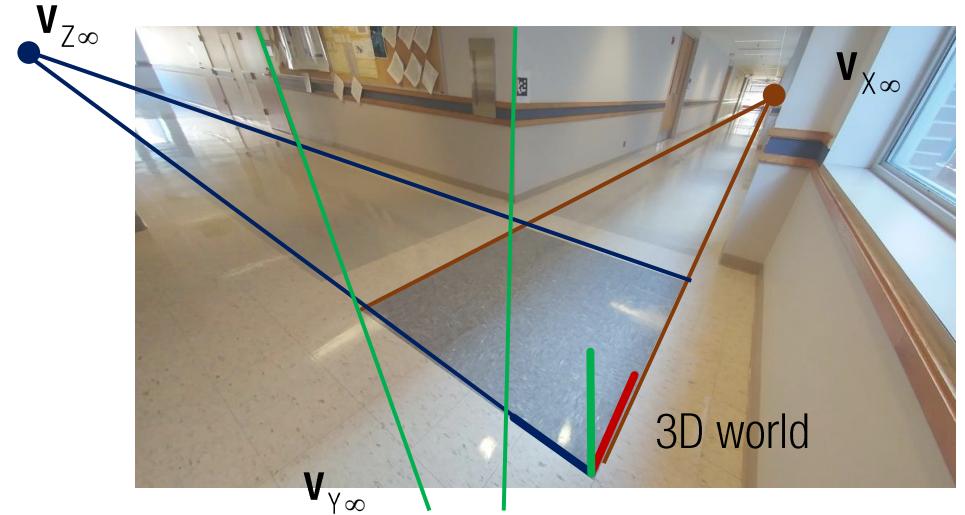
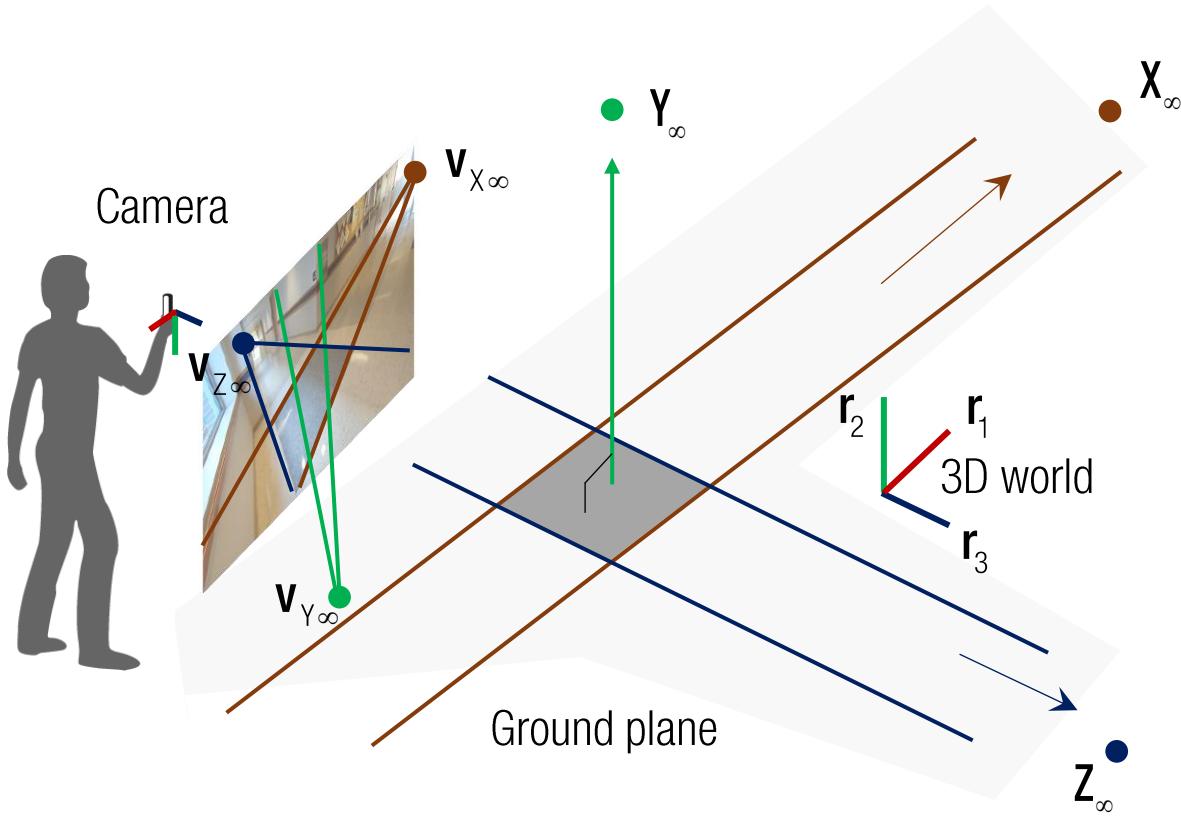
$$py = \text{size(im,1)}/2 = 1080$$

Linear solve
using SVD

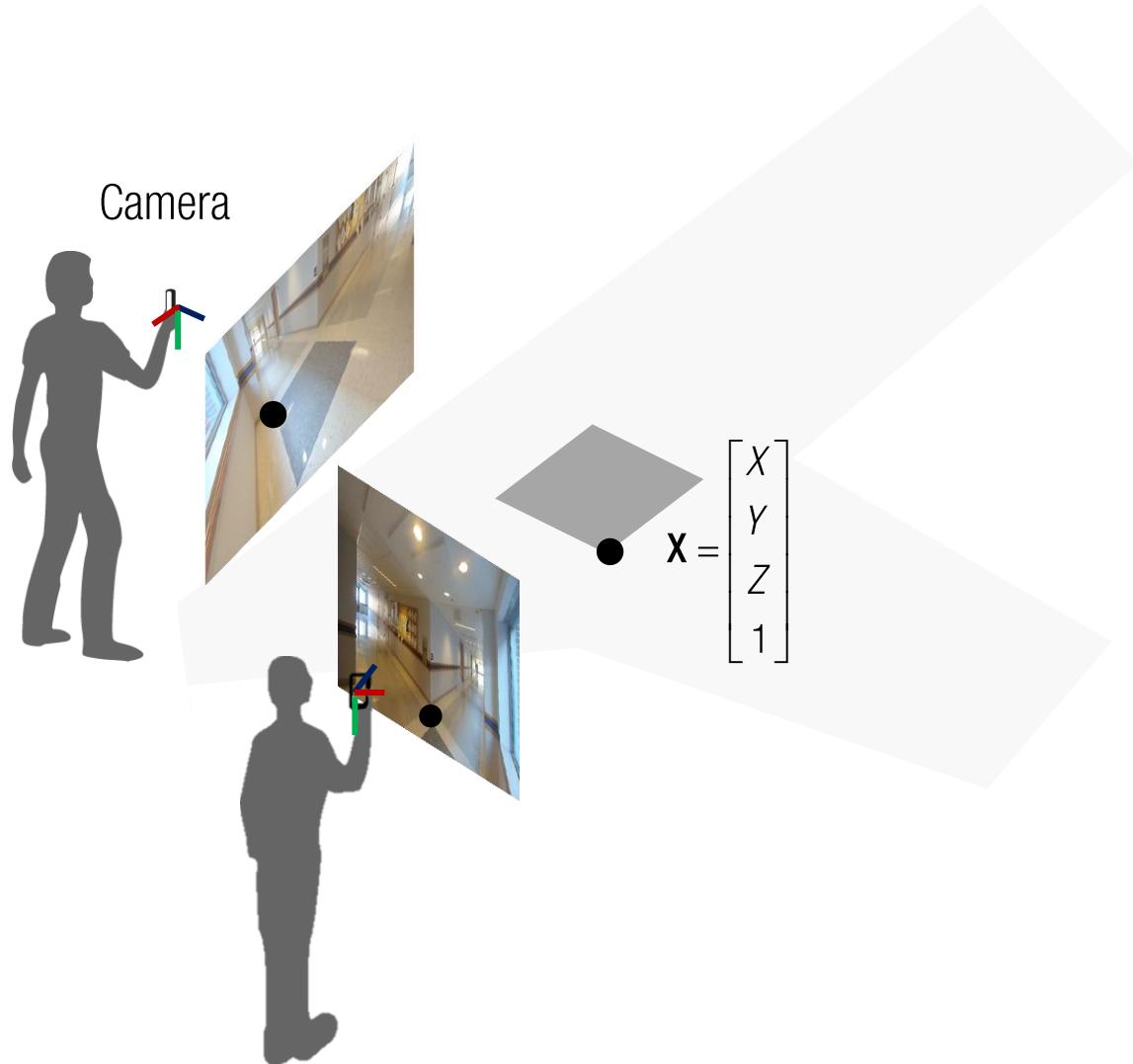
$$p_x = -\frac{b_2}{b_1}, \quad p_y = -\frac{b_3}{b_1}, \quad f = \sqrt{\frac{b_4}{b_1} - (p_x^2 + p_y^2)}$$

Previous manual estimate

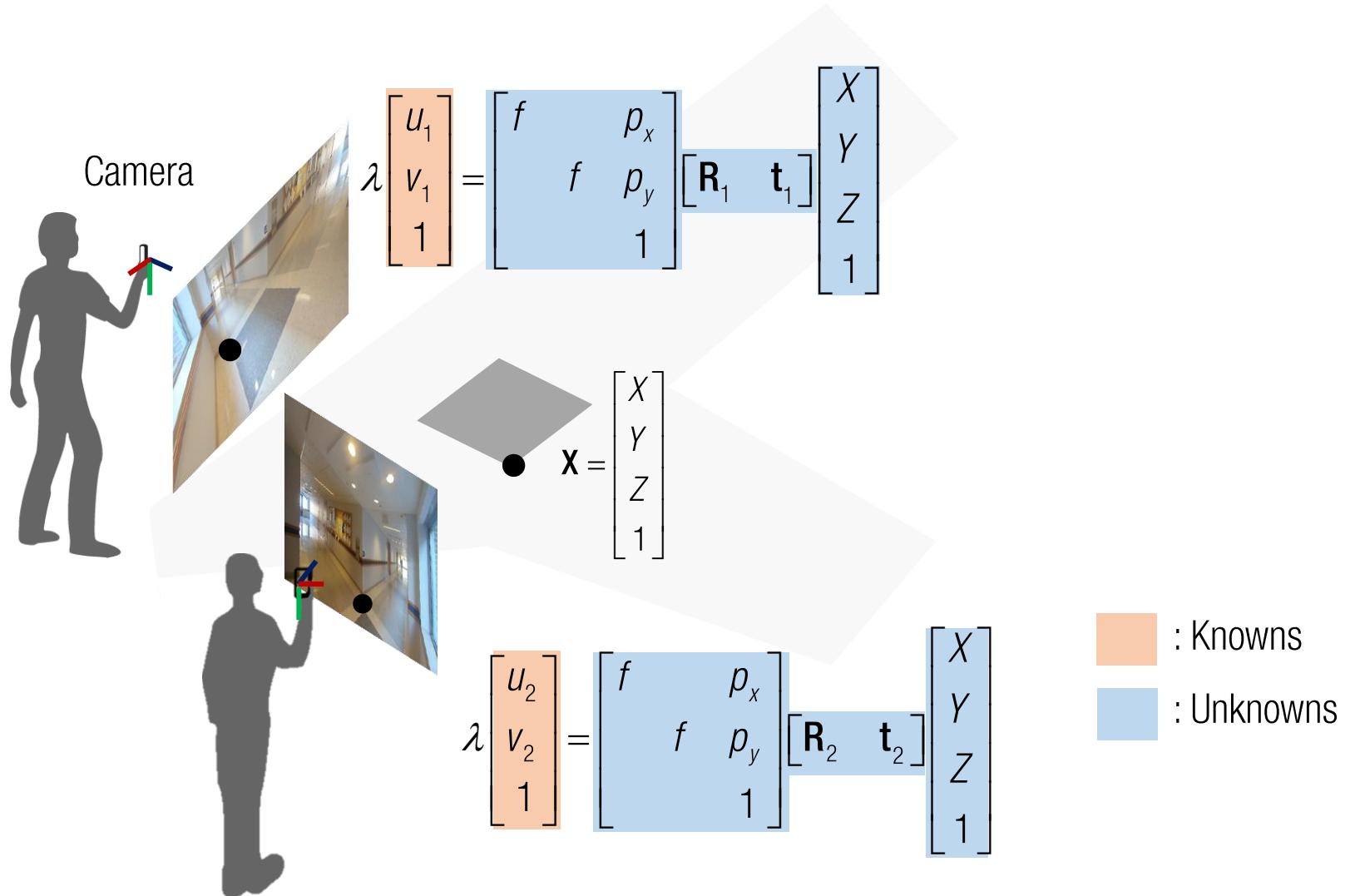
Camera Calibration via Vanishing Points



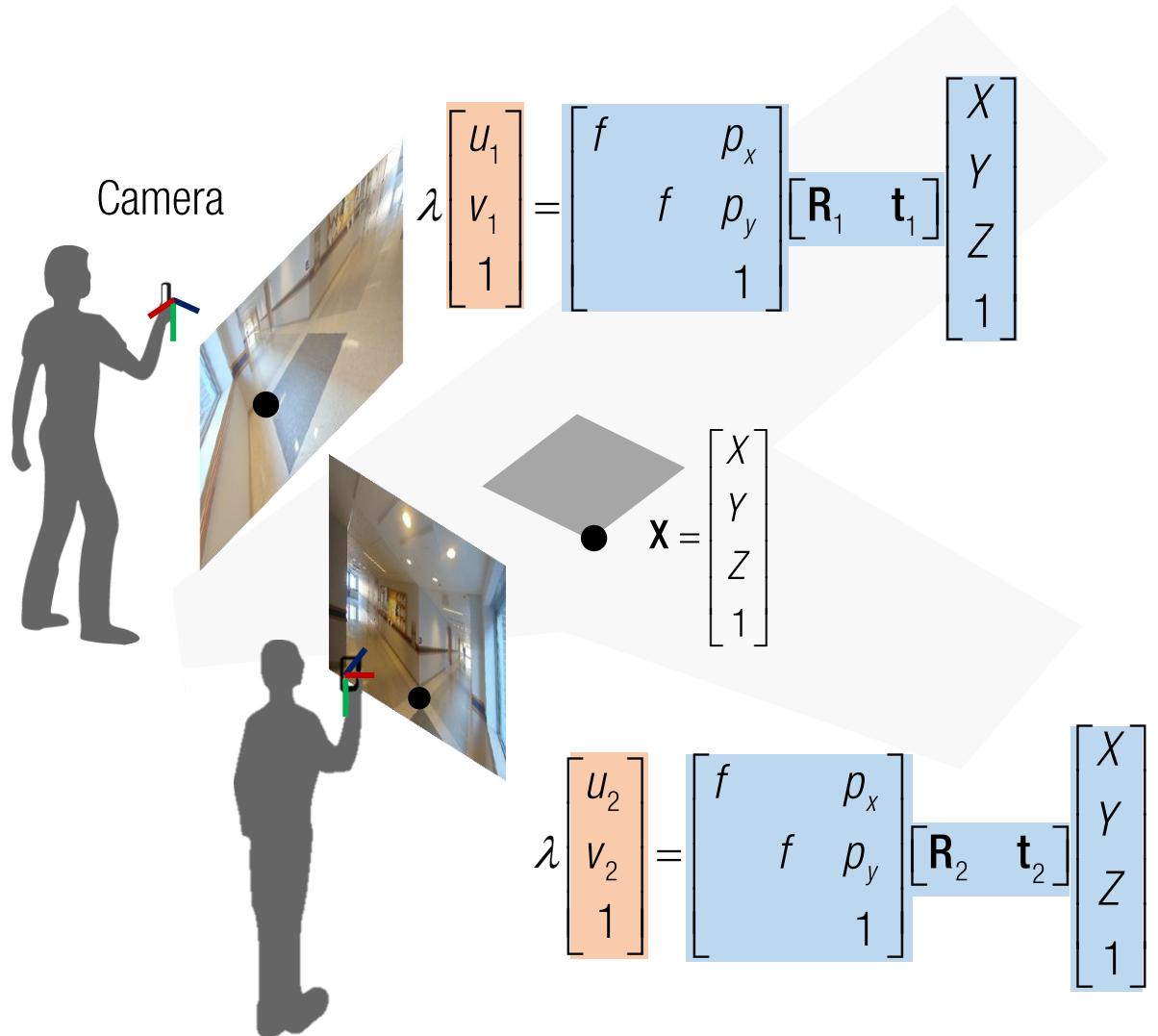
Multiview Camera Calibration



Multiview Camera Calibration



Multiview Camera Calibration



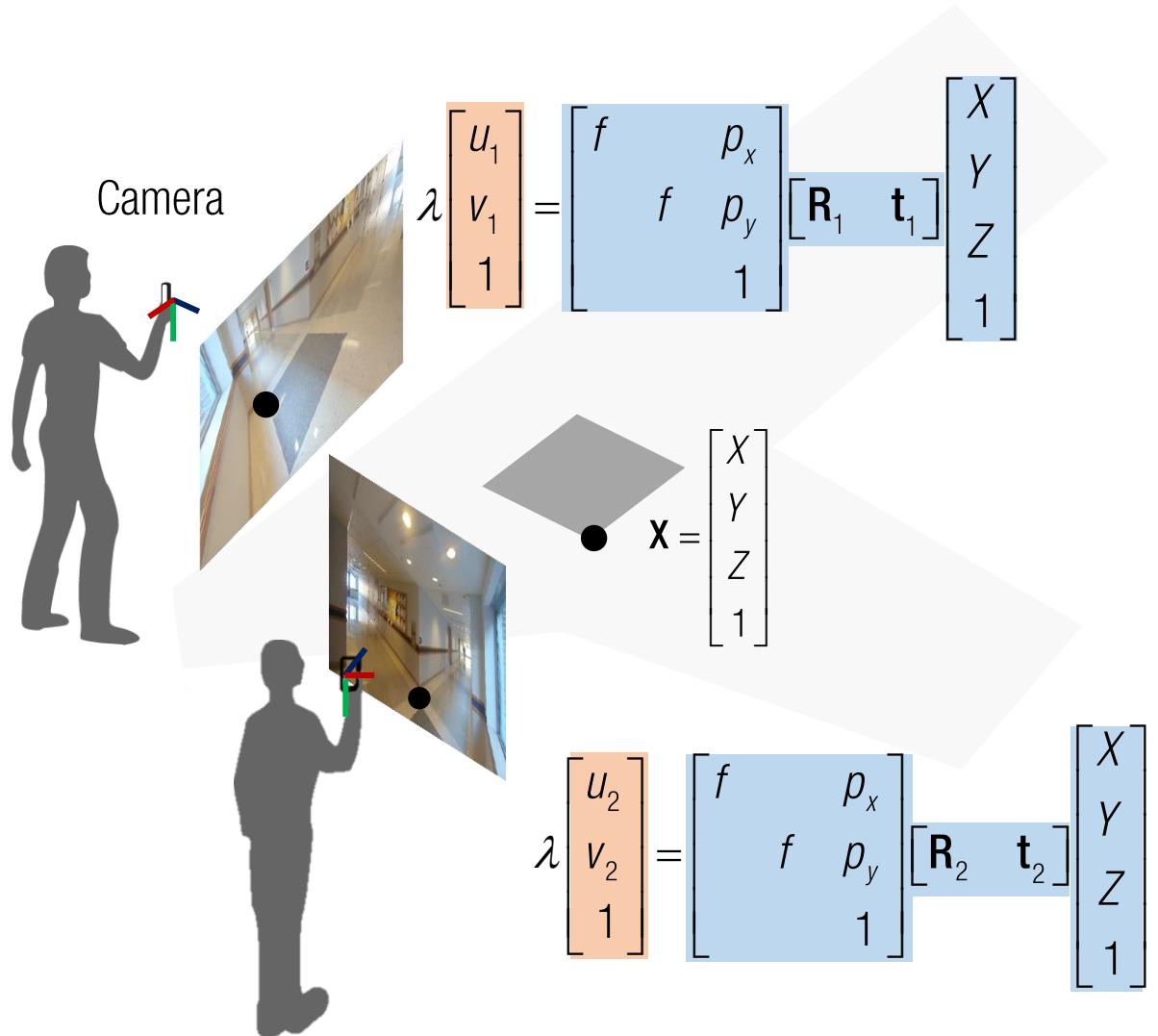
of unknowns:

n : the number of images

p : the number of points

of equations:

Multiview Camera Calibration



of unknowns: $3(\mathbf{K}) + 6n(\mathbf{R} \text{ and } \mathbf{t}) + 3p(\mathbf{X})$

n : the number of images

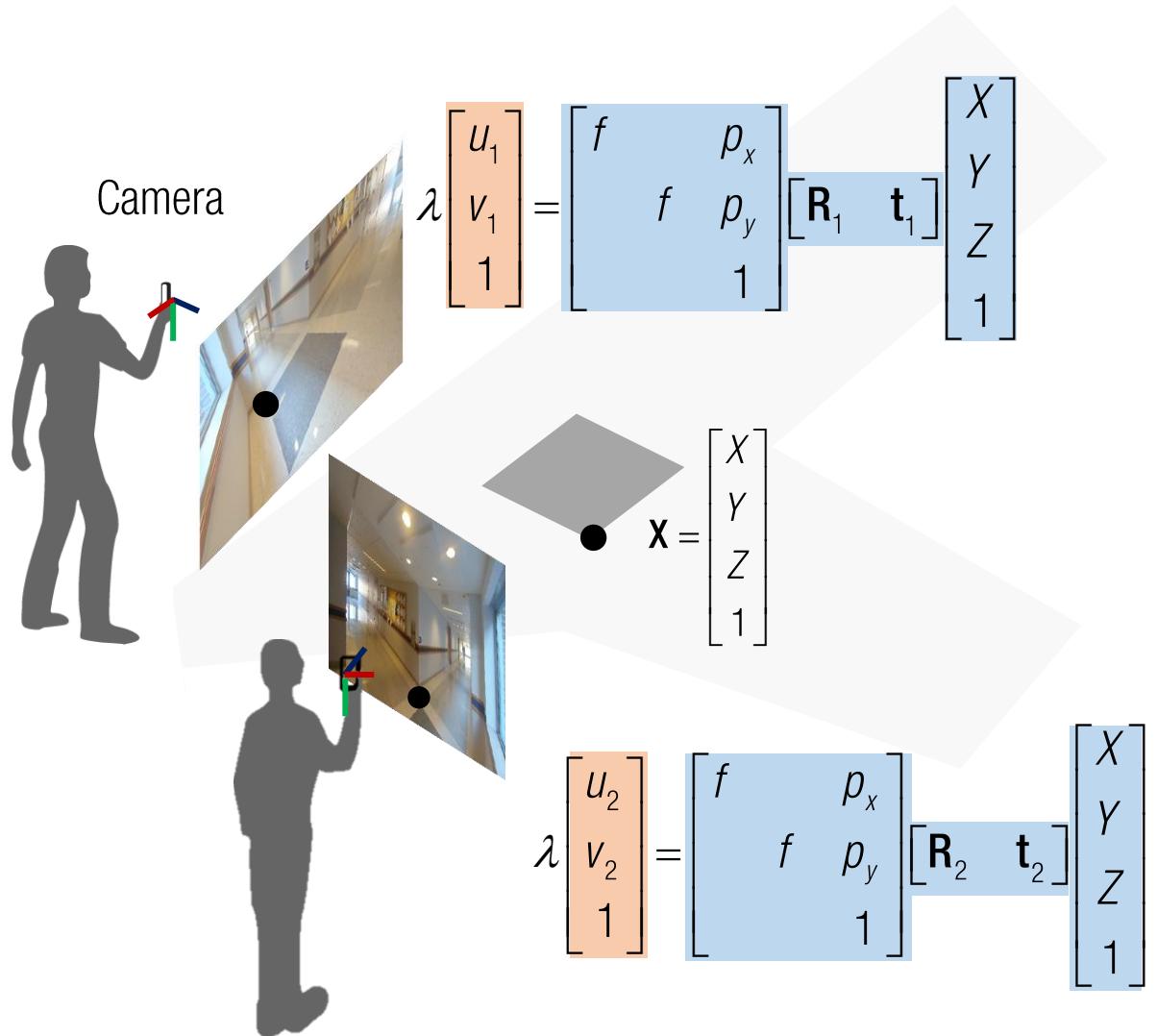
p : the number of points

of equations:

Knowns

Unknowns

Multiview Camera Calibration



of unknowns: $3(\mathbf{K}) + 6n(\mathbf{R} \text{ and } \mathbf{t}) + 3p(\mathbf{X})$

n : the number of images

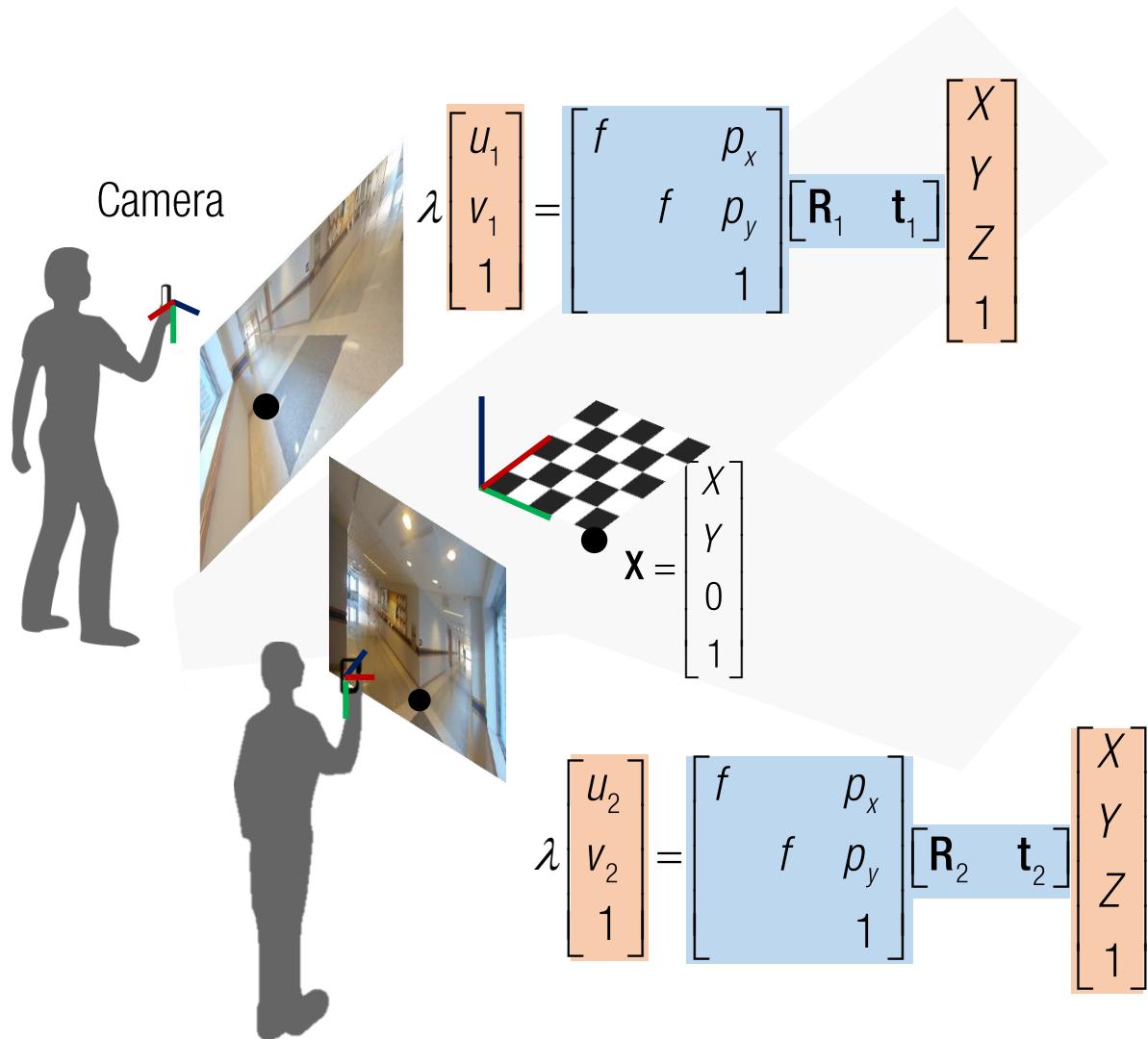
p : the number of points

of equations: $2np$

: Knowns

: Unknowns

Insight: Known Common 3D Points



of unknowns: 3 (**K**) + 6n (**R** and **t**)

n: the number of images

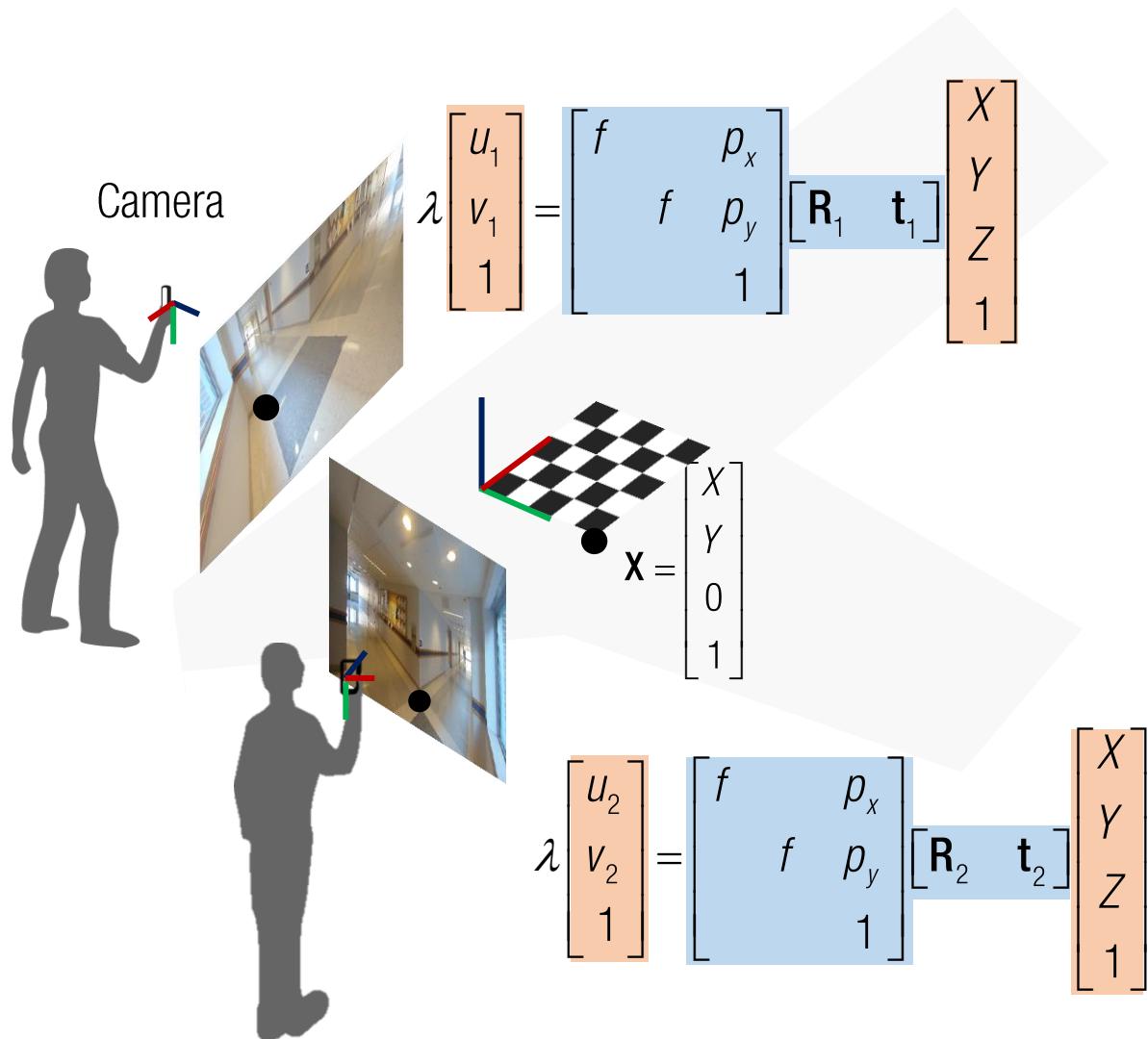
p: the number of points

of equations: $2np$

Knowns

Unknowns

Insight: Known Common 3D Points



of unknowns: 3 (**K**) + 6n (**R** and **t**)

n: the number of images

p: the number of points

of equations: 2np

We can solve for **K**, **R**, **t** if $3 + 6n < 2 np$

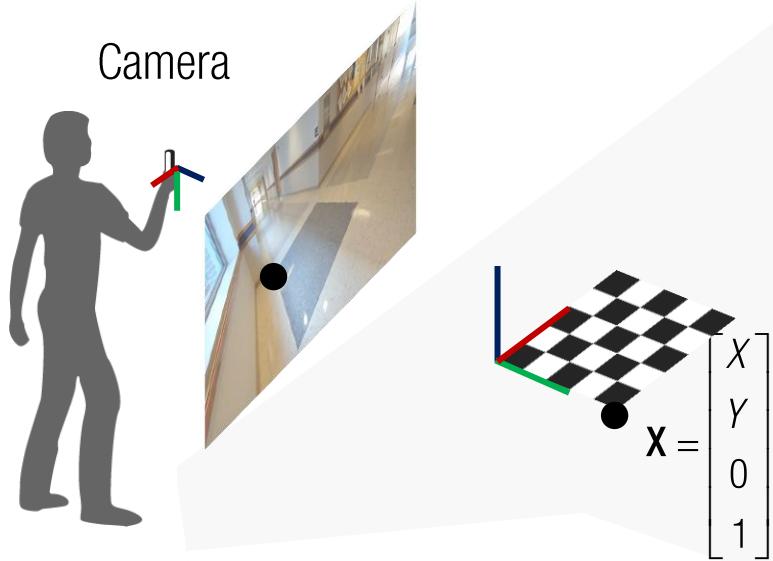
: Knowns

: Unknowns

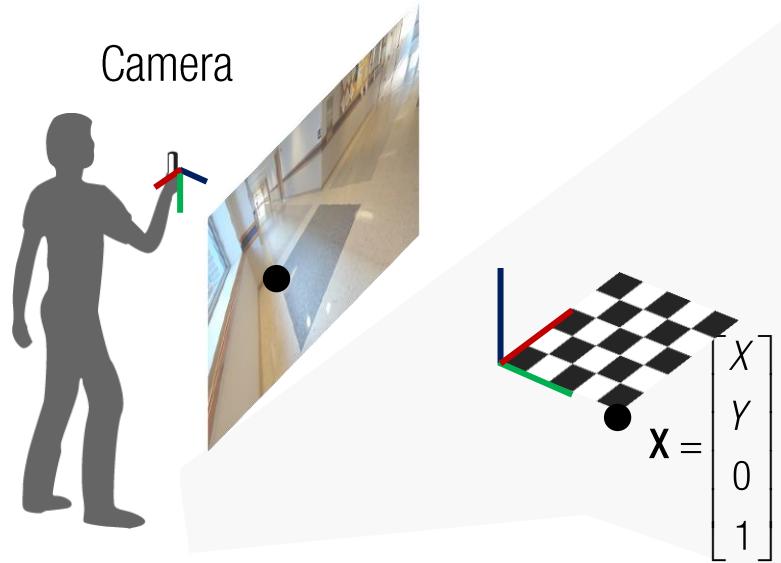
Homography

Points in 2D plane are mapped to an image with homography:

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} r_1 & r_2 & r_3 & t \end{bmatrix} \begin{bmatrix} X \\ Y \\ 0 \\ 1 \end{bmatrix}$$



Homography



Points in 2D plane are mapped to an image with homography:

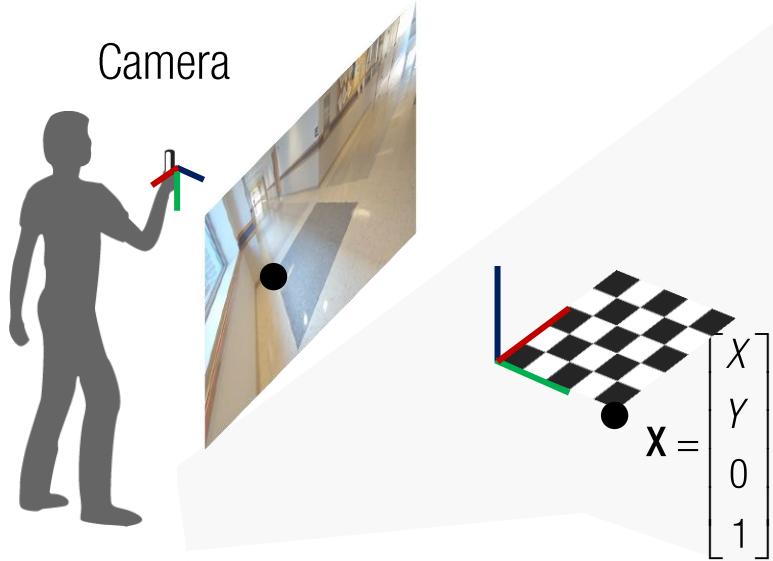
$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \begin{bmatrix} r_1 & r_2 & r_3 & t \end{bmatrix} \begin{bmatrix} X \\ Y \\ 0 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \begin{bmatrix} r_1 & r_2 & t \end{bmatrix} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$

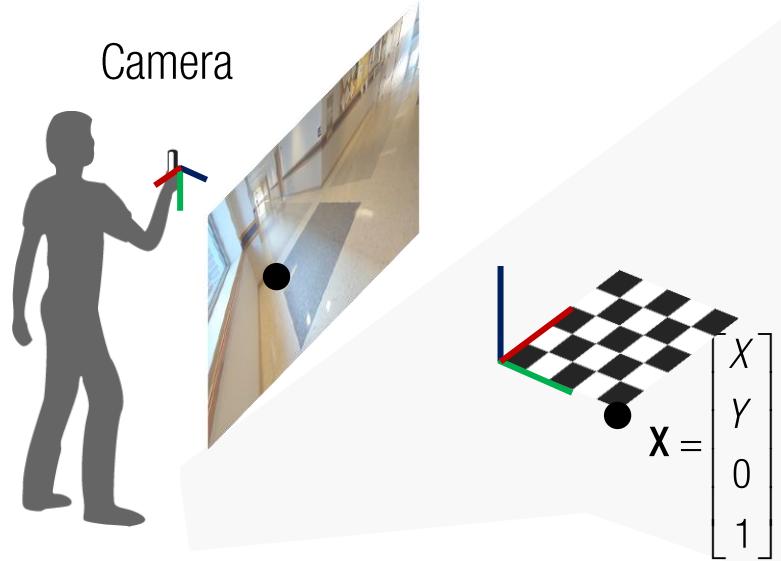
Homography

Points in 2D plane are mapped to an image with homography:

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{r}_3 & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ 0 \\ 1 \end{bmatrix}$$
$$= \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix} = \mathbf{H} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$



Homography



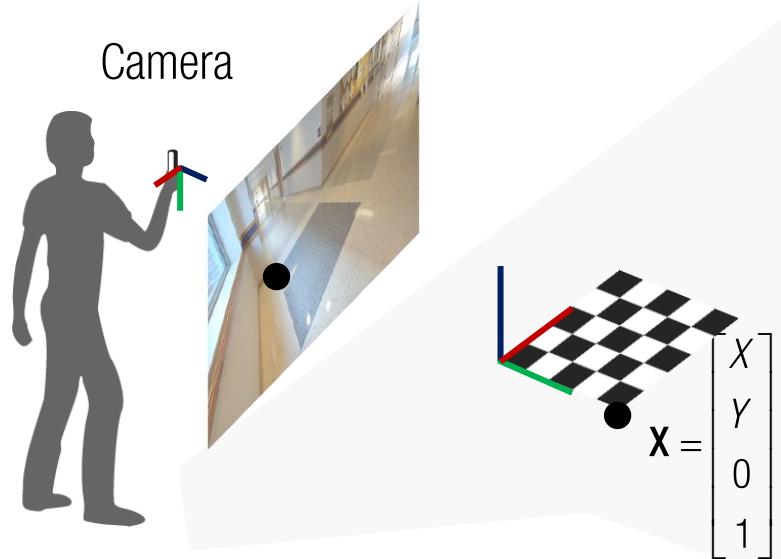
Points in 2D plane are mapped to an image with homography:

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} r_1 & r_2 & r_3 & t \end{bmatrix} \begin{bmatrix} X \\ Y \\ 0 \\ 1 \end{bmatrix}$$
$$= \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} r_1 & r_2 & t \end{bmatrix} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix} = \mathbf{H} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$

1. Compute homography

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \mathbf{H} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$

Homography



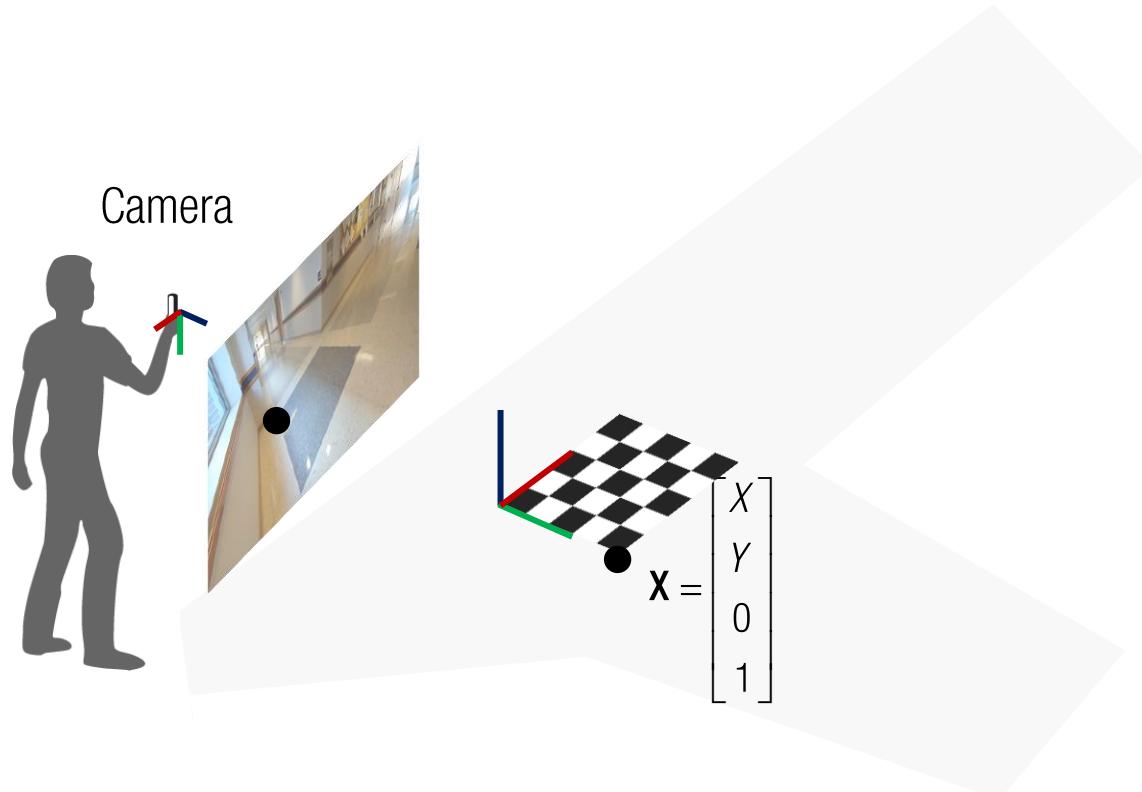
Points in 2D plane are mapped to an image with homography:

$$\lambda \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{r}_3 & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ 0 \\ 1 \end{bmatrix}$$

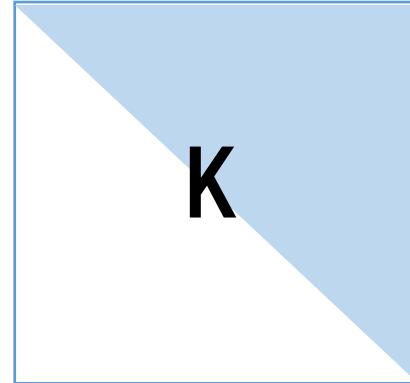
$$= \begin{bmatrix} f & p_x \\ f & p_y \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{t} \end{bmatrix} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix} = \mathbf{H} \begin{bmatrix} X \\ Y \\ 1 \end{bmatrix}$$

$$\mathbf{H} = \begin{array}{c|c} \mathbf{K} & \mathbf{Q} \end{array}$$

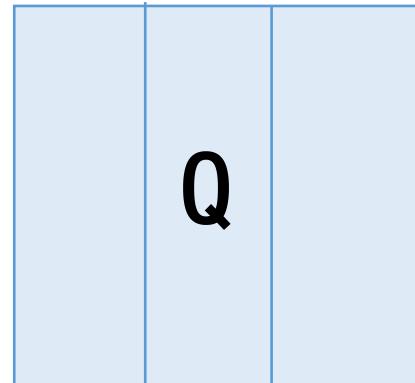
Method1: RQ Decomposition



$H =$

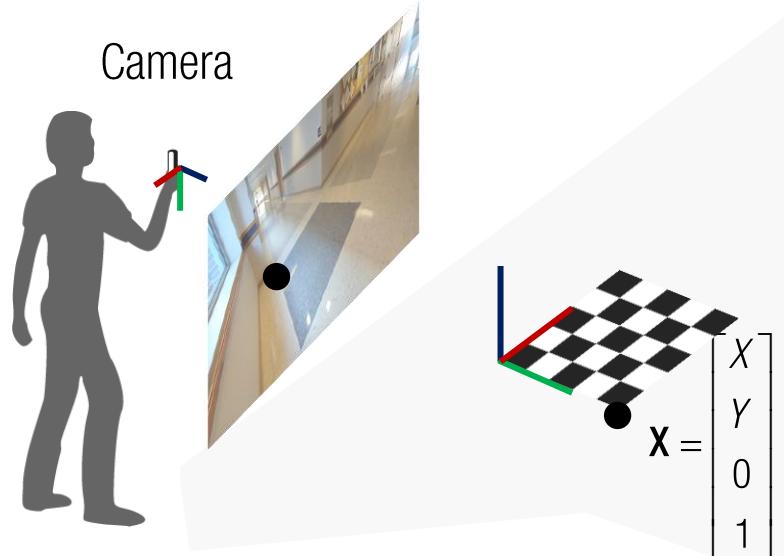


Upper triangle matrix

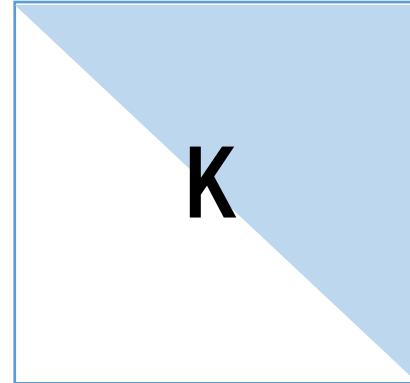


Orthogonal matrix

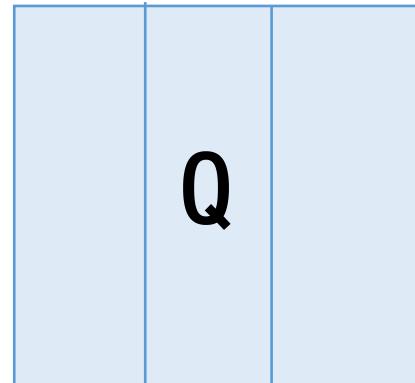
Method1: RQ Decomposition



$H =$



Upper triangle matrix



Orthogonal matrix

QR decomposition:

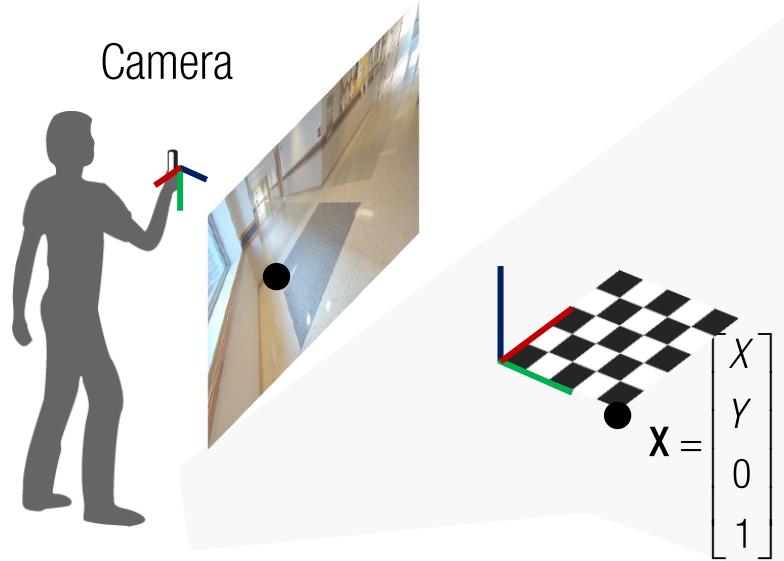
A

$=$

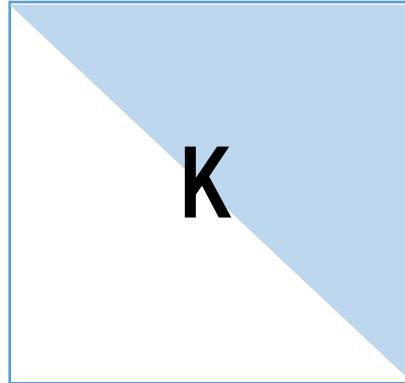
Q

R

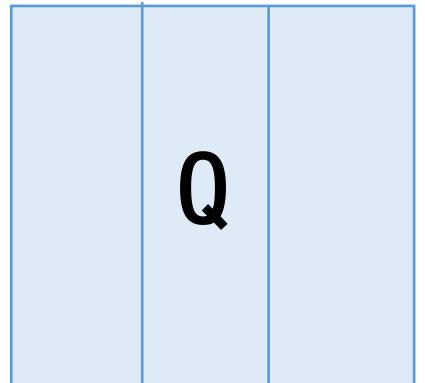
Method1: RQ Decomposition



$H =$

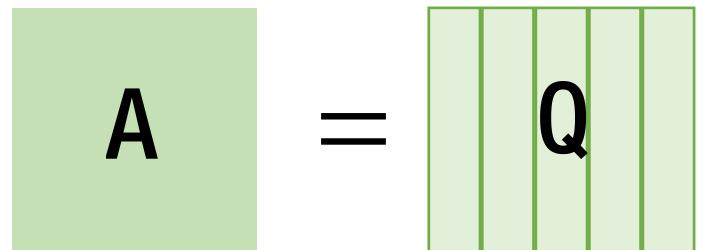


Upper triangle matrix



Orthogonal matrix

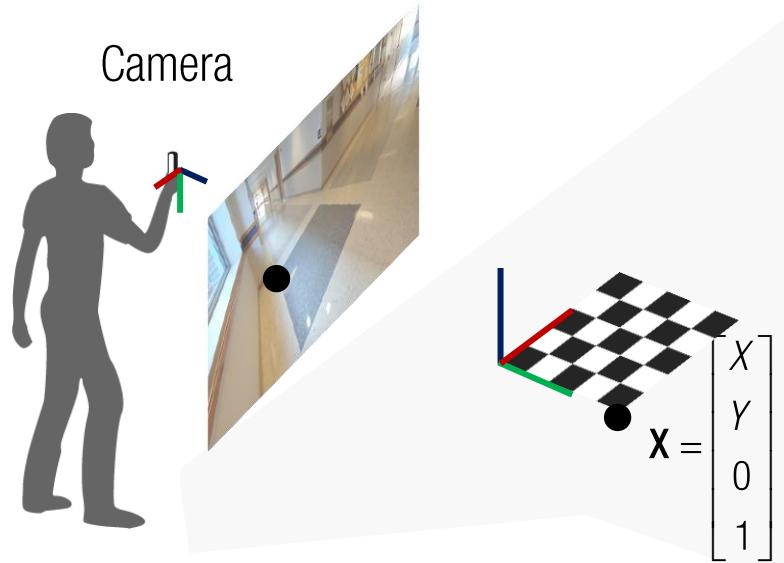
QR decomposition:



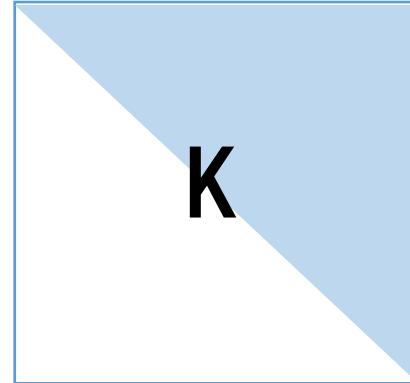
MATLAB

`[Q R] = qr(A)`

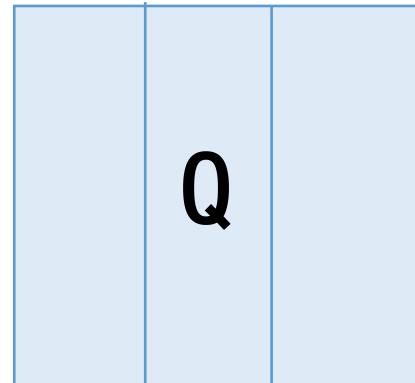
Method1: RQ Decomposition



$H =$

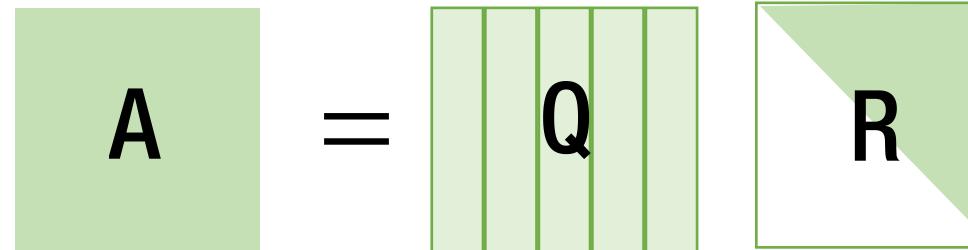


Upper triangle matrix



Orthogonal matrix

QR decomposition:

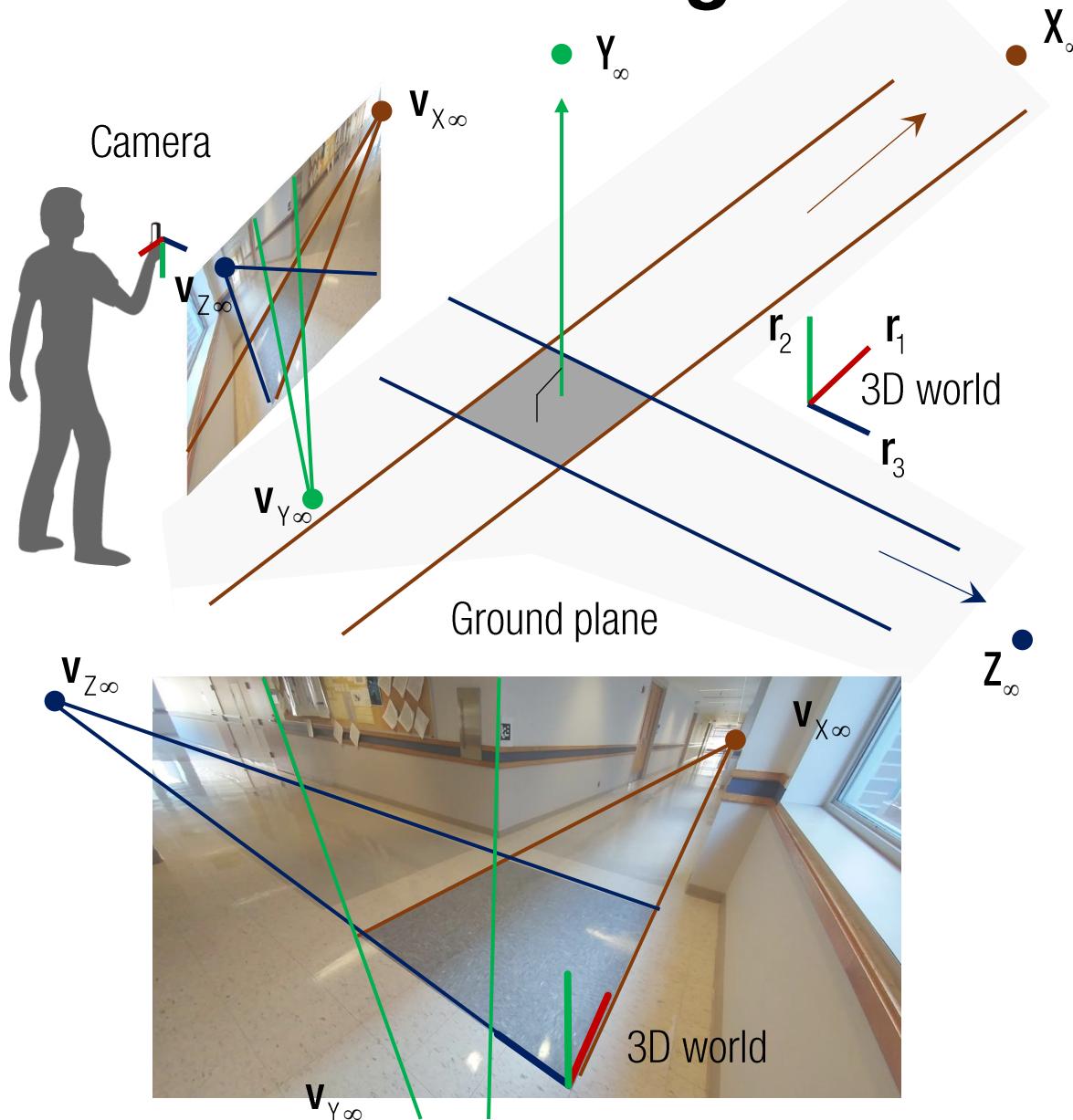


MATLAB

`[Q R] = qr(A)`

HW: How to convert **QR** to **RQ**?

Recall: Vanishing Points



$$\lambda v_{X\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R X_\infty \quad \lambda v_{Z\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R Z_\infty \quad \lambda v_{Y\infty} = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} R Y_\infty$$

$$\lambda K^{-1} v_{X\infty} = R X_\infty \quad \lambda K^{-1} v_{Y\infty} = R Y_\infty \quad \lambda K^{-1} v_{Z\infty} = R Z_\infty$$

Note that the camera extrinsic is still unknown (R and t).

Known property of points at infinity:

$$(X_\infty)^\top (Y_\infty) = 0 \qquad (RX_\infty)^\top (RY_\infty) = 0$$

$$(Y_\infty)^\top (Z_\infty) = 0 \quad \longleftrightarrow \quad (RY_\infty)^\top (RZ_\infty) = 0$$

$$(Z_\infty)^\top (X_\infty) = 0 \qquad (RZ_\infty)^\top (RX_\infty) = 0$$

$$(K^{-1}v_{X\infty})^\top (K^{-1}v_{Y\infty}) = (K^{-1}v_{Y\infty})^\top (K^{-1}v_{Z\infty}) = (K^{-1}v_{Z\infty})^\top (K^{-1}v_{X\infty}) = 0$$

: 3 unknowns and 3 equations

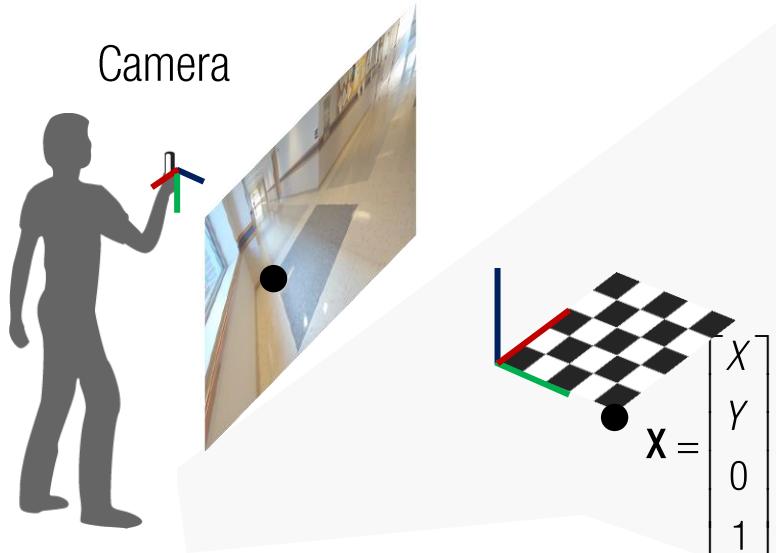
Method2: Rotation

: Knowns

: Unknowns

Homography factorization:

$$[\mathbf{h}_1 \ \mathbf{h}_2 \ \mathbf{h}_3] = \begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} [\mathbf{r}_1 \ \mathbf{r}_2 \ \mathbf{t}]$$

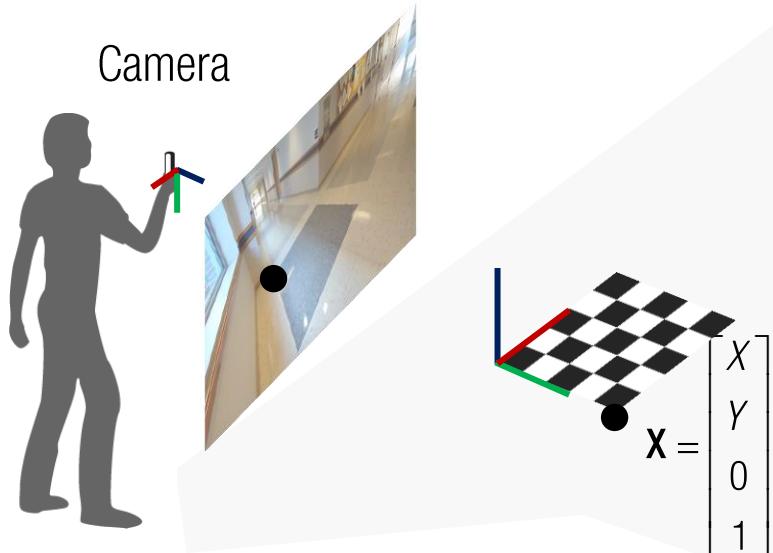


Method2: Rotation

: Knowns
: Unknowns

Homography factorization:

$$\left(\begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} \mathbf{h}_1 & \mathbf{h}_2 & \mathbf{h}_3 \end{bmatrix} = \begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{t} \end{bmatrix}$$



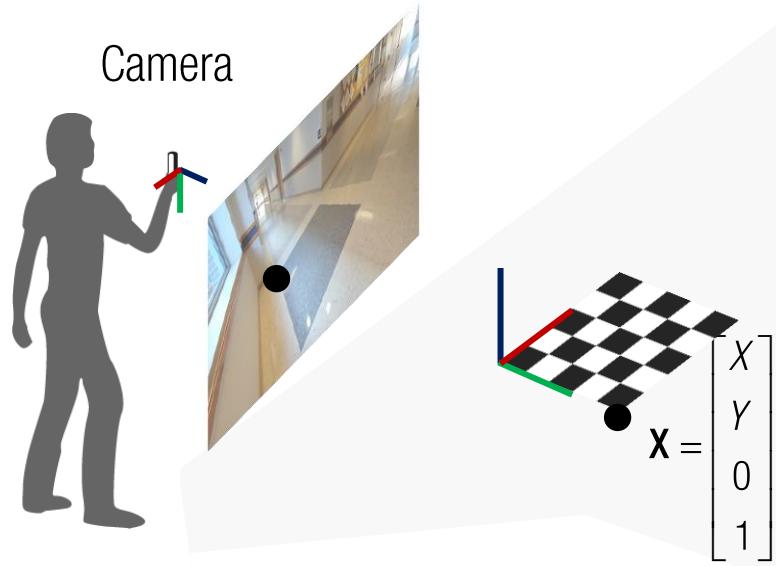
Method2: Rotation

: Knowns

: Unknowns

Homography factorization:

$$\left(\begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} \mathbf{h}_1 & \mathbf{h}_2 & \mathbf{h}_3 \end{bmatrix} = \begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{t} \end{bmatrix}$$



$$\mathbf{r}_1 = \mathbf{K}^{-1} \mathbf{h}_1 \quad \mathbf{r}_2 = \mathbf{K}^{-1} \mathbf{h}_2 \quad \mathbf{r}_3 = \mathbf{K}^{-1} \mathbf{h}_3$$

Method2: Rotation

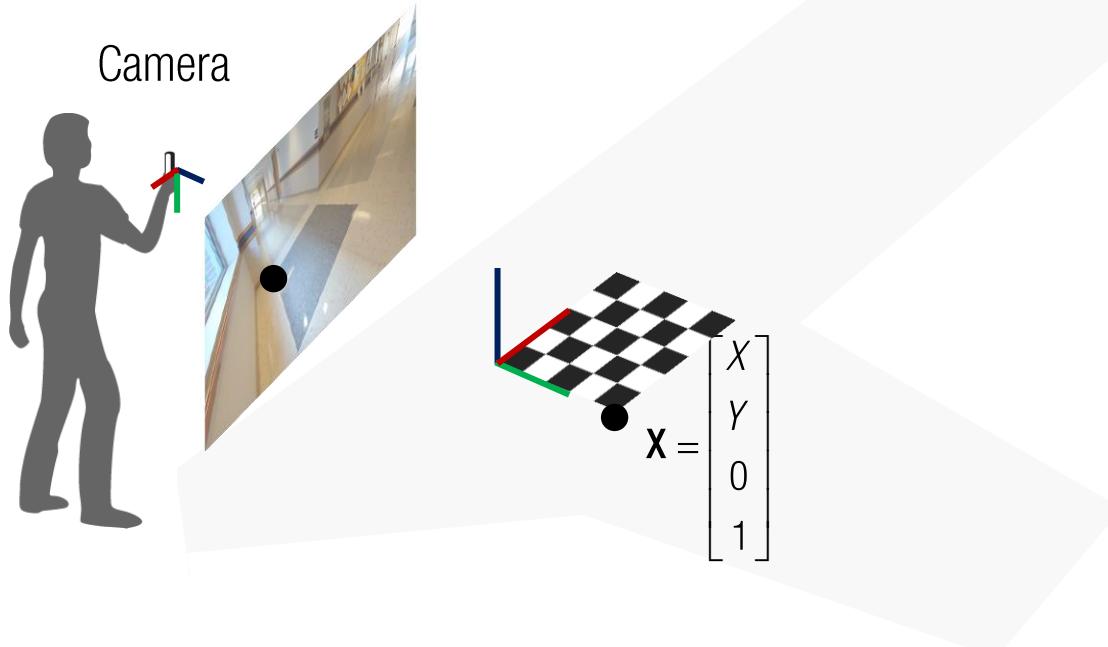
$$\mathbf{r}_1 = \mathbf{K}^{-1} \mathbf{h}_1$$

$$\mathbf{r}_2 = \mathbf{K}^{-1} \mathbf{h}_2$$

$$\mathbf{r}_3 = \mathbf{K}^{-1} \mathbf{h}_3$$

Orthogonality of rotation matrix property:

$$\mathbf{r}_1^T \mathbf{r}_2 = 0 \quad \|\mathbf{r}_1\| = 1 \quad \|\mathbf{r}_2\| = 1$$



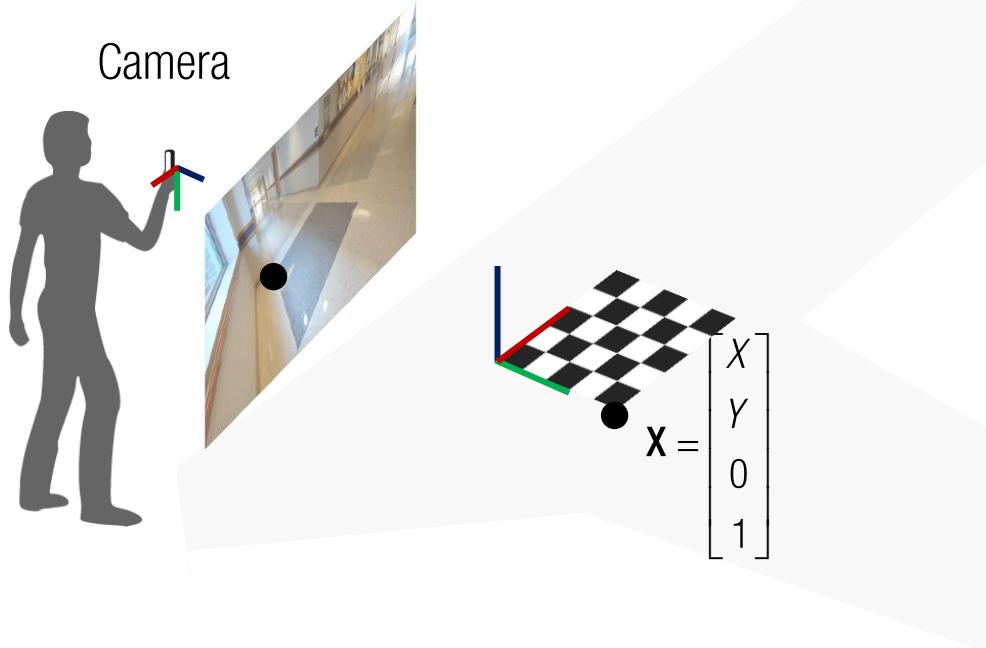
Method2: Rotation

$$\mathbf{r}_1 = \mathbf{K}^{-1} \mathbf{h}_1 \quad \mathbf{r}_2 = \mathbf{K}^{-1} \mathbf{h}_2 \quad \mathbf{r}_3 = \mathbf{K}^{-1} \mathbf{h}_3$$

Orthogonality of rotation matrix property:

$$\mathbf{r}_1^T \mathbf{r}_2 = 0 \quad \| \mathbf{r}_1 \| = 1 \quad \| \mathbf{r}_2 \| = 1$$

$$\rightarrow (\mathbf{K}^{-1} \mathbf{h}_1)^T (\mathbf{K}^{-1} \mathbf{h}_2) = \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2 = 0$$



Method2: Rotation

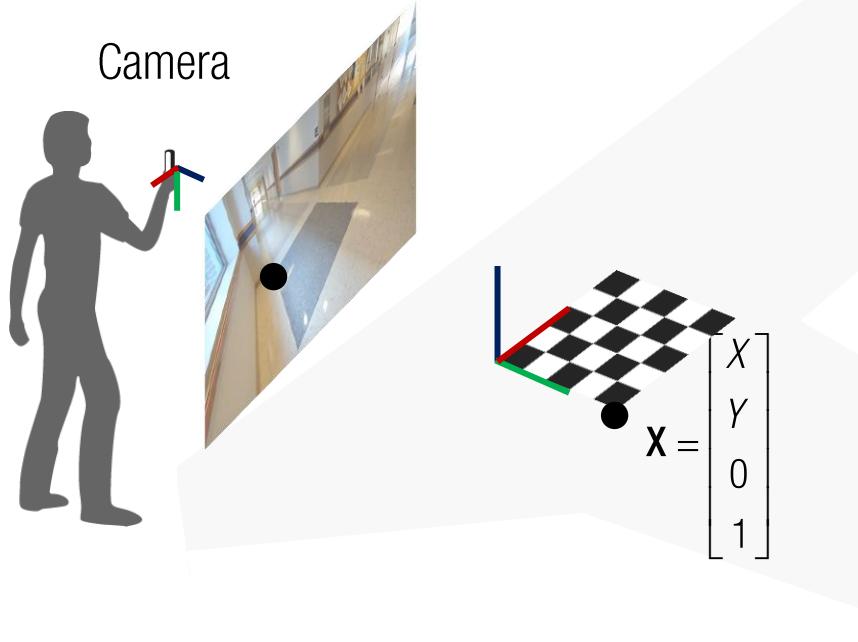
$$\mathbf{r}_1 = \mathbf{K}^{-1}\mathbf{h}_1 \quad \mathbf{r}_2 = \mathbf{K}^{-1}\mathbf{h}_2 \quad \mathbf{r}_3 = \mathbf{K}^{-1}\mathbf{h}_3$$

Orthogonality of rotation matrix property:

$$\mathbf{r}_1^T \mathbf{r}_2 = 0 \quad \|\mathbf{r}_1\| = 1 \quad \|\mathbf{r}_2\| = 1$$

$$\rightarrow (\mathbf{K}^{-1}\mathbf{h}_1)^T (\mathbf{K}^{-1}\mathbf{h}_2) = \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2 = 0$$

$$\|\mathbf{K}^{-1}\mathbf{h}_1\| = \|\mathbf{K}^{-1}\mathbf{h}_2\| \quad \text{or, } \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2$$



Method2: Rotation

$$\mathbf{r}_1 = \mathbf{K}^{-1} \mathbf{h}_1 \quad \mathbf{r}_2 = \mathbf{K}^{-1} \mathbf{h}_2 \quad \mathbf{r}_3 = \mathbf{K}^{-1} \mathbf{h}_3$$

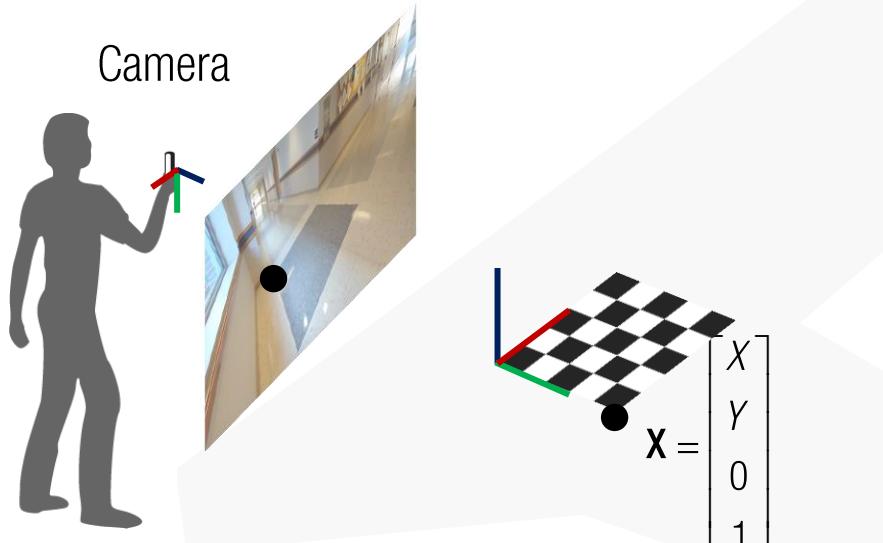
Orthogonality of rotation matrix property:

$$\mathbf{r}_1^T \mathbf{r}_2 = 0 \quad \|\mathbf{r}_1\| = 1 \quad \|\mathbf{r}_2\| = 1$$

$$\rightarrow (\mathbf{K}^{-1} \mathbf{h}_1)^T (\mathbf{K}^{-1} \mathbf{h}_2) = \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2 = 0$$

$$\|\mathbf{K}^{-1} \mathbf{h}_1\| = \|\mathbf{K}^{-1} \mathbf{h}_2\| \quad \text{or, } \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2$$

$$\mathbf{K}^{-T} \mathbf{K}^{-1} =$$



Method2: Rotation

$$\mathbf{r}_1 = \mathbf{K}^{-1} \mathbf{h}_1 \quad \mathbf{r}_2 = \mathbf{K}^{-1} \mathbf{h}_2 \quad \mathbf{r}_3 = \mathbf{K}^{-1} \mathbf{h}_3$$

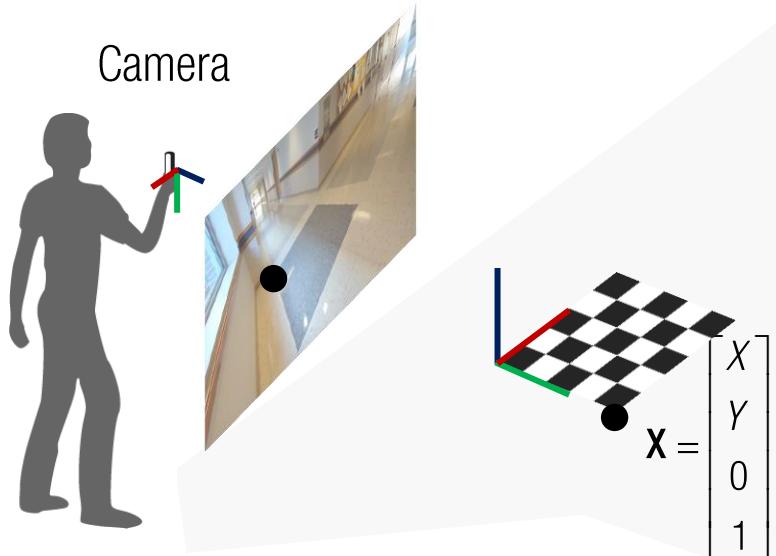
Orthogonality of rotation matrix property:

$$\mathbf{r}_1^T \mathbf{r}_2 = 0 \quad \|\mathbf{r}_1\| = 1 \quad \|\mathbf{r}_2\| = 1$$

$$\rightarrow (\mathbf{K}^{-1} \mathbf{h}_1)^T (\mathbf{K}^{-1} \mathbf{h}_2) = \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2 = 0$$

$$\|\mathbf{K}^{-1} \mathbf{h}_1\| = \|\mathbf{K}^{-1} \mathbf{h}_2\| \quad \text{or, } \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2$$

$$\mathbf{K}^{-T} \mathbf{K}^{-1} = \begin{bmatrix} 1/f & & & 1/f & -p_x/f \\ & 1/f & & 1/f & -p_y/f \\ & & 1 & & 1 \end{bmatrix}$$



Method2: Rotation

$$\mathbf{r}_1 = \mathbf{K}^{-1} \mathbf{h}_1 \quad \mathbf{r}_2 = \mathbf{K}^{-1} \mathbf{h}_2 \quad \mathbf{r}_3 = \mathbf{K}^{-1} \mathbf{h}_3$$

Orthogonality of rotation matrix property:

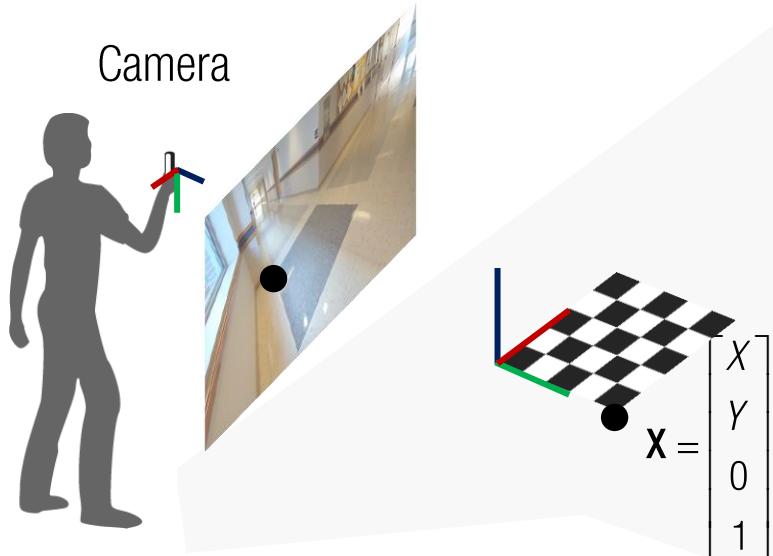
$$\mathbf{r}_1^T \mathbf{r}_2 = 0 \quad \|\mathbf{r}_1\| = 1 \quad \|\mathbf{r}_2\| = 1$$

$$\rightarrow (\mathbf{K}^{-1} \mathbf{h}_1)^T (\mathbf{K}^{-1} \mathbf{h}_2) = \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2 = 0$$

$$\|\mathbf{K}^{-1} \mathbf{h}_1\| = \|\mathbf{K}^{-1} \mathbf{h}_2\| \quad \text{or, } \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2$$

$$\mathbf{K}^{-T} \mathbf{K}^{-1} = \begin{bmatrix} 1/f & & \\ & 1/f & \\ -p_x/f & -p_y/f & 1 \end{bmatrix} \begin{bmatrix} 1/f & -p_x/f \\ 1/f & -p_y/f \\ 1 \end{bmatrix} = \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix}$$

$$\text{where } b_1 = \frac{1}{f^2}, \quad b_2 = -\frac{p_x}{f^2}, \quad b_3 = -\frac{p_y}{f^2}, \quad b_4 = \frac{p_x^2}{f^2} + \frac{p_y^2}{f^2} + 1$$



Method2: Rotation

$$\mathbf{r}_1 = \mathbf{K}^{-1} \mathbf{h}_1 \quad \mathbf{r}_2 = \mathbf{K}^{-1} \mathbf{h}_2 \quad \mathbf{r}_3 = \mathbf{K}^{-1} \mathbf{h}_3$$

Orthogonality of rotation matrix property:

$$\mathbf{r}_1^T \mathbf{r}_2 = 0 \quad \|\mathbf{r}_1\| = 1 \quad \|\mathbf{r}_2\| = 1$$

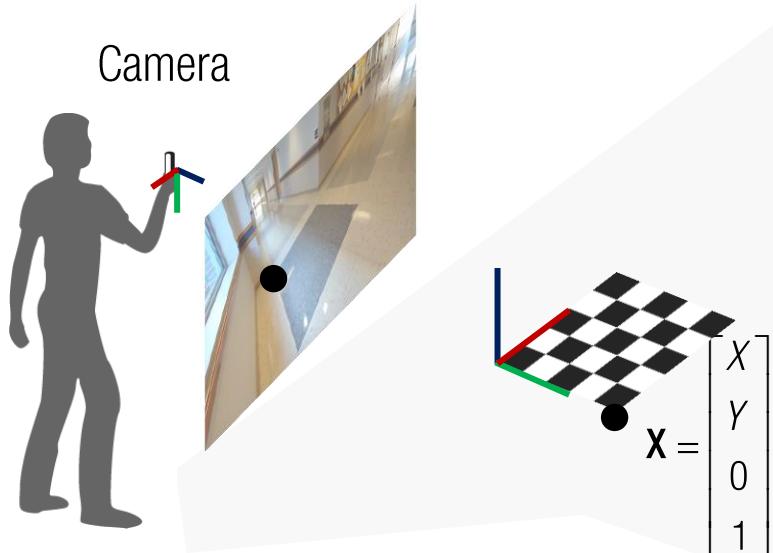
$$\rightarrow (\mathbf{K}^{-1} \mathbf{h}_1)^T (\mathbf{K}^{-1} \mathbf{h}_2) = \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2 = 0$$

$$\|\mathbf{K}^{-1} \mathbf{h}_1\| = \|\mathbf{K}^{-1} \mathbf{h}_2\| \quad \text{or, } \mathbf{h}_1^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{K}^{-T} \mathbf{K}^{-1} \mathbf{h}_2$$

$$\mathbf{K}^{-T} \mathbf{K}^{-1} = \begin{bmatrix} 1/f & & \\ & 1/f & \\ -p_x/f & -p_y/f & 1 \end{bmatrix} \begin{bmatrix} 1/f & -p_x/f \\ 1/f & -p_y/f \\ 1 \end{bmatrix} = \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \frac{\mathbf{B}}{\mathbf{B}}$$

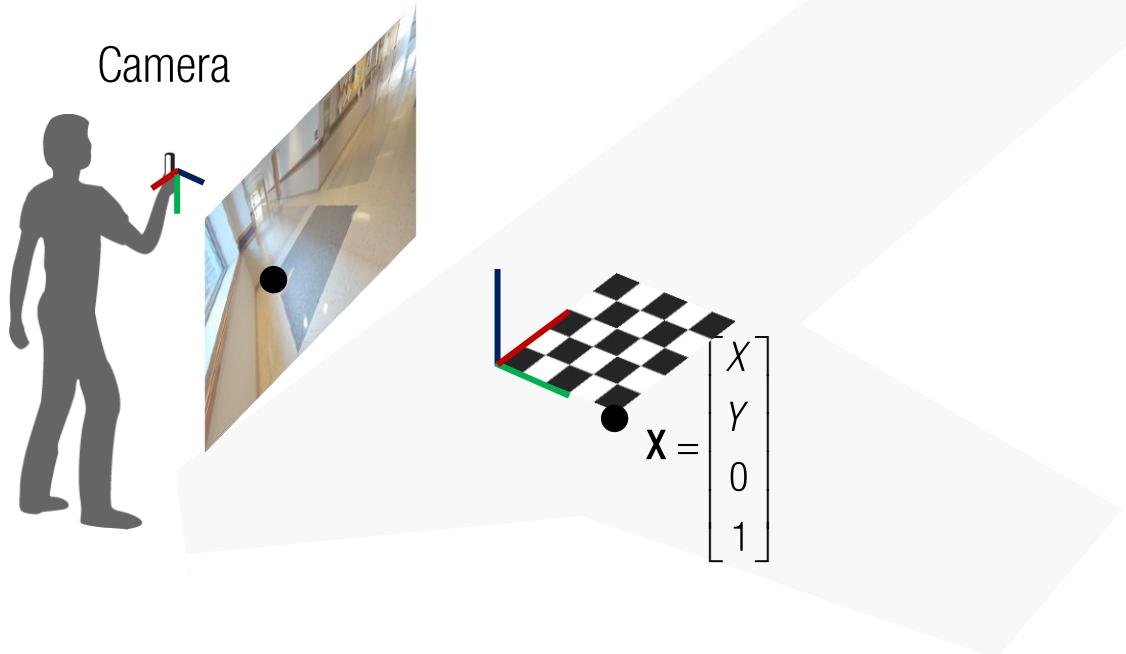
$$\text{where } b_1 = \frac{1}{f^2}, \quad b_2 = -\frac{p_x}{f^2}, \quad b_3 = -\frac{p_y}{f^2}, \quad b_4 = \frac{p_x^2}{f^2} + \frac{p_y^2}{f^2} + 1$$

$$\text{Linear in } \mathbf{B}: \quad \mathbf{h}_1^T \mathbf{B} \mathbf{h}_2 = 0 \quad \mathbf{h}_1^T \mathbf{B} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{B} \mathbf{h}_2$$

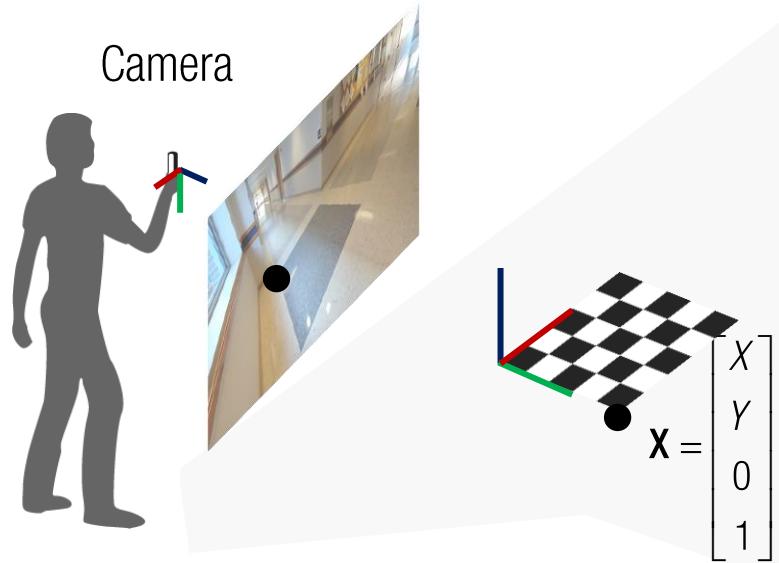


Method2: Rotation

$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_2 = \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix} = 0$$



Method2: Rotation

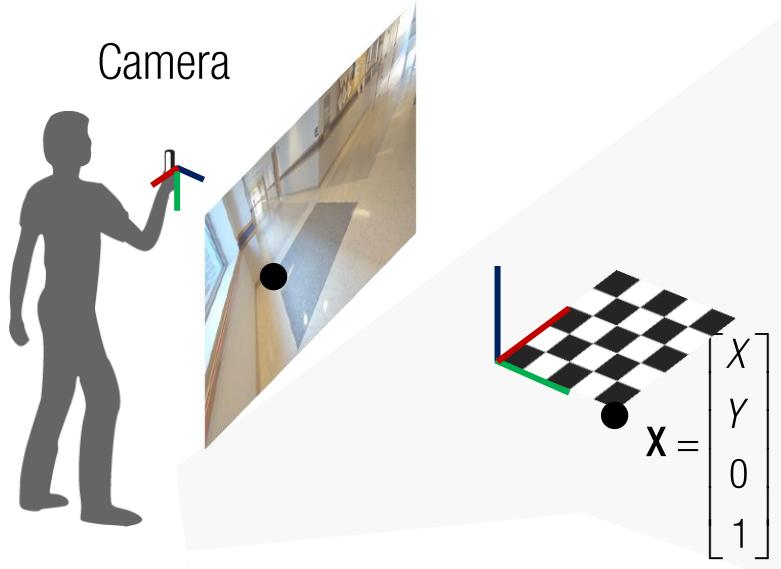


$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_2 = \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix} = 0$$

$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{B} \mathbf{h}_2$$

$$\begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix} = \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}$$

Method2: Rotation



$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_2 = \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix} = 0$$

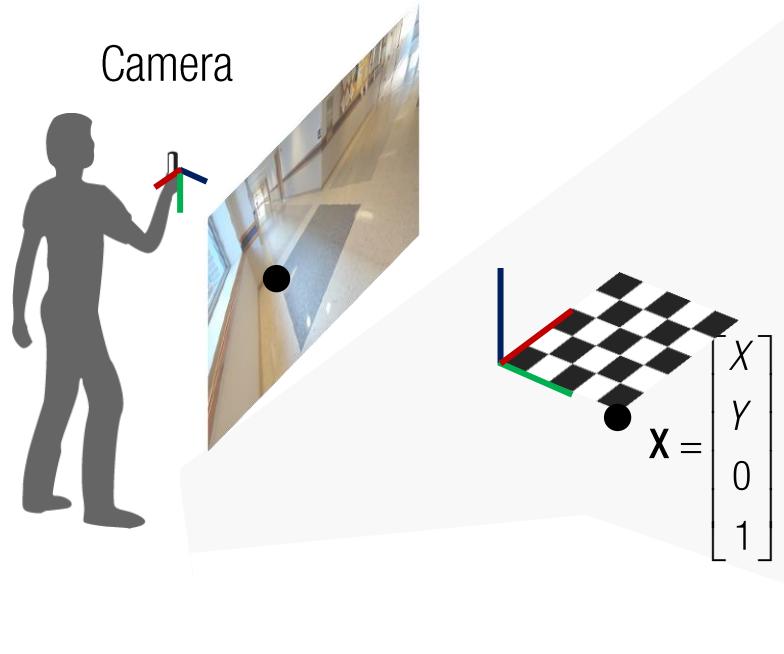
$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{B} \mathbf{h}_2$$

$$\begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix} = \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}$$

$$\begin{bmatrix} h_{11}h_{12} + h_{21}h_{22} & h_{11} + h_{12} & h_{21} + h_{22} & 1 \\ h_{11}^2 - h_{12}^2 + h_{21}^2 - h_{22}^2 & 2(h_{11} - h_{12}) & 2(h_{21} - h_{22}) & 0 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = \mathbf{0}$$

2x4

Method2: Rotation



$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_2 = \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 & b_3 \\ b_1 & b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix} = 0$$

$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{B} \mathbf{h}_2$$

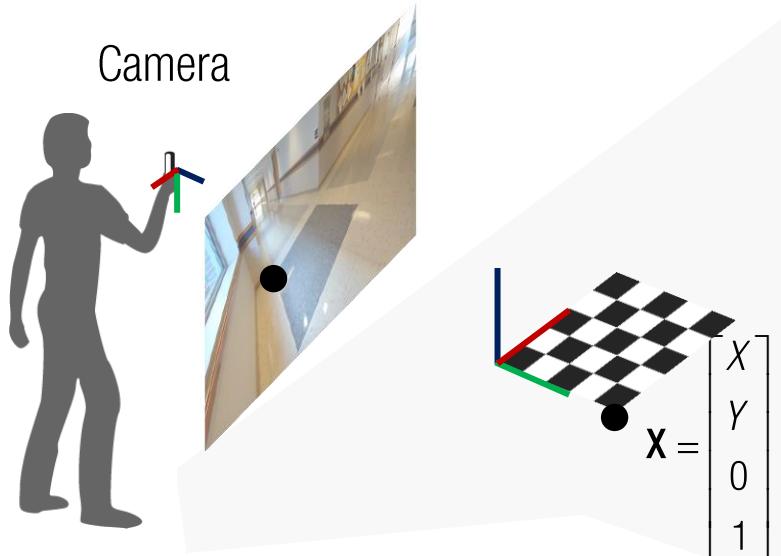
$$\begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 & b_3 \\ b_1 & b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix} = \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 & b_3 \\ b_1 & b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}$$

$$\begin{bmatrix} h_{11}h_{12} + h_{21}h_{22} & h_{11} + h_{12} & h_{21} + h_{22} & 1 \\ h_{11}^2 - h_{12}^2 + h_{21}^2 - h_{22}^2 & 2(h_{11} - h_{12}) & 2(h_{21} - h_{22}) & 0 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = \mathbf{0}$$

2x4

$$p_x = -\frac{b_2}{b_1}, \quad p_y = -\frac{b_3}{b_1}, \quad f = \sqrt{\frac{b_4}{b_1} - (p_x^2 + p_y^2)}$$

Method2: Rotation



$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_2 = \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 & b_3 \\ b_1 & b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix} = 0$$

$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{B} \mathbf{h}_2$$

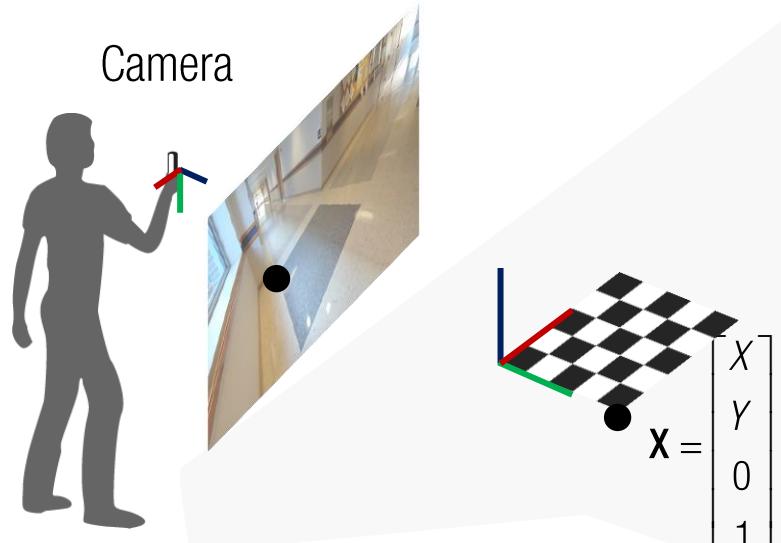
$$\begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 & b_3 \\ b_1 & b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix} = \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 & b_3 \\ b_1 & b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}$$

$$\begin{bmatrix} h_{11}h_{12} + h_{21}h_{22} & h_{11} + h_{12} & h_{21} + h_{22} & 1 \\ h_{11}^2 - h_{12}^2 + h_{21}^2 - h_{22}^2 & 2(h_{11}h_{12}) & 2(h_{21} - h_{22}) & 0 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = \mathbf{0}$$

2x4

Each image produces 2 equations and therefore, \mathbf{x} can be computed with minimum 2 images.

Method2: Rotation



$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_2 = \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix} = 0$$

$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{B} \mathbf{h}_2$$

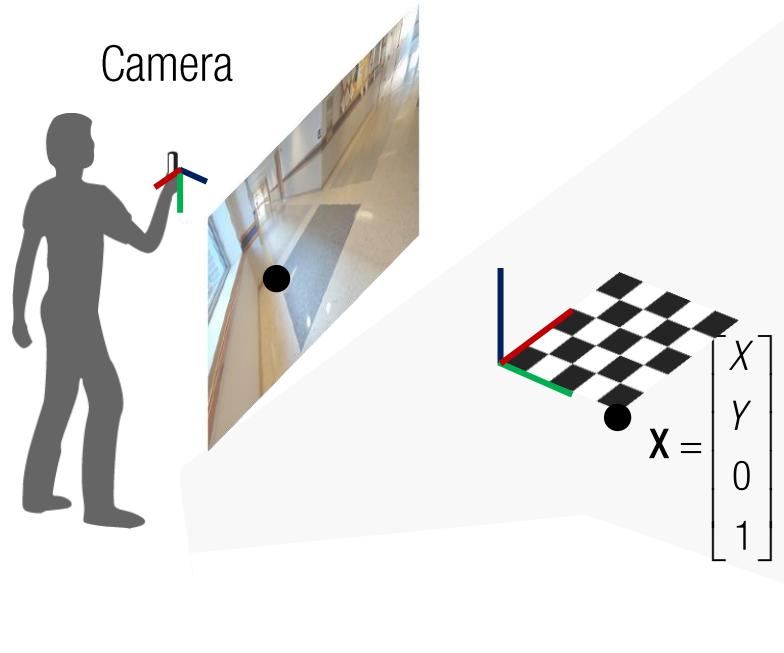
$$\begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix} = \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 \\ b_1 & b_3 \\ b_2 & b_3 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}$$

$$\begin{bmatrix} h_{11}h_{12} + h_{21}h_{22} & h_{11} + h_{12} & h_{21} + h_{22} & 1 \\ h_{11}^2 - h_{12}^2 + h_{21}^2 - h_{22}^2 & 2(h_{11}h_{12}) & 2(h_{21} - h_{22}) & 0 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = \mathbf{0}$$

2x4

$$p_x = -\frac{b_2}{b_1}, \quad p_y = -\frac{b_3}{b_1}, \quad f = \sqrt{\frac{b_4}{b_1} - (p_x^2 + p_y^2)}$$

Method2: Rotation



$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_2 = \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 & b_3 \\ b_1 & b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix} = 0$$

$$\mathbf{h}_1^T \mathbf{B} \mathbf{h}_1 = \mathbf{h}_2^T \mathbf{B} \mathbf{h}_2$$

$$\begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 & b_3 \\ b_1 & b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \begin{bmatrix} h_{11} \\ h_{21} \\ h_{31} \end{bmatrix} = \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}^T \begin{bmatrix} b_1 & b_2 & b_3 \\ b_1 & b_1 & b_3 \\ b_2 & b_3 & b_4 \end{bmatrix} \begin{bmatrix} h_{12} \\ h_{22} \\ h_{32} \end{bmatrix}$$

$$\begin{bmatrix} h_{11}h_{12} + h_{21}h_{22} & h_{11} + h_{12} & h_{21} + h_{22} & 1 \\ h_{11}^2 - h_{12}^2 + h_{21}^2 - h_{22}^2 & 2(h_{11}h_{12}) & 2(h_{21} - h_{22}) & 0 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} = \mathbf{0}$$

2x4

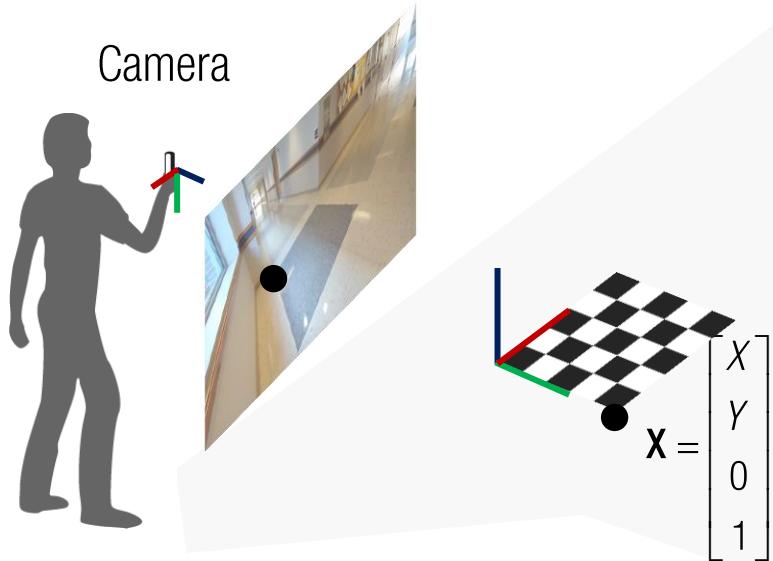
$$p_x = -\frac{b_2}{b_1}, \quad p_y = -\frac{b_3}{b_1}, \quad f = \sqrt{\frac{b_4}{b_1} - (p_x^2 + p_y^2)}$$

Method2: Rotation

: Knowns
: Unknowns

Homography factorization:

$$\left(\begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} \mathbf{h}_1 & \mathbf{h}_2 & \mathbf{h}_3 \end{bmatrix} = \begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{t} \end{bmatrix}$$



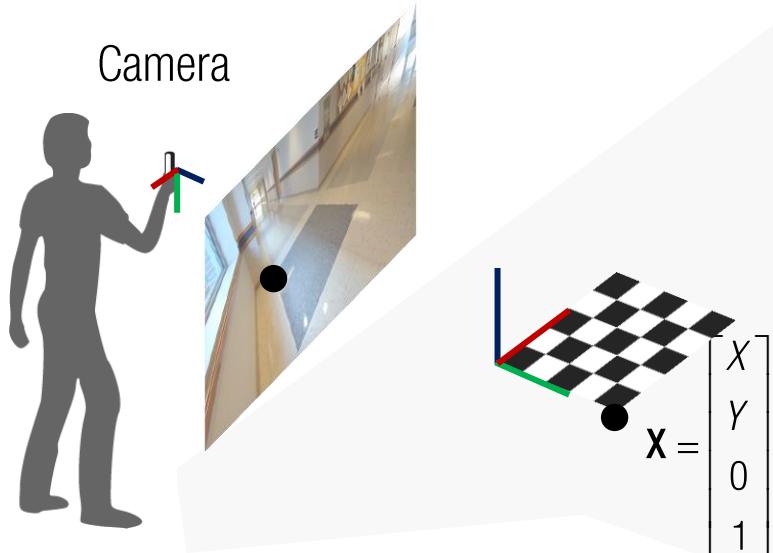
$$\mathbf{r}_1 = \mathbf{K}^{-1} \mathbf{h}_1 \quad \mathbf{r}_2 = \mathbf{K}^{-1} \mathbf{h}_2 \quad \mathbf{r}_3 = \mathbf{K}^{-1} \mathbf{h}_3$$

Method2: Rotation

: Knowns
: Unknowns

Homography factorization:

$$\left(\begin{bmatrix} f & p_x \\ f & p_y \\ 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} \mathbf{h}_1 & \mathbf{h}_2 & \mathbf{h}_3 \end{bmatrix} = \begin{bmatrix} \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{t} \end{bmatrix}$$



$$\mathbf{r}_1 = \mathbf{K}^{-1} \mathbf{h}_1 \quad \mathbf{r}_2 = \mathbf{K}^{-1} \mathbf{h}_2 \quad \mathbf{r}_3 = \mathbf{K}^{-1} \mathbf{h}_3$$

$$\underline{\mathbf{r}_1 = \frac{\mathbf{K}^{-1} \mathbf{h}_1}{\|\mathbf{K}^{-1} \mathbf{h}_1\|}}, \quad \underline{\mathbf{r}_2 = \frac{\mathbf{K}^{-1} \mathbf{h}_2}{\|\mathbf{K}^{-1} \mathbf{h}_2\|}}, \quad \underline{\mathbf{t} = \frac{\mathbf{K}^{-1} \mathbf{h}_3}{\|\mathbf{K}^{-1} \mathbf{h}_3\|}}, \quad \underline{\mathbf{r}_3 = \mathbf{r}_1 \times \mathbf{r}_2}$$

Divided by constant factor

MATLAB Calibration Toolbox Demo

https://www.vision.caltech.edu/bouguetj/calib_doc/

Current Folder

Name
image0000001.jpg
image0000002.jpg
image0000003.jpg
image0000004.jpg
image0000005.jpg
image0000006.jpg
image0000007.jpg
image0000008.jpg
image0000009.jpg
image0000010.jpg
image0000011.jpg
image0000012.jpg
image0000013.jpg
image0000014.jpg
image0000015.jpg
image0000016.jpg
image0000017.jpg
image0000018.jpg
image0000019.jpg
image0000020.jpg
image0000021.jpg
image0000022.jpg
image0000023.jpg
image0000024.jpg
image0000025.jpg
image0000026.jpg
image0000027.jpg
image0000028.jpg

Command Window

New to MATLAB? See resources for [Getting Started](#).

```
>> cd C:\Users\hspark\Desktop\Lecture\calib\image
>> addpath C:\Users\hspark\Desktop\Lecture\calib\toolbox_calib\TOOLBOX_calib
>> calib
fx >
```

Run calibration

Standard mode

Read image

The screenshot shows the MATLAB graphical user interface. On the left, the 'Current Folder' browser displays a list of 28 image files named 'image0000001.jpg' through 'image0000028.jpg'. In the center, the 'Command Window' contains MATLAB commands to change the current directory to 'C:\Users\hspark\Desktop\Lecture\calib\image', add the toolbox path, call the 'calib' function, and then run it again. A small 'fx' icon is next to the second 'run' command. On the right, a 'Camera Calibration Toolbox - Select m...' dialog box is open, showing three options: 'Standard (all the images are stored in memory)', 'Memory efficient (the images are loaded one by one)', and 'Exit'. An arrow points from the text 'Standard mode' to the 'Standard' option in the dialog, and another arrow points from the text 'Read image' to the 'Memory efficient' option.

Current Folder

Name
image0000001.jpg
image0000002.jpg
image0000003.jpg
image0000004.jpg
image0000005.jpg
image0000006.jpg
image0000007.jpg
image0000008.jpg
image0000009.jpg
image0000010.jpg
image0000011.jpg
image0000012.jpg
image0000013.jpg
image0000014.jpg
image0000015.jpg
image0000016.jpg
image0000017.jpg
image0000018.jpg
image0000019.jpg
image0000020.jpg
image0000021.jpg
image0000022.jpg
image0000023.jpg
image0000024.jpg
image0000025.jpg
image0000026.jpg
image0000027.jpg
image0000028.jpg

Command Window

New to MATLAB? See resources for [Getting Started](#).

```
>> cd C:\Users\hspark\Desktop\Lecture\calib\image
>> addpath C:\Users\hspark\Desktop\Lecture\calib\toolbox_calib\TOOLBOX_calib
>> calib

.
..
image000001.jpg image000008.jpg image000015.jpg image000022.jpg
image000002.jpg image000009.jpg image000016.jpg image000023.jpg
image000003.jpg image000010.jpg image000017.jpg image000024.jpg
image000004.jpg image000011.jpg image000018.jpg image000025.jpg
image000005.jpg image000012.jpg image000019.jpg image000026.jpg
```

Image prefix

Basename camera calibration images (without number nor suffix): >> image0000

Image format: ([]='r'='ras', 'b'='bmp', 't'='tif', 'p'='pgm', 'j'='jpg', 'm'='ppm') j

Loading image 1...2...3...4...5...6...7...8...9...10...11...12...13...14...15...16...17...18...19...2

done

Image extension

fxt >> |

Current Folder

Name

- image0000001.jpg
- image0000002.jpg
- image0000003.jpg
- image0000004.jpg
- image0000005.jpg
- image0000006.jpg
- image0000007.jpg
- image0000008.jpg
- image0000009.jpg
- image0000010.jpg
- image0000011.jpg
- image0000012.jpg
- image0000013.jpg
- image0000014.jpg
- image0000015.jpg
- image0000016.jpg
- image0000017.jpg
- image0000018.jpg
- image0000019.jpg
- image0000020.jpg
- image0000021.jpg
- image0000022.jpg
- image0000023.jpg
- image0000024.jpg
- image0000025.jpg
- image0000026.jpg
- image0000027.jpg
- image0000028.jpg

Command Window

New to MATLAB? See resources for [Getting Started](#).

```
>> cd C:\Users\hspark\Desktop\Lecture\calib\image
>> addpath C:\Users\hspark\Desktop\Lecture\calib\toolbox_calib\TOOLBOX_calib
>> calib
```

Camera Calibration Toolbox - Standard Version

Image names	Read images	Extract grid corners	Calibration
Show Extrinsic	Reproject on images	Analyse error	Recomp. corners
Add/Suppress images	Save	Load	Exit
Comp. Extrinsic	Undistort image	Export calib data	Show calib results

image000005.jpg image000012.jpg image000019.jpg image000026.jpg

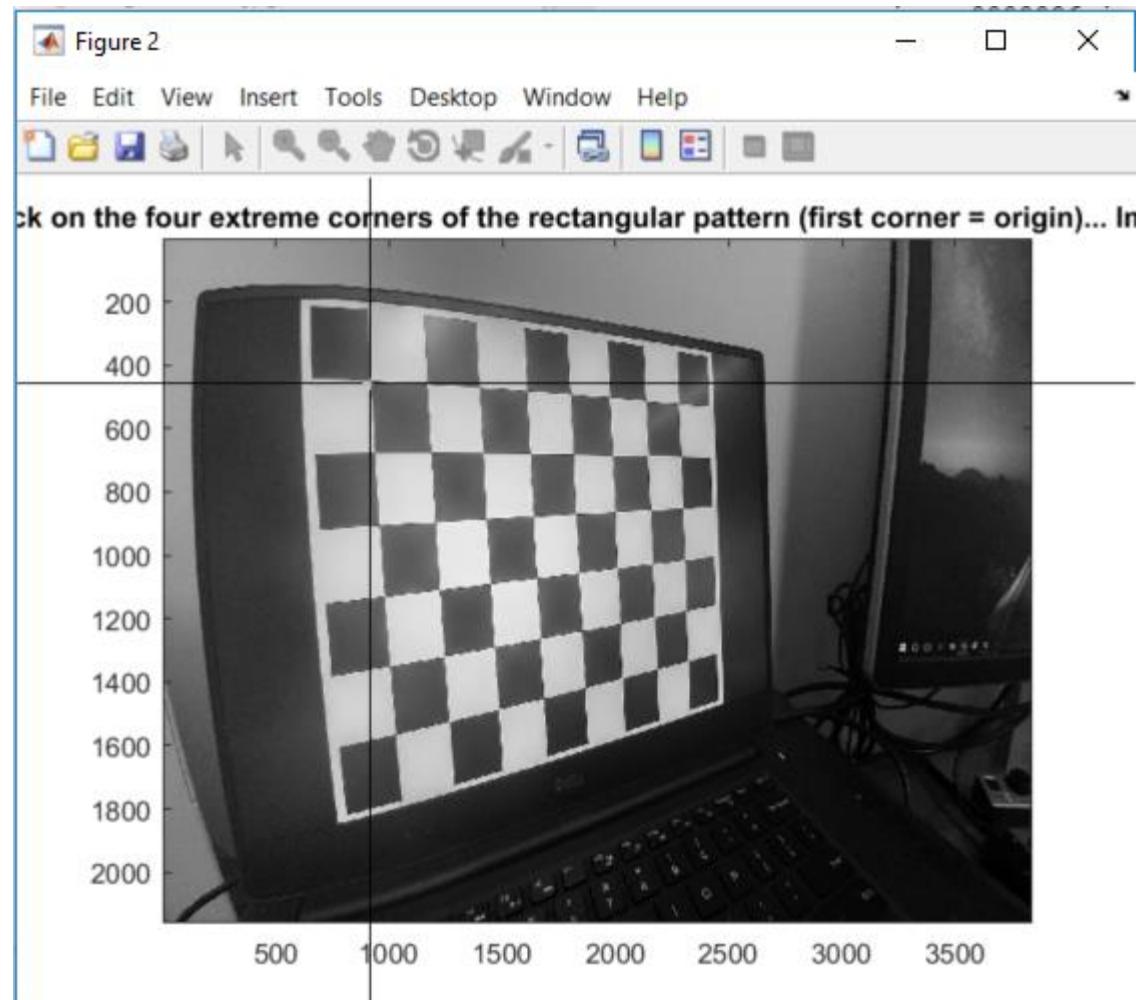
Basename camera calibration images (without number nor suffix): >> image00000

Image format: ([]='r'='ras', 'b'='bmp', 't'='tif', 'p'='pgm', 'j'='jpg', 'm'='ppm') j

Loading image 1...2...3...4...5...6...7...8...9...10...11...12...13...14...15...16...17...18...19...20

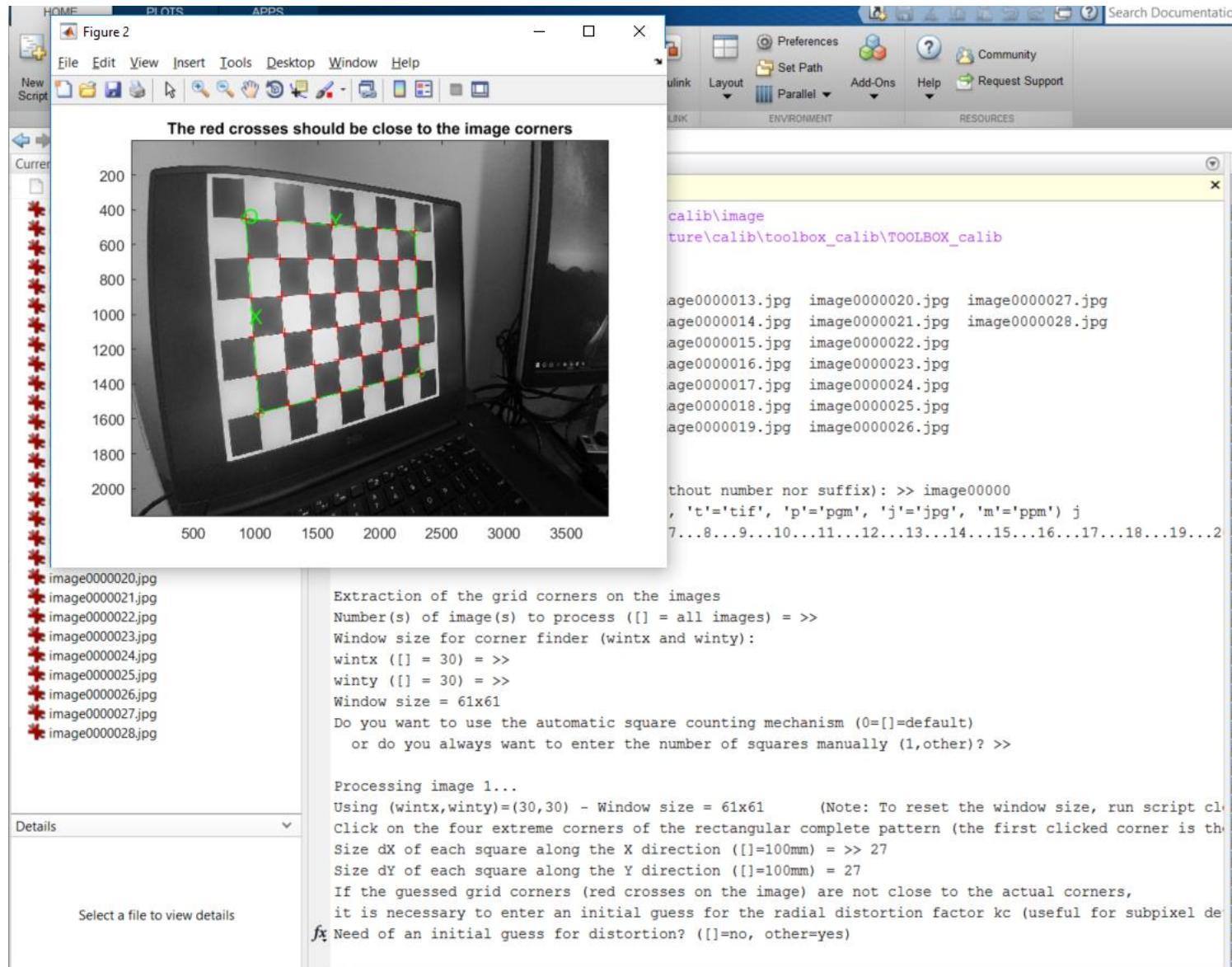
done

fx >>



Click four corner in the following order:

1. Top left
2. Top right
3. Bottom right
4. Bottom left

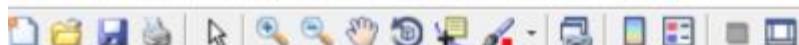


Default mode (press Enter)

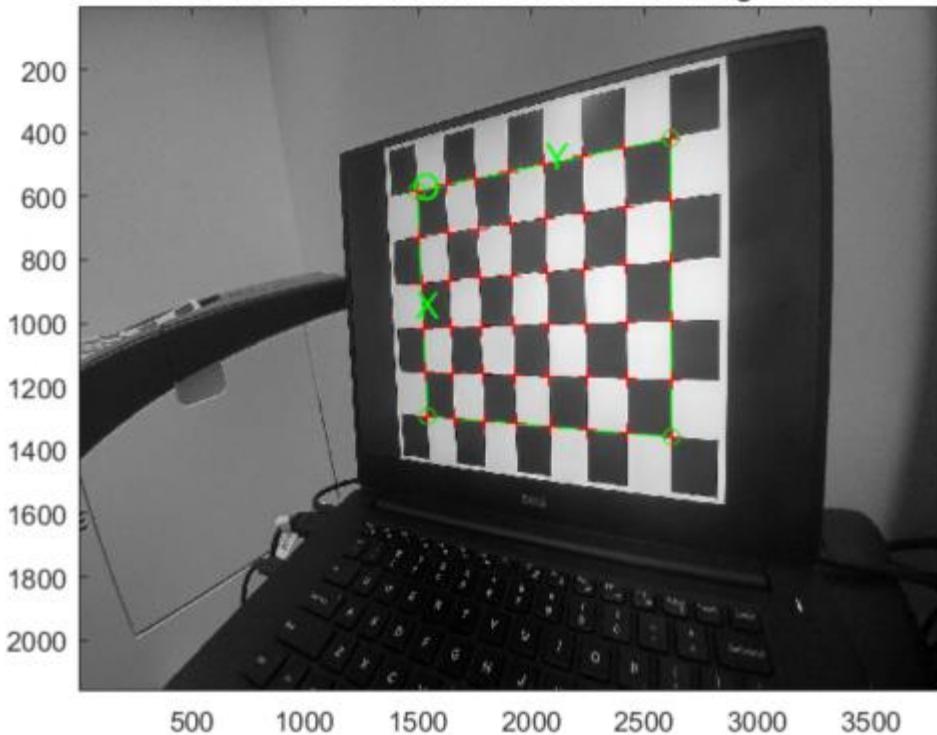
Set grid size (27mm)

Figure 2

File Edit View Insert Tools Desktop Window Help

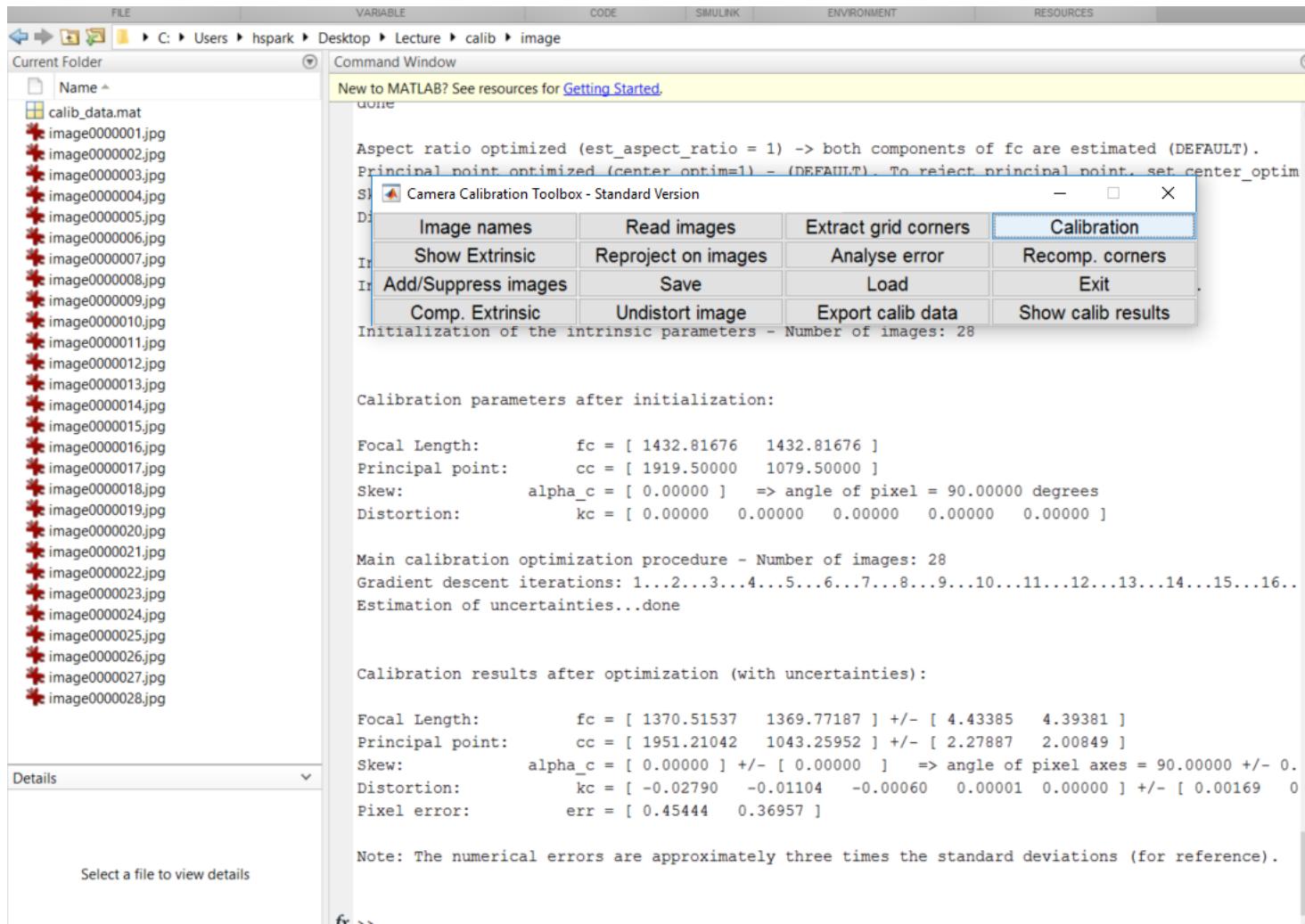


The red crosses should be close to the image corners



```
Need of an initial guess for distortion? ([]=no, other=yes)
Corner extraction...
 Camera Calibration Toolbox - Standard Version
P Image names Read images Extract grid corners Calibration
U Show Extrinsic Reproject on images Analyse error Recomp. corners
C Add/Suppress images Save Load Exit
S Comp. Extrinsic Undistort image Export calib data Show calib results
If the guessed grid corners (red crosses on the image) are not close to the actual corners,
it is necessary to enter an initial guess for the radial distortion factor kc (useful for subp
Need of an initial guess for distortion? ([]=no, other=yes)
Corner extraction...

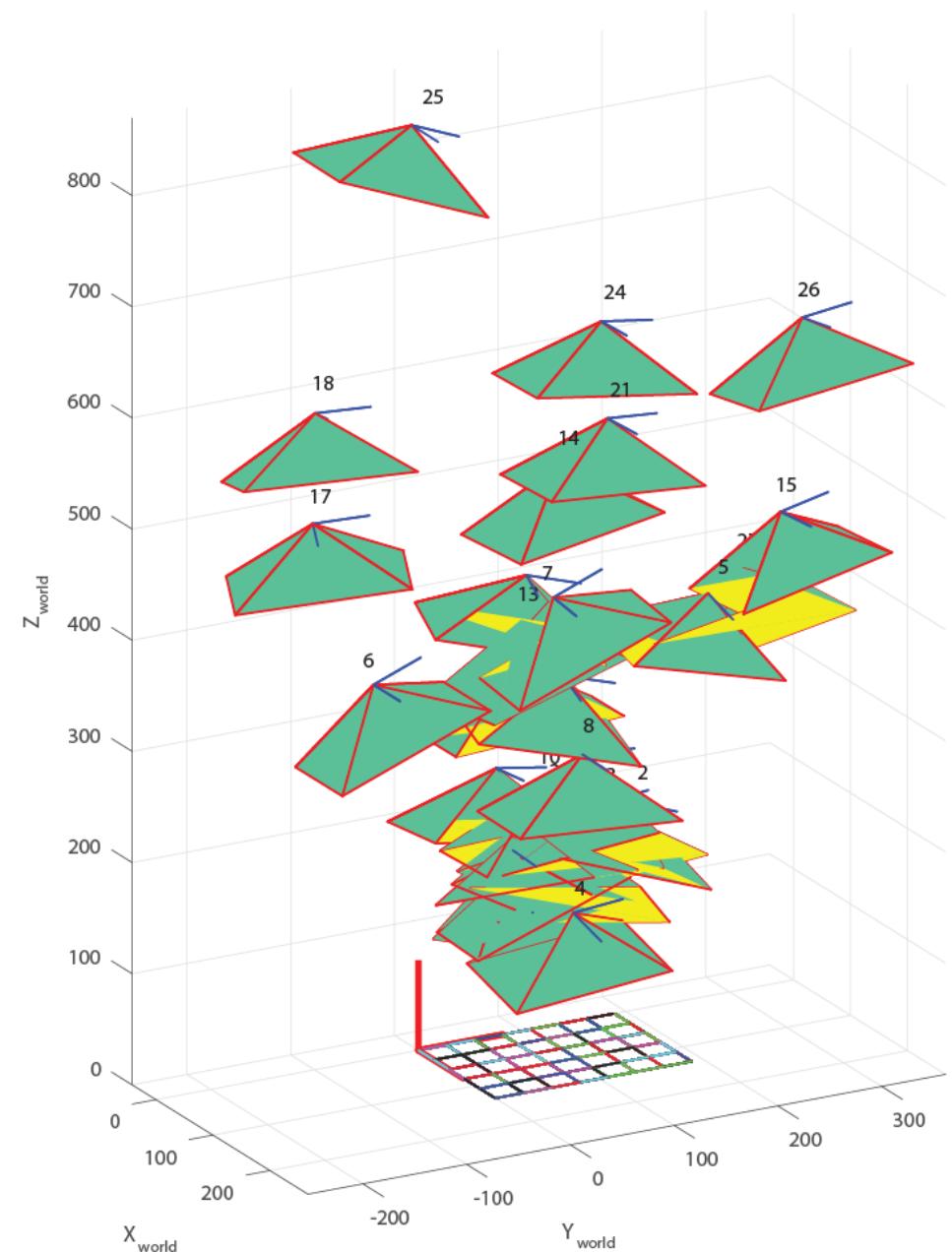
Processing image 27...
Using (wintx,winty)=(30,30) - Window size = 61x61      (Note: To reset the window size, run sc
Click on the four extreme corners of the rectangular complete pattern (the first clicked corne
Size of each square along the X direction: dX=27mm
Size of each square along the Y direction: dY=27mm      (Note: To reset the size of the squares,
```



Cf) calibration with vanishing points

$$K = \begin{bmatrix} 1317.2 & 0 & 1931.8 \\ 0 & 1317.2 & 1146.1 \\ 0 & 0 & 1 \end{bmatrix}$$

Extrinsic parameters (world-centered)



FILE VARIABLE CODE SIMULINK ENVIRONMENT RESOURCES

C: > Users > hspark > Desktop > Lecture > calib > image

Current Folder

Name

- calib_data.mat
- image000001.jpg
- image000002.jpg
- image000003.jpg
- image000004.jpg
- image000005.jpg
- image000006.jpg
- image000007.jpg
- image000008.jpg
- image000009.jpg
- image000010.jpg
- image000011.jpg
- image000012.jpg
- image000013.jpg
- image000014.jpg
- image000015.jpg
- image000016.jpg
- image000017.jpg
- image000018.jpg
- image000019.jpg
- image000020.jpg
- image000021.jpg
- image000022.jpg
- image000023.jpg
- image000024.jpg
- image000025.jpg
- image000026.jpg
- image000027.jpg
- image000028.jpg

Command Window

New to MATLAB? See resources for [Getting Started](#).

done

Aspect ratio optimized (est_aspect_ratio = 1) -> both components of fc are estimated (DEFAULT).

Principal point optimized (center_optim=1) - (DEFAULT). To reject principal point, set center_optim

SI Camera Calibration Toolbox - Standard Version

Image names	Read images	Extract grid corners	Calibration
Show Extrinsic	Reproject on images	Analyse error	Recomp. corners
Add/Suppress images	Save	Load	Exit
Comp. Extrinsic	Undistort image	Export calib data	Show calib results

Initialization of the intrinsic parameters - Number of images: 28

Calibration parameters after initialization:

```
Focal Length:      fc = [ 1432.81676   1432.81676 ]
Principal point:  cc = [ 1919.50000   1079.50000 ]
Skew:             alpha_c = [ 0.00000 ] => angle of pixel = 90.00000 degrees
Distortion:       kc = [ 0.00000   0.00000   0.00000   0.00000   0.00000 ]
```

Main calibration optimization procedure - Number of images: 28

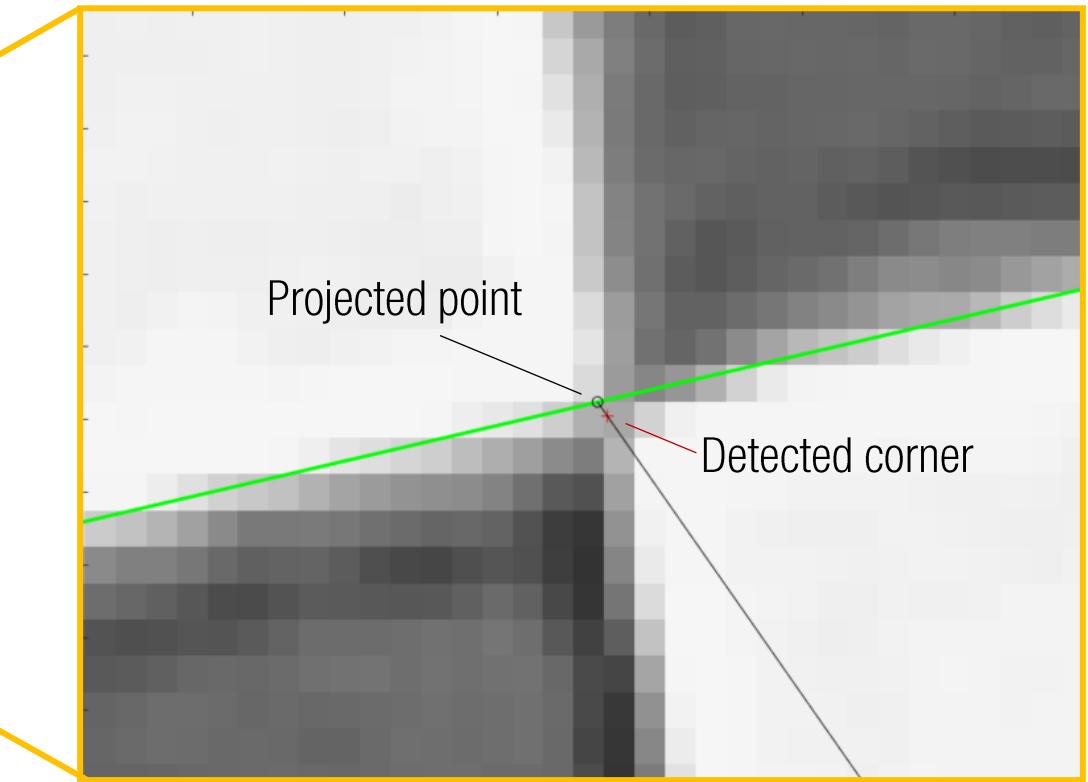
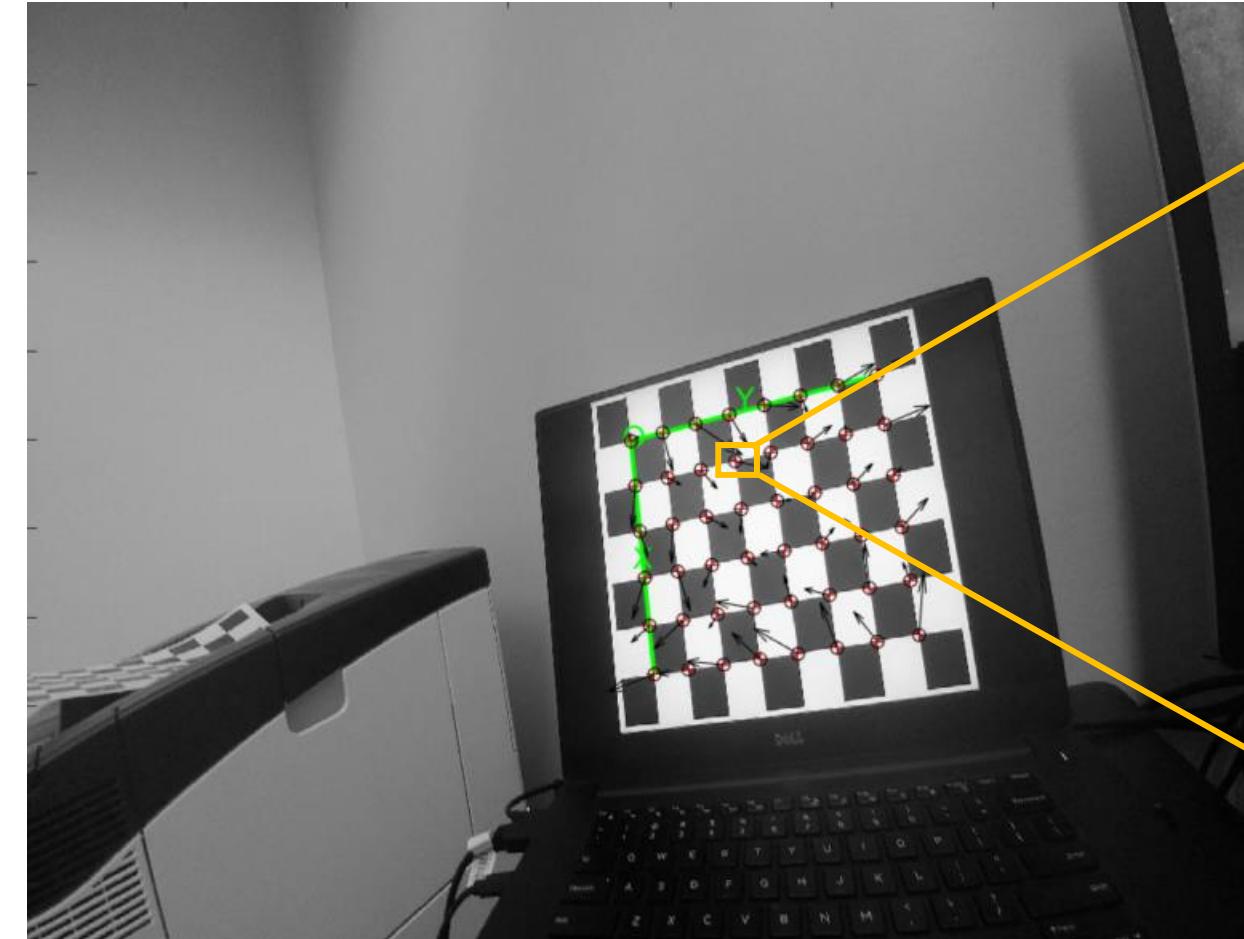
Gradient descent iterations: 1...2...3...4...5...6...7...8...9...10...11...12...13...14...15...16...

Estimation of uncertainties...done

Calibration results after optimization (with uncertainties):

```
Focal Length:      fc = [ 1370.51537   1369.77187 ] +/- [ 4.43385   4.39381 ]
Principal point:  cc = [ 1951.21042   1043.25952 ] +/- [ 2.27887   2.00849 ]
Skew:             alpha_c = [ 0.00000 ] +/- [ 0.00000 ] => angle of pixel axes = 90.00000 +/- 0.
Distortion:       kc = [ -0.02790   -0.01104   -0.00060   0.00001   0.00000 ] +/- [ 0.00169   0
Pixel error:     err = [ 0.45444   0.36957 ]
```

Note: The numerical errors are approximately three times the standard deviations (for reference).



Projected point

Detected corner



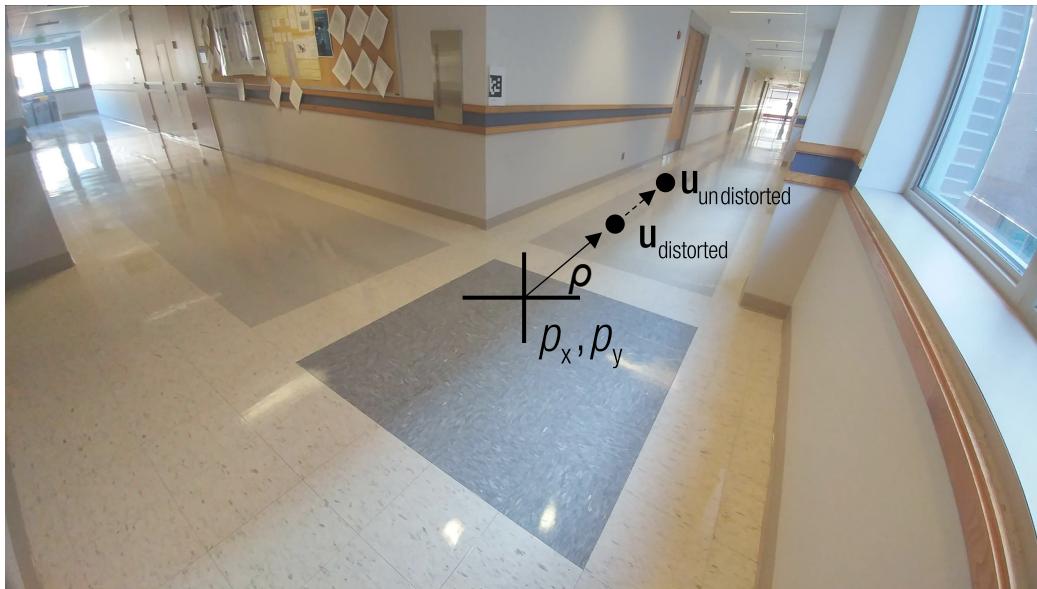
Lens Radial Distortion



Lens Radial Distortion Correction

Radial Distortion Model

Assumption: Lens distortion is a function of distance from the principal point.

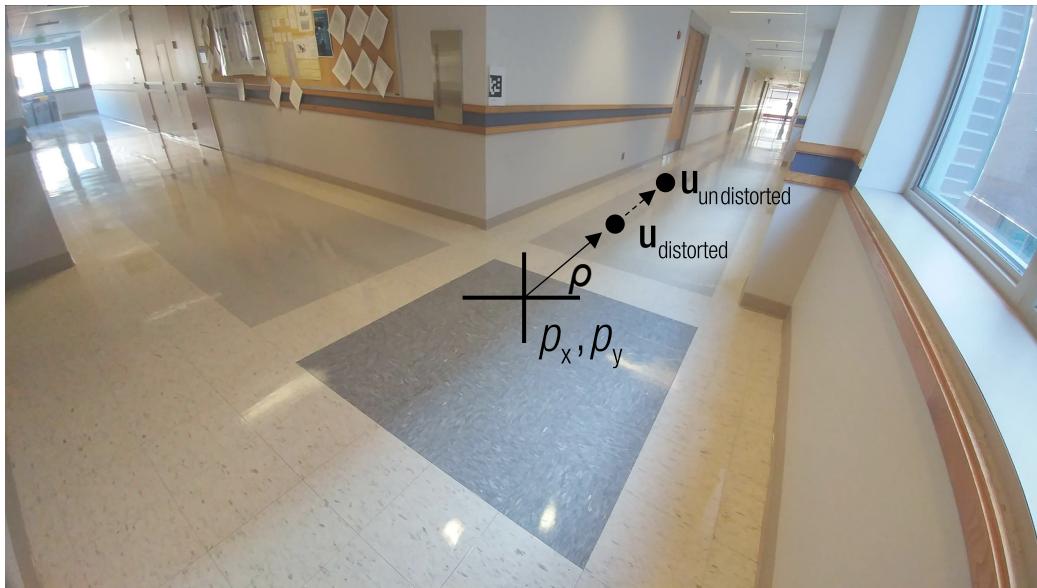


Normalized point:

$$\bar{\mathbf{u}}_{\text{distorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{distorted}}, \quad \bar{\mathbf{u}}_{\text{undistorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{undistorted}}$$

Radial Distortion Model

Assumption: Lens distortion is a function of distance from the principal point.



Normalized point:

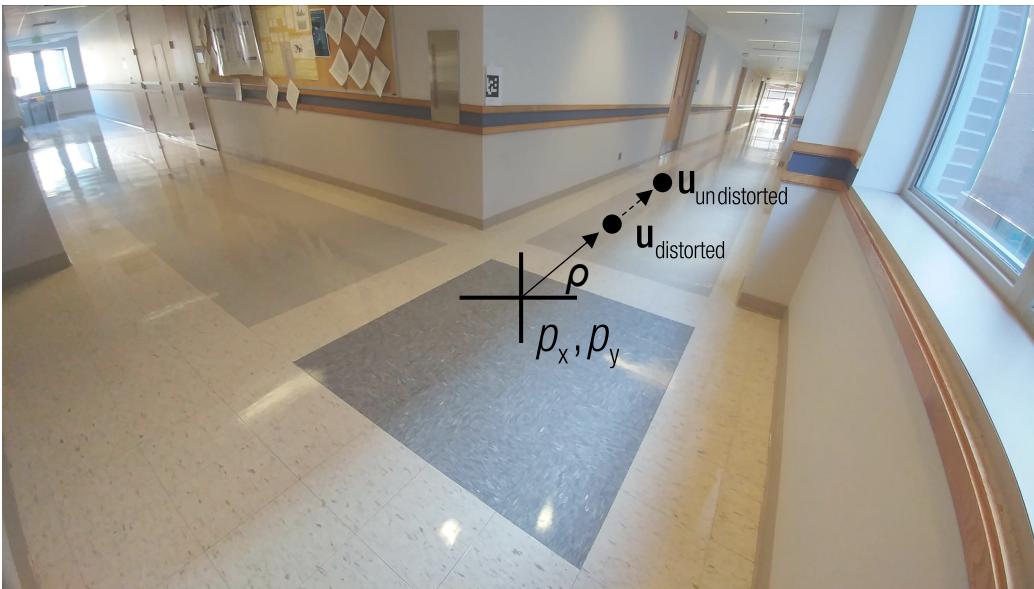
$$\bar{\mathbf{u}}_{\text{distorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{distorted}}, \quad \bar{\mathbf{u}}_{\text{undistorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{undistorted}}$$

$$\bar{\mathbf{u}}_{\text{distorted}} = L(\rho) \bar{\mathbf{u}}_{\text{undistorted}}$$

$$\text{where } \rho = \left\| \mathbf{K}^{-1} \bar{\mathbf{u}}_{\text{distorted}} \right\|$$

$$L(\rho) = 1 + k_1 \rho^2 + k_2 \rho^4 + \dots$$

Radial Distortion Parameter Estimation (2nd order)

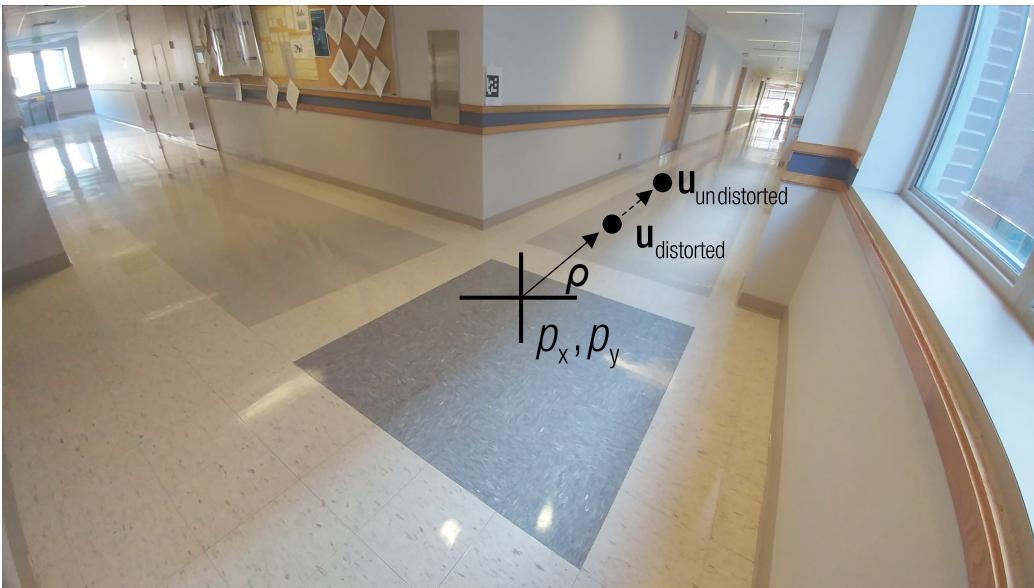


Normalized point:

$$\bar{\mathbf{u}}_{\text{distorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{distorted}}, \quad \bar{\mathbf{u}}_{\text{undistorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{undistorted}}$$

$$\bar{\mathbf{u}}_{\text{distorted}} = (1 + k_1 \rho^2 + k_2 \rho^4) \bar{\mathbf{u}}_{\text{undistorted}}$$

Radial Distortion Parameter Estimation (2nd order)



Normalized point:

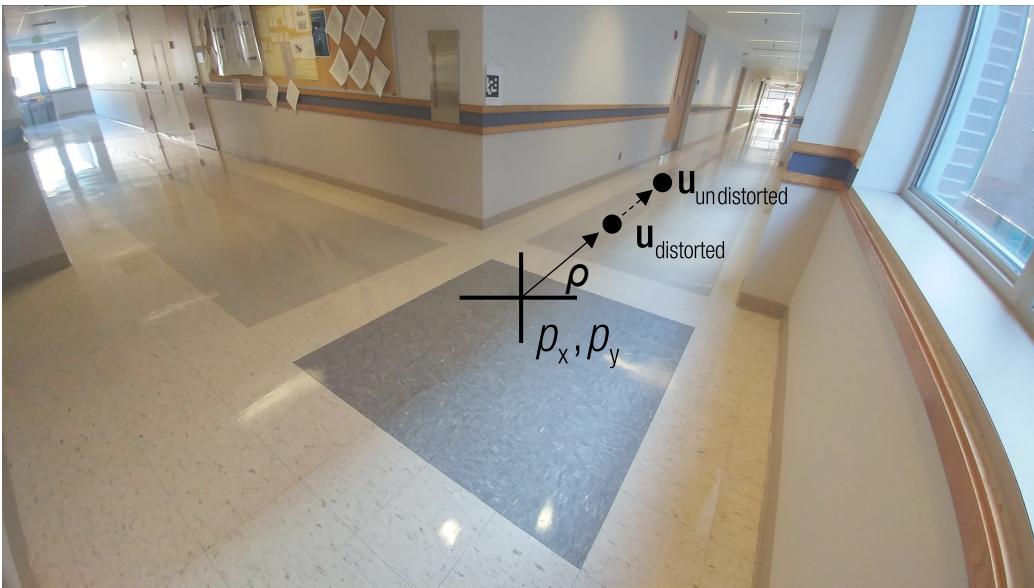
$$\bar{\mathbf{u}}_{\text{distorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{distorted}}, \quad \bar{\mathbf{u}}_{\text{undistorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{undistorted}}$$

$$\bar{\mathbf{u}}_{\text{distorted}} = (1 + k_1 \rho^2 + k_2 \rho^4) \bar{\mathbf{u}}_{\text{undistorted}}$$

$$\begin{bmatrix} \rho^2 \bar{\mathbf{u}}_{\text{undistorted}}^1 & \rho^4 \bar{\mathbf{u}}_{\text{undistorted}}^1 \\ \vdots & \vdots \\ \rho^2 \bar{\mathbf{u}}_{\text{undistorted}}^m & \rho^4 \bar{\mathbf{u}}_{\text{undistorted}}^m \end{bmatrix} \begin{bmatrix} k_1 \\ k_2 \end{bmatrix} = \begin{bmatrix} \bar{\mathbf{u}}_{\text{distorted}}^1 - \bar{\mathbf{u}}_{\text{undistorted}}^1 \\ \vdots \\ \bar{\mathbf{u}}_{\text{distorted}}^m - \bar{\mathbf{u}}_{\text{undistorted}}^m \end{bmatrix}$$

m: # of points

Radial Distortion Parameter Estimation (2nd order)



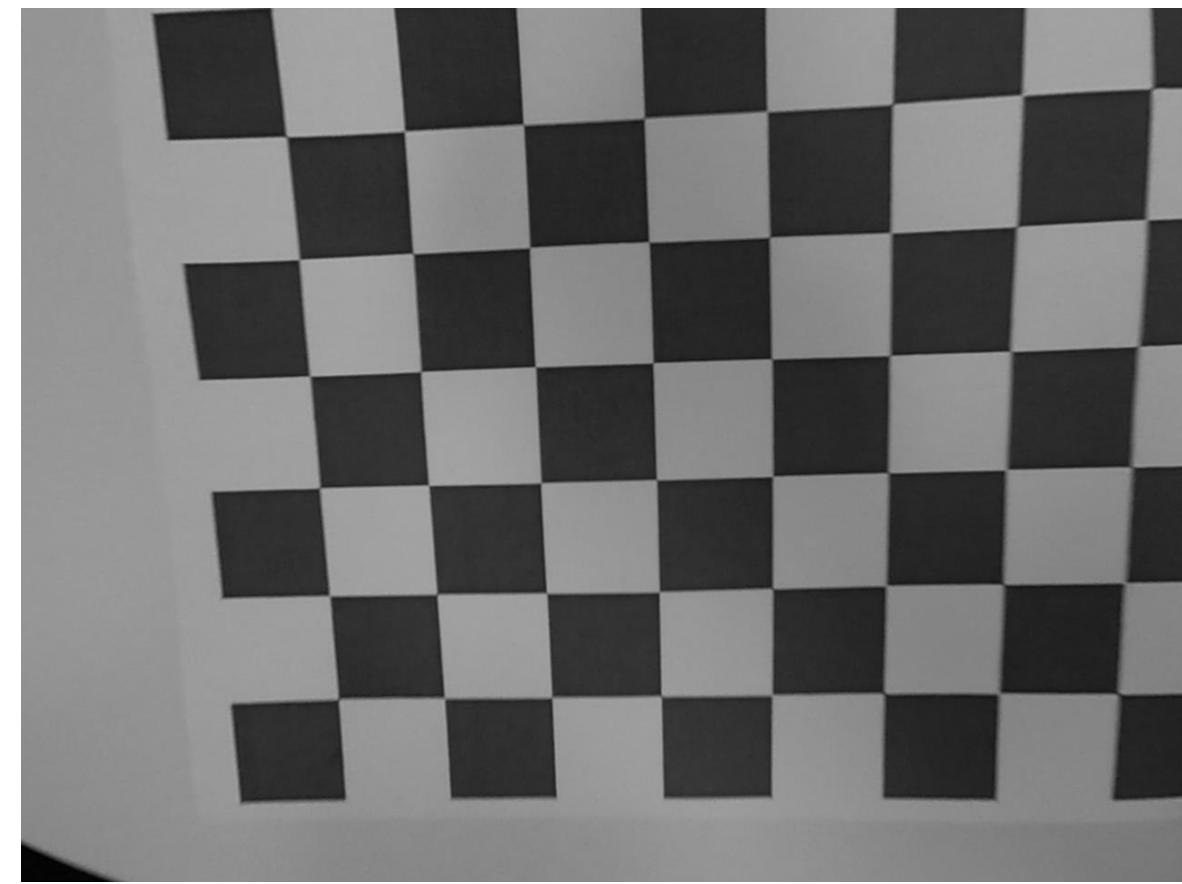
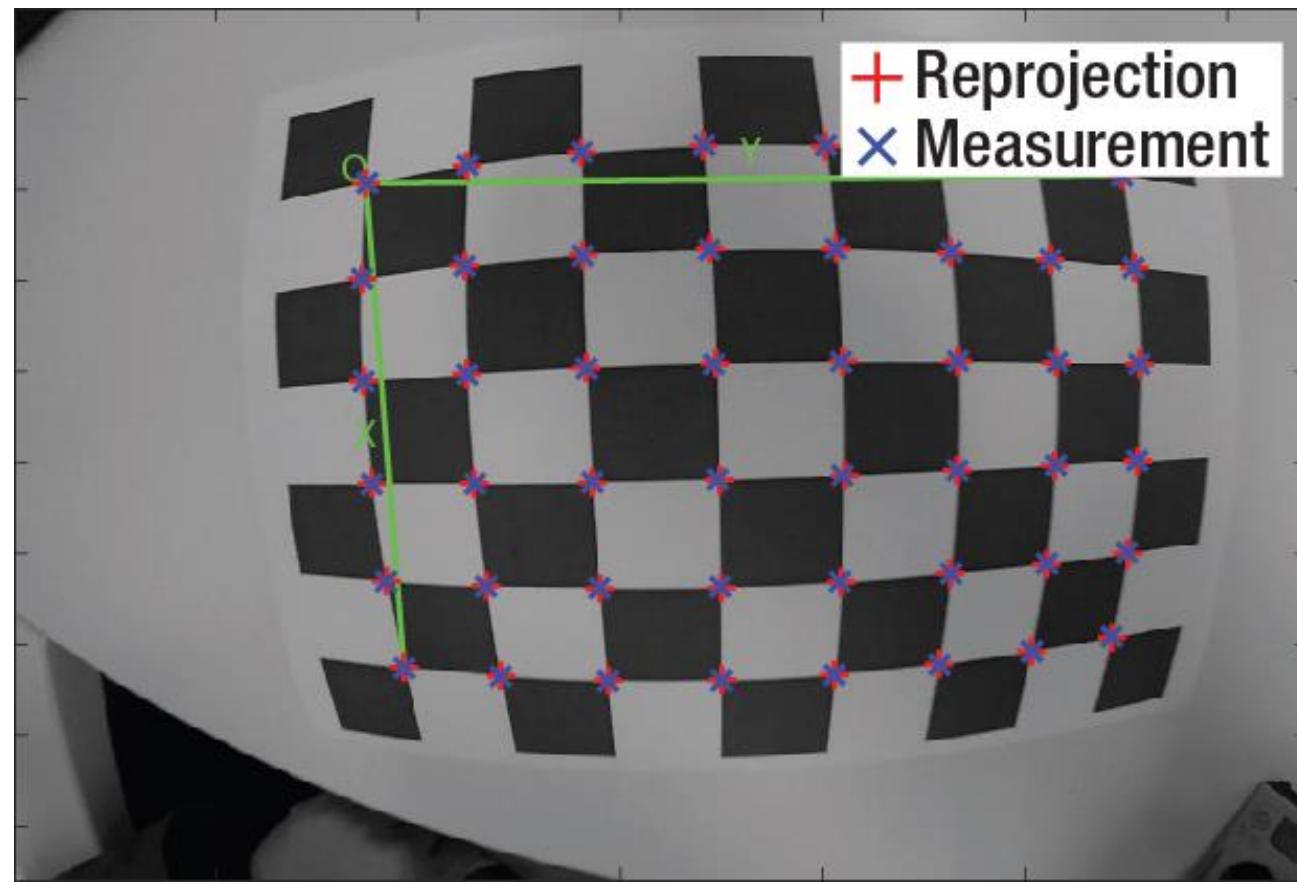
Normalized point:

$$\bar{\mathbf{u}}_{\text{distorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{distorted}}, \quad \bar{\mathbf{u}}_{\text{undistorted}} = \mathbf{K}^{-1} \mathbf{u}_{\text{undistorted}}$$

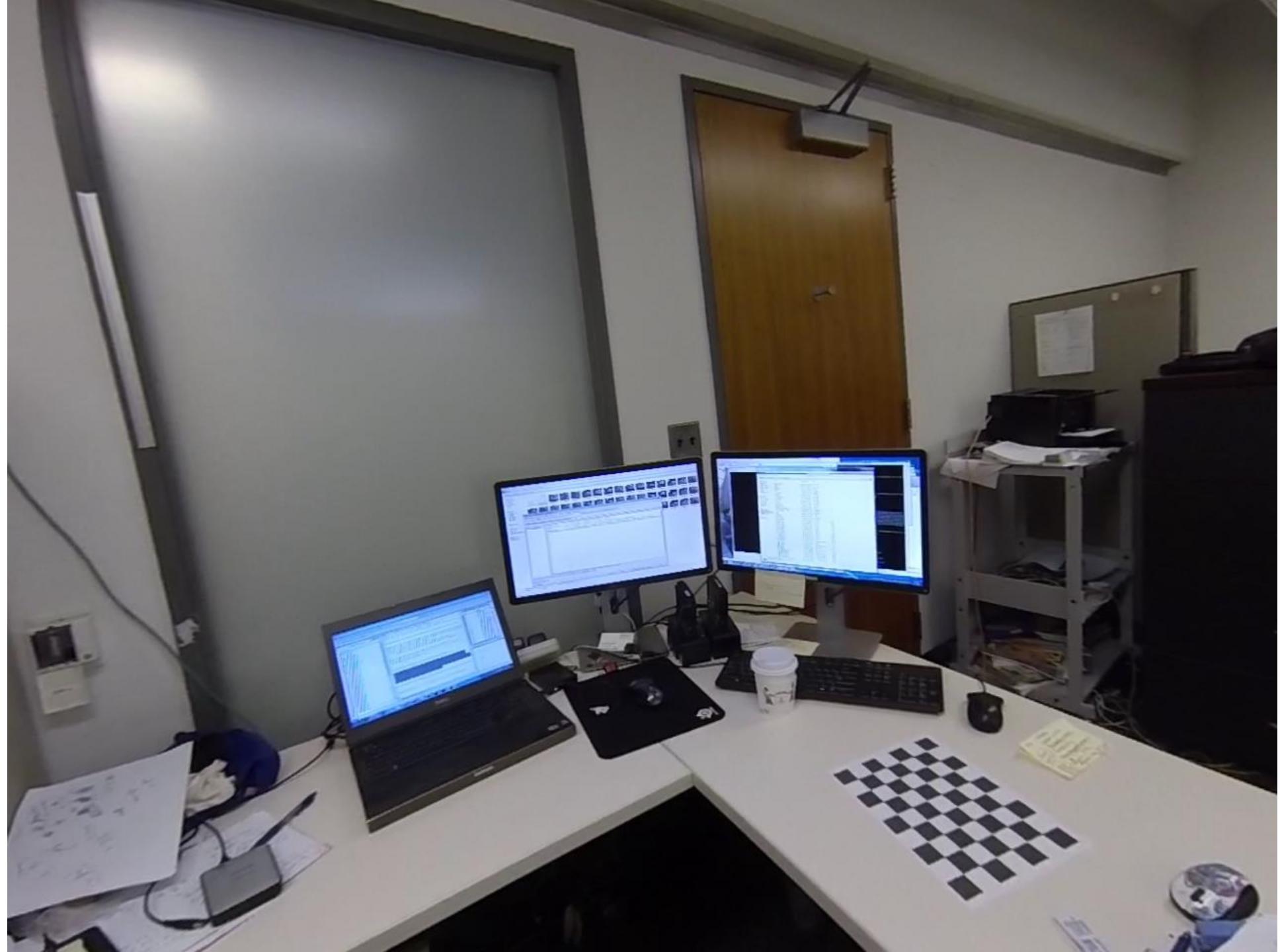
$$\bar{\mathbf{u}}_{\text{distorted}} = (1 + k_1 \rho^2 + k_2 \rho^4) \bar{\mathbf{u}}_{\text{undistorted}}$$

$$\begin{bmatrix} \rho^2 \bar{\mathbf{u}}_{\text{undistorted}}^1 & \rho^4 \bar{\mathbf{u}}_{\text{undistorted}}^1 \\ \vdots & \vdots \\ \rho^2 \bar{\mathbf{u}}_{\text{undistorted}}^m & \rho^4 \bar{\mathbf{u}}_{\text{undistorted}}^m \end{bmatrix} \begin{bmatrix} \mathbf{X} \\ \mathbf{k} \end{bmatrix} = \begin{bmatrix} \bar{\mathbf{u}}_{\text{distorted}}^1 - \bar{\mathbf{u}}_{\text{undistorted}}^1 \\ \bar{\mathbf{u}}_{\text{distorted}}^m - \bar{\mathbf{u}}_{\text{undistorted}}^m \end{bmatrix}$$

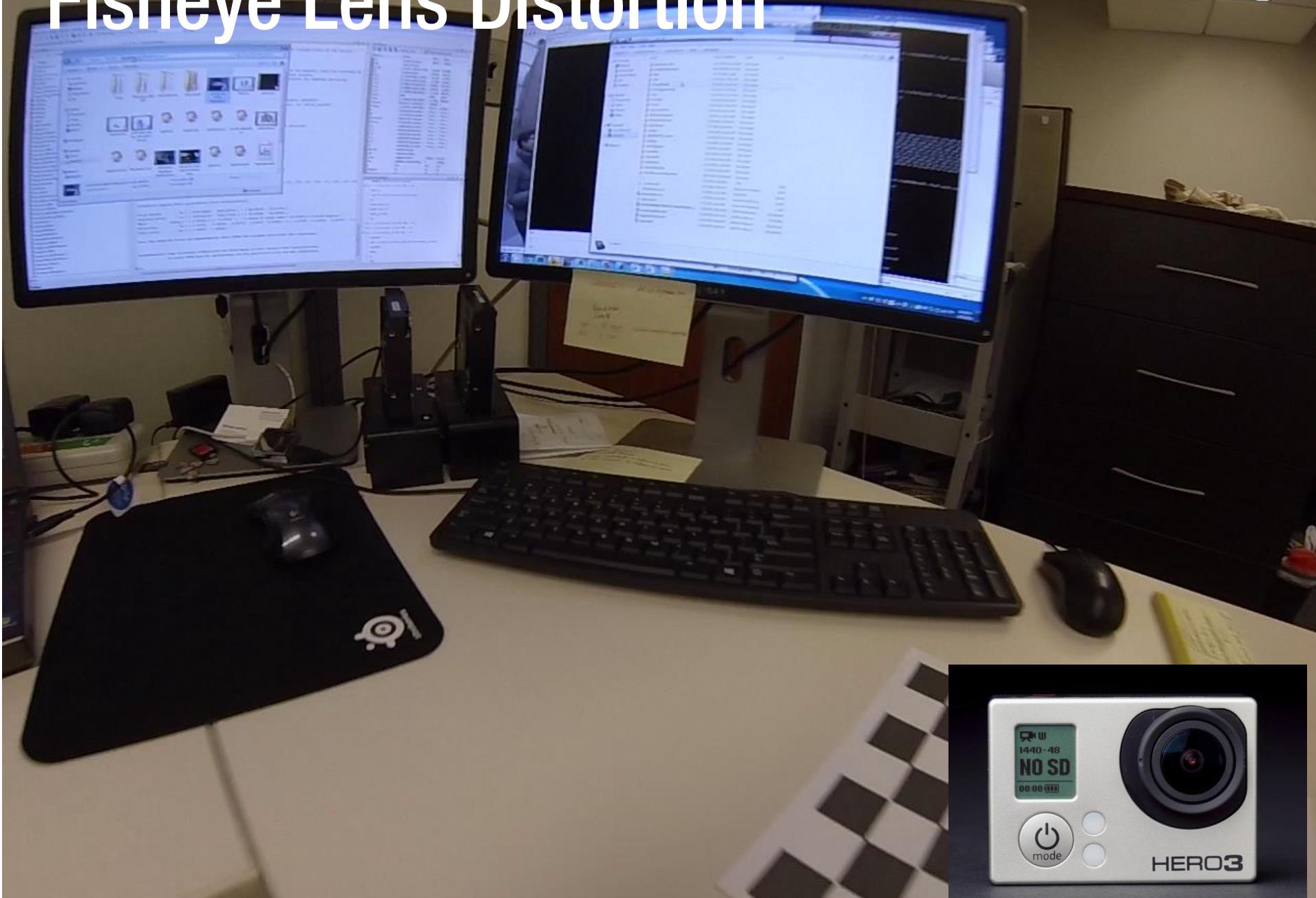
m: # of points







Fisheye Lens Distortion



Fisheye Lens Distortion

