

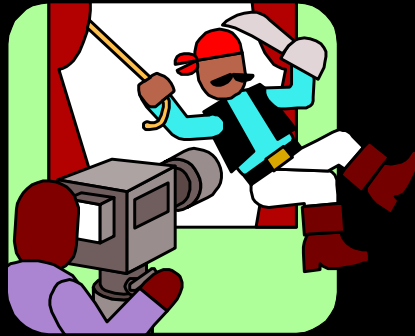


Introduction to

Computer Vision

3D Vision

CSC I6716
Spring 2004



Topic 6 of Part 2
Calibration

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<http://www-cs.engr.ccny.cuny.edu/~zhu/VisionCourse-2004.html>

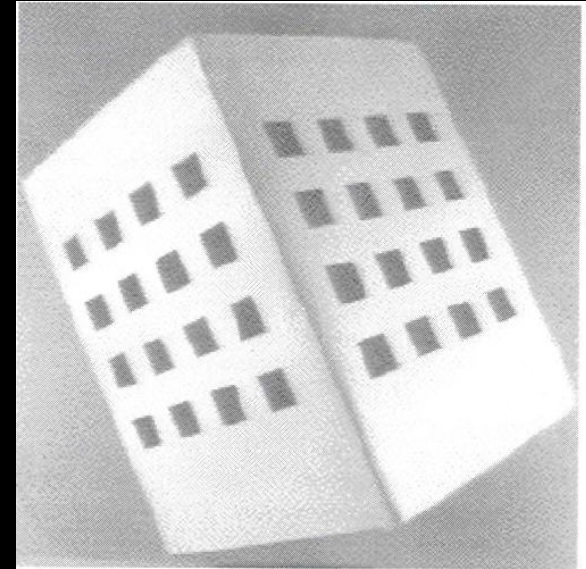
- Calibration: Find the intrinsic and extrinsic parameters
 - Problem and assumptions
 - Direct parameter estimation approach
 - Projection matrix approach

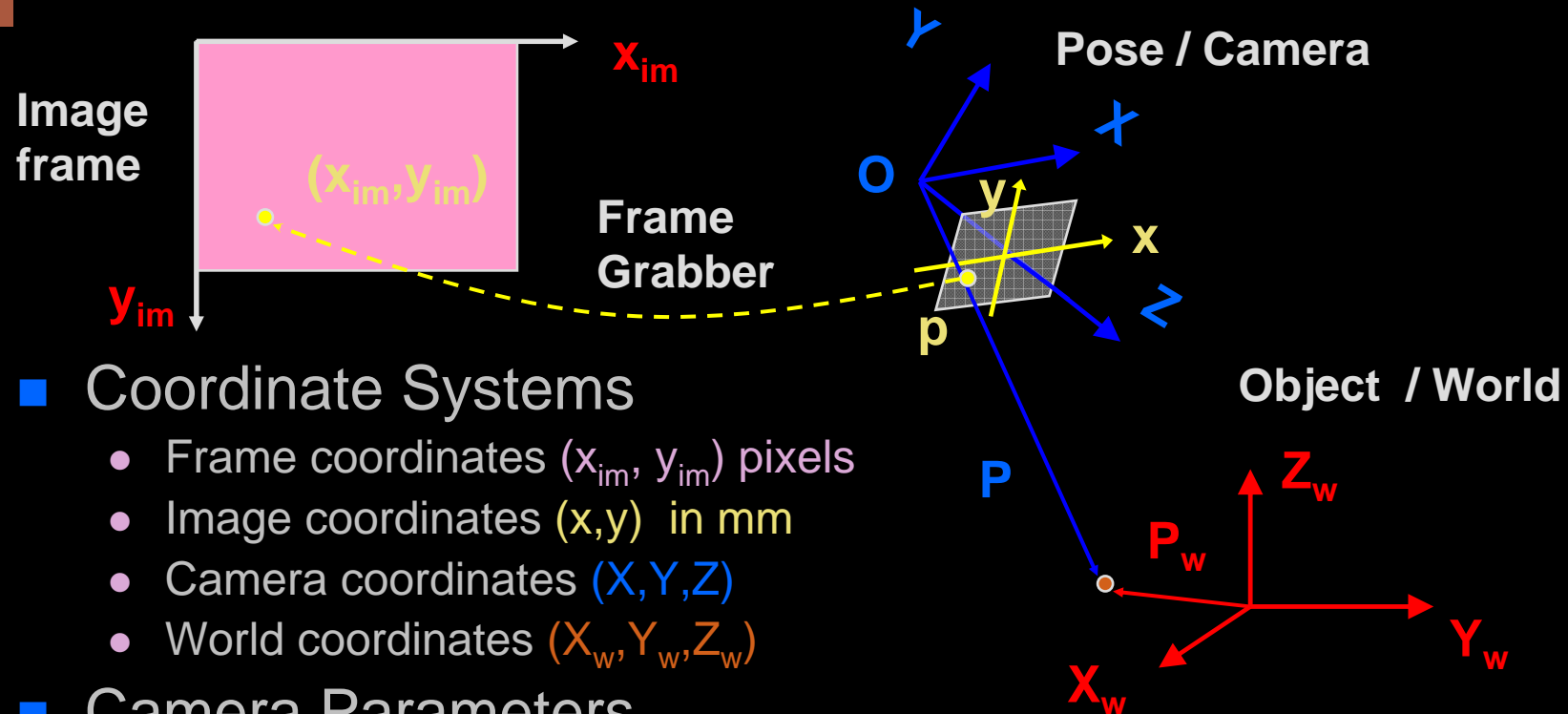
- Direct Parameter Estimation Approach
 - Basic equations (from Lecture 5)
 - Estimating the Image center using vanishing points
 - **SVD (Singular Value Decomposition) and Homogeneous System**
 - Focal length, Aspect ratio, and extrinsic parameters
 - **Discussion: Why not do all the parameters together?**

- Projection Matrix Approach (...after-class reading)
 - Estimating the projection matrix M
 - Computing the camera parameters from M
 - Discussion

- Comparison and Summary
 - Any difference?

- Given one or more images of a calibration pattern,
- Estimate
 - The intrinsic parameters
 - The extrinsic parameters, or
 - **BOTH**
- Issues: Accuracy of Calibration
 - How to design and measure the calibration pattern
 - Distribution of the control points to assure stability of solution – **not coplanar**
 - Construction tolerance one or two order of magnitude smaller than the desired accuracy of calibration
 - e.g. 0.01 mm tolerance versus 0.1mm desired accuracy
 - How to extract the image correspondences
 - Corner detection?
 - Line fitting?
 - Algorithms for camera calibration given both 3D-2D pairs
- **Alternative approach: 3D from un-calibrated camera**





■ Coordinate Systems

- Frame coordinates (x_{im}, y_{im}) pixels
- Image coordinates (x, y) in mm
- Camera coordinates (X, Y, Z)
- World coordinates (X_w, Y_w, Z_w)

■ Camera Parameters

- Intrinsic Parameters (of the camera and the frame grabber): link the **frame coordinates** of an image point with its corresponding **camera coordinates**
- Extrinsic parameters: define the location and orientation of the **camera coordinate system** with respect to the **world coordinate system**

Linear Version of Perspective Projection

World to Camera

- Camera: $P = (X, Y, Z)^T$
- World: $P_w = (X_w, Y_w, Z_w)$
- Transform: R, T

$$\mathbf{P} = \mathbf{R}\mathbf{P}_w + \mathbf{T} = \begin{pmatrix} r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x \\ r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y \\ r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z \end{pmatrix} = \begin{bmatrix} \mathbf{R}_1^T \mathbf{P}_w + T_x \\ \mathbf{R}_2^T \mathbf{P}_w + T_y \\ \mathbf{R}_3^T \mathbf{P}_w + T_z \end{bmatrix}$$

Camera to Image

- Camera: $P = (X, Y, Z)^T$
- Image: $p = (x, y)^T$
- Not linear equations

$$(x, y) = \left(f \frac{X}{Z}, f \frac{Y}{Z} \right)$$

Image to Frame

- Neglecting distortion
- Frame $(x_{im}, y_{im})^T$

$$x = -(x_{im} - o_x)s_x$$

$$y = -(y_{im} - o_y)s_y$$

World to Frame

- $(X_w, Y_w, Z_w)^T \rightarrow (x_{im}, y_{im})^T$
- Effective focal lengths
 - $f_x = f/s_x, f_y = f/s_y$

$$x_{im} - o_x = -f_x \frac{r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

$$y_{im} - o_y = -f_y \frac{r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

- Extrinsic Parameters
 - R , 3x3 rotation matrix
 - Three angles α, β, γ
 - T , 3-D translation vector

$$x' = x_{im} - o_x = -f_x \frac{r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

$$y' = y_{im} - o_y = -f_y \frac{r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

- Intrinsic Parameters
 - f_x, f_y : effective focal length in pixel
 - $\alpha = f_x/f_y = s_y/s_x$, and f_x
 - (o_x, o_y) : known Image center $\rightarrow (x, y)$ known
 - k_1 , radial distortion coefficient: **neglect it in the basic algorithm**

- Same Denominator in the two Equations

- Known : (X_w, Y_w, Z_w) and its (x, y)
- Unknown: $r_{pq}, T_x, T_y, f_x, f_y$

$$f_y (r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y) / y' = f_x (r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x) / x'$$

$$x' f_y (r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y) = y' f_x (r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x)$$

- Linear Equation of 8 unknowns $\mathbf{v} = (v_1, \dots, v_8)$
 - Aspect ratio: $\alpha = f_x/f_y$
 - Point pairs , $\{(X_i, Y_i, Z_i) \leftrightarrow (x_i, y_i)\}$ drop the ' and subscript "w"

$$x'(r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y) = y'\alpha(r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x)$$



$$x_i X_i r_{21} + x_i Y_i r_{22} + x_i Z_i r_{23} + x_i T_y - y_i X_i (\alpha r_{11}) - y_i Y_i (\alpha r_{12}) - y_i Z_i (\alpha r_{13}) - y_i (\alpha T_x) = 0$$



$$x_i X_i v_1 + x_i Y_i v_2 + x_i Z_i v_3 + x_i v_4 - y_i X_i v_5 - y_i Y_i v_6 - y_i Z_i v_7 - y_i v_8 = 0$$

$$(v_1, v_2, v_3, v_4, v_5, v_6, v_7, v_8)$$

$$= (r_{21}, r_{22}, r_{23}, T_y, \alpha r_{11}, \alpha r_{12}, \alpha r_{13}, \alpha T_x)$$

- Homogeneous System of N Linear Equations
 - Given N corresponding pairs $\{(X_i, Y_i, Z_i) \leftrightarrow (x_i, y_i)\}$, $i=1,2,\dots,N$
 - 8 unknowns $\mathbf{v} = (v_1, \dots, v_8)^T$, **7 independent parameters**

$$x_i X_i v_1 + x_i Y_i v_2 + x_i Z_i v_3 + x_i v_4 - y_i X_i v_5 - y_i Y_i v_6 - y_i Z_i v_7 - y_i v_8 = 0$$



$$\mathbf{A} \mathbf{v} = \mathbf{0}$$

$$\mathbf{A} = \begin{bmatrix} x_1 X_1 & x_1 Y_1 & x_1 Z_1 & x_1 & -y_1 X_1 & -y_1 Y_1 & -y_1 Z_1 & -y_1 \\ x_2 X_2 & x_2 Y_2 & x_2 Z_2 & x_2 & -y_2 X_2 & -y_2 Y_2 & -y_2 Z_2 & -y_2 \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \\ x_N X_N & x_N Y_N & x_N Z_N & x_N & -y_N X_N & -y_N Y_N & -y_N Z_N & -y_N \end{bmatrix}$$

- The system has a nontrivial solution (up to a scale)
 - IF $N \geq 7$ and N points are not coplanar $\Rightarrow \text{Rank}(\mathbf{A}) = 7$
 - Can be determined from the SVD of A

■ Singular Value Decomposition:

- Any $m \times n$ matrix can be written as the product of three matrices

$$\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{V}^T$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{1n} \\ a_{21} & a_{22} & a_{2n} \\ \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & a_{mn} \end{bmatrix} = \begin{bmatrix} u_{11} & u_{12} & u_{1m} \\ u_{21} & u_{22} & u_{2m} \\ \vdots & \vdots & \vdots \\ u_{m1} & u_{m2} & u_{mm} \end{bmatrix} \begin{bmatrix} \sigma_1 & 0 & 0 \\ 0 & \sigma_2 & 0 \\ \vdots & \vdots & \vdots \\ 0 & 0 & \sigma_n \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} v_{11} & v_{21} & v_{n1} \\ v_{12} & v_{22} & v_{n2} \\ \vdots & \vdots & \vdots \\ v_{1n} & v_{2n} & v_{nn} \end{bmatrix}$$

The first column of \mathbf{U} is labeled \mathbf{U}_1 and the first column of \mathbf{V} is labeled \mathbf{V}_1 .

- Singular values σ_i are fully determined by \mathbf{A}
 - \mathbf{D} is diagonal: $d_{ij} = 0$ if $i \neq j$; $d_{ii} = \sigma_i$ ($i=1,2,\dots,n$)
 - $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n \geq 0$
- Both \mathbf{U} and \mathbf{V} are not unique
 - Columns of each are mutual orthogonal vectors

1. Singularity and Condition Number

$$\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{V}^T$$

- $n \times n$ \mathbf{A} is nonsingular IFF all singular values are nonzero
- Condition number : degree of singularity of \mathbf{A}
 - \mathbf{A} is ill-conditioned if $1/C$ is comparable to the arithmetic precision of your machine; almost singular

$$C = \sigma_1 / \sigma_n$$

2. Rank of a square matrix \mathbf{A}

- Rank (\mathbf{A}) = number of nonzero singular values

3. Inverse of a square Matrix

- If \mathbf{A} is nonsingular
- In general, the pseudo-inverse of \mathbf{A}

$$\mathbf{A}^{-1} = \mathbf{V}\mathbf{D}^{-1}\mathbf{U}^T$$

$$\mathbf{A}^+ = \mathbf{V}\mathbf{D}_0^{-1}\mathbf{U}^T$$

4. Eigenvalues and Eigenvectors (questions)

- Eigenvalues of both $\mathbf{A}^T\mathbf{A}$ and $\mathbf{A}\mathbf{A}^T$ are σ_i^2 ($\sigma_i > 0$)
- The columns of \mathbf{U} are the eigenvectors of $\mathbf{A}\mathbf{A}^T$ ($m \times m$)
- The columns of \mathbf{V} are the eigenvectors of $\mathbf{A}^T\mathbf{A}$ ($n \times n$)

$$\mathbf{A}\mathbf{A}^T \mathbf{u}_i = \sigma_i^2 \mathbf{u}_i$$

$$\mathbf{A}^T \mathbf{A} \mathbf{v}_i = \sigma_i^2 \mathbf{v}_i$$

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

■ Least Square

- Solve a system of m equations for n unknowns \mathbf{x} ($m \geq n$)
- A is a $m \times n$ matrix of the coefficients
- \mathbf{b} ($\neq 0$) is the m -D vector of the data
- Solution:

$$\underbrace{\mathbf{A}^T \mathbf{A} \mathbf{x} = \mathbf{A}^T \mathbf{b}}_{n \times n \text{ matrix}} \quad \longrightarrow \quad \mathbf{x} = \underbrace{(\mathbf{A}^T \mathbf{A})^+}_{\text{Pseudo-inverse}} \mathbf{A}^T \mathbf{b}$$

- **How to solve: compute the pseudo-inverse of $A^T A$ by SVD**
 - $(A^T A)^+$ is more likely to coincide with $(A^T A)^{-1}$ given $m > n$
 - Always a good idea to look at the condition number of $A^T A$

■ Homogeneous System

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

- m equations for n unknowns \mathbf{x} ($m \geq n-1$)
- Rank (A) = n-1 (by looking at the SVD of A)
- A non-trivial solution (up to an arbitrary scale) by SVD:
- **Simply proportional to the eigenvector corresponding to the only zero eigenvalue of $A^T A$ (n x n matrix)**

■ Note:

- All the other eigenvalues are positive because Rank (A)=n-1
- For a proof, see Textbook p. 324-325
- In practice, the eigenvector (i.e. \mathbf{v}_n) corresponding to the minimum eigenvalue of $A^T A$, i.e. σ_n^2

■ Problem Statements

- Numerical estimate of a matrix A whose entries are not independent
- Errors introduced by noise alter the estimate to \hat{A}

■ Enforcing Constraints by SVD

- Take orthogonal matrix A as an example
- Find the closest matrix to \hat{A} , which satisfies the constraints exactly

- SVD of \hat{A}

$$\hat{A} = \mathbf{U}\mathbf{D}\mathbf{V}^T$$

- Observation: $\mathbf{D} = \mathbf{I}$ (all the singular values are 1) if A is orthogonal
- Solution: changing the singular values to those expected

$$\mathbf{A} = \mathbf{U}\mathbf{I}\mathbf{V}^T$$

$$\mathbf{A}\mathbf{v} = \mathbf{0}$$

- Homogeneous System of N Linear Equations
 - Given N corresponding pairs $\{(X_i, Y_i, Z_i) \leftrightarrow (x_i, y_i)\}$, $i=1,2,\dots,N$
 - 8 unknowns $\mathbf{v} = (v_1, \dots, v_8)^T$, **7 independent parameters**
- The system has a nontrivial solution (up to a scale)
 - IF $N \geq 7$ and N points are not coplanar $\Rightarrow \text{Rank}(\mathbf{A}) = 7$
 - Can be determined from the SVD of A
 - Rows of \mathbf{V}^T : eigenvectors $\{\mathbf{e}_i\}$ of $\mathbf{A}^T\mathbf{A}$
 - Solution: **the 8th row \mathbf{e}_8 corresponding to the only zero singular value $\lambda_8=0$**
- Practical Consideration
 - The errors in localizing image and world points may make the rank of A to be maximum (8)
 - In this case select the eigenvector corresponding to the smallest eigenvalue.

$$\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{V}^T$$

$$\bar{\mathbf{v}} = c\mathbf{e}_8$$

- Equations for scale factor γ and aspect ratio α

$$\bar{\mathbf{v}} = \gamma (r_{21}, r_{22}, r_{23}, T_y, \alpha r_{11}, \alpha r_{12}, \alpha r_{13}, \alpha T_x)$$

$\mathbf{v}_1 \quad \mathbf{v}_2 \quad \mathbf{v}_3 \quad \mathbf{v}_4 \quad \mathbf{v}_5 \quad \mathbf{v}_6 \quad \mathbf{v}_7 \quad \mathbf{v}_8$

- Knowledge: \mathbf{R} is an orthogonal matrix

$$\mathbf{R}_i^T \mathbf{R}_j = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$$

$$\mathbf{R} = (r_{ij})_{3 \times 3} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} = \begin{bmatrix} \mathbf{R}_1^T \\ \mathbf{R}_2^T \\ \mathbf{R}_3^T \end{bmatrix}$$

- Second row ($i=j=2$):

$$r_{21}^2 + r_{22}^2 + r_{23}^2 = 1$$



$$|\gamma| = \sqrt{\bar{v}_1^2 + \bar{v}_2^2 + \bar{v}_3^2}$$



$$|\gamma|$$

- First row ($i=j=1$):

$$r_{11}^2 + r_{12}^2 + r_{13}^2 = 1$$



$$\alpha |\gamma| = \sqrt{\bar{v}_5^2 + \bar{v}_6^2 + \bar{v}_7^2}$$



$$\alpha$$

- Equations for first 2 rows of R and T given α and $|\gamma|$

$$\bar{\mathbf{v}} = s |\gamma| (r_{21}, r_{22}, r_{23}, T_y, \alpha r_{11}, \alpha r_{12}, \alpha r_{13}, \alpha T_x)$$

- First 2 rows of R and T can be found up to a common sign s (+ or -)

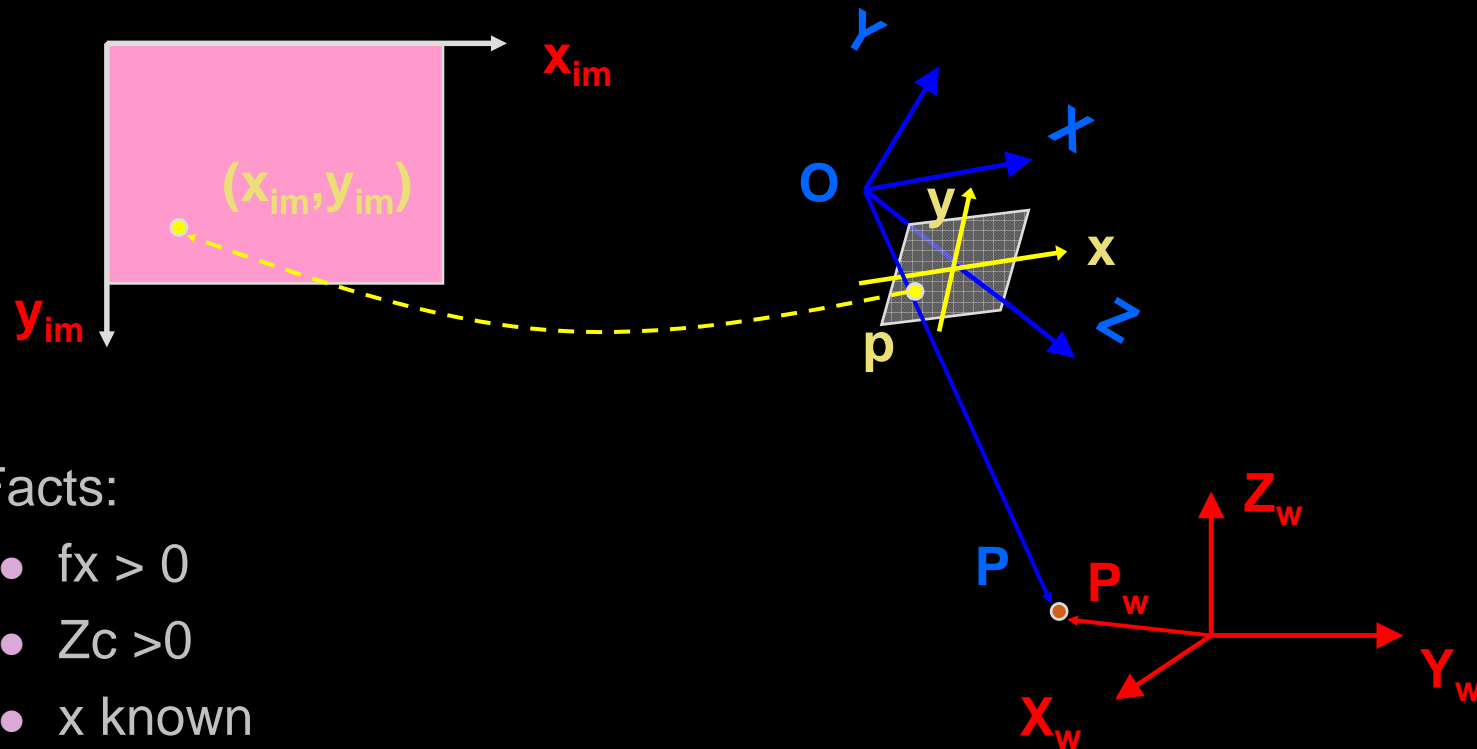
$$s\mathbf{R}_1^T, s\mathbf{R}_2^T, sT_x, sT_y$$

- The third row of the rotation matrix by vector product

$$\mathbf{R}_3^T = \mathbf{R}_1^T \times \mathbf{R}_2^T = s\mathbf{R}_1^T \times s\mathbf{R}_2^T$$

- Remaining Questions :
 - How to find the sign s ?
 - Is R orthogonal?
 - How to find T_z and f_x, f_y ?

$$\mathbf{R} = (r_{ij})_{3 \times 3} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} = \begin{bmatrix} \mathbf{R}_1^T \\ \mathbf{R}_2^T \\ \mathbf{R}_3^T \end{bmatrix}$$



- Facts:

- $f_x > 0$
- $Z_c > 0$
- x known
- X_w, Y_w, Z_w known

- Solution

- ⇒ Check the sign of X_c
- ⇒ Should be opposite to x

$$x = -f_x \frac{X_c}{Z_c} = -f_x \frac{r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

$$y = -f_y \frac{Y_c}{Z_c} = -f_y \frac{r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

- Question:
 - First 2 rows of R are calculated without using the mutual orthogonal constraint

$$\hat{\mathbf{R}}^T \hat{\mathbf{R}} = \mathbf{I}?$$

$$\mathbf{R} = (r_{ij})_{3 \times 3} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} = \begin{bmatrix} \mathbf{R}_1^T \\ \mathbf{R}_2^T \\ \mathbf{R}_3^T \end{bmatrix}$$

$$\mathbf{R}_3^T = \mathbf{R}_1^T \times \mathbf{R}_2^T = s\mathbf{R}_1^T \times s\mathbf{R}_2^T$$

- Solution:
 - Use SVD of estimate R

$$\hat{\mathbf{R}} = \mathbf{U}\mathbf{D}\mathbf{V}^T$$



$$\mathbf{R} = \mathbf{U}\mathbf{I}\mathbf{V}^T$$



Replace the diagonal matrix D with the 3x3 identity matrix



■ Solution

- Solve the system of N linear equations with two unknown
 - T_x, f_x

$$x = -f_x \frac{r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

$$xT_z + \underbrace{(r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x)}_{a_{i2}} f_x = -x \underbrace{(r_{31}X_w + r_{32}Y_w + r_{33}Z_w)}_{b_i}$$

a_{i1} is labeled under xT_z .

- Least Square method

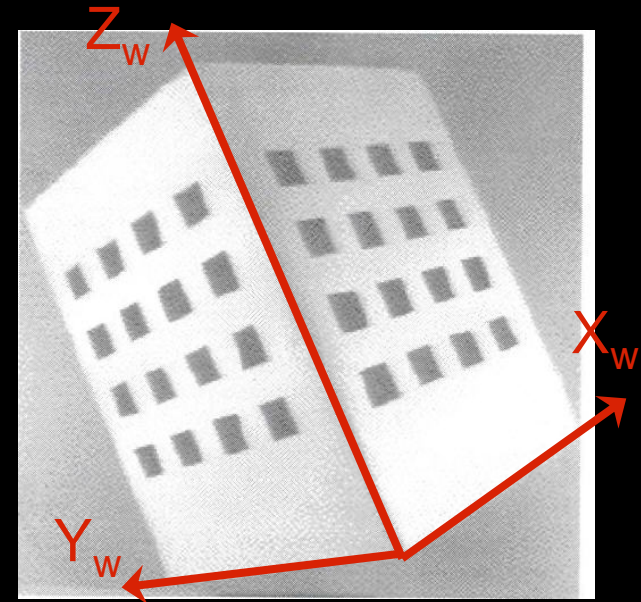
$$\begin{pmatrix} \hat{T}_z \\ \hat{f}_x \end{pmatrix} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}$$

$$\mathbf{A} \begin{pmatrix} T_z \\ f_x \end{pmatrix} = \mathbf{b}$$

- SVD method to find inverse

Algorithm (p130-131)

1. Measure N 3D coordinates (X_i, Y_i, Z_i)
2. Locate their corresponding image points (x_i, y_i) - Edge, Corner, Hough
3. Build matrix A of a homogeneous system $Av = 0$
4. Compute **SVD** of A , solution v
5. Determine aspect ratio α and scale $|\gamma|$
6. Recover the first two rows of R and the first two components of T up to a sign
7. Determine sign s of γ by checking the projection equation
8. Compute the 3rd row of R by vector product, and enforce orthogonality constraint by **SVD**
9. Solve T_z and f_x using Least Square and **SVD**, then $f_y = f_x / \alpha$



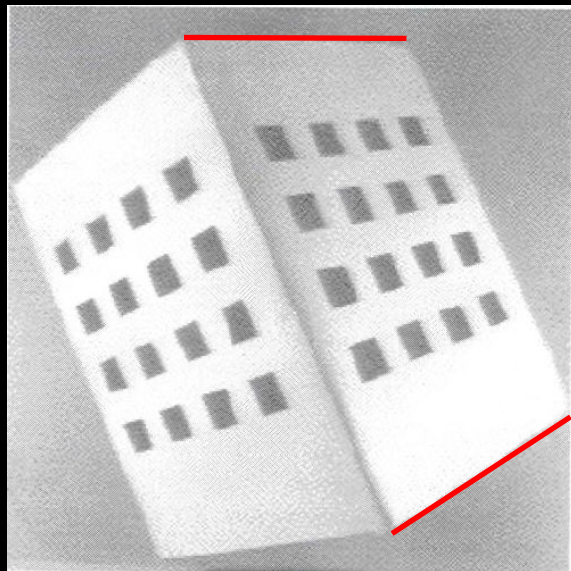
- Questions

- Can we select an arbitrary image center for solving other parameters?
- **How to find the image center (ox,oy)?**
- How about to include the radial distortion?
- Why not solve all the parameters once ?
 - How many unknown with ox, oy? --- 20 ??? – projection matrix method

$$x = x_{im} - o_x = -f_x \frac{r_{11}X_w + r_{12}Y_w + r_{13}Z_w + T_x}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$
$$y = y_{im} - o_y = -f_y \frac{r_{21}X_w + r_{22}Y_w + r_{23}Z_w + T_y}{r_{31}X_w + r_{32}Y_w + r_{33}Z_w + T_z}$$

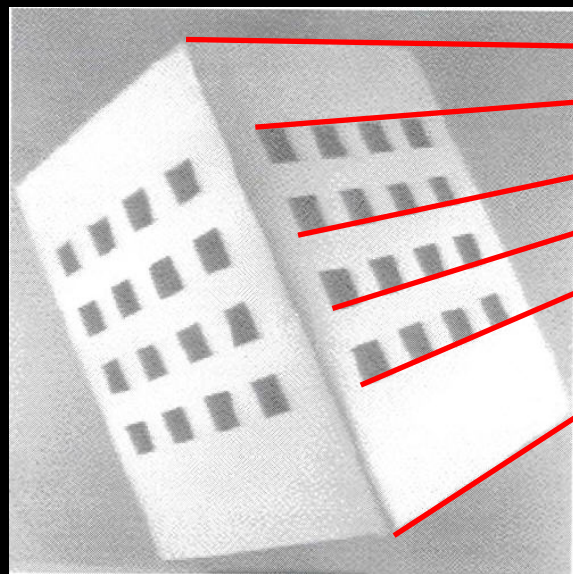
- Vanishing points:

- Due to perspective, all parallel lines in 3D space appear to meet in a point on the image - the vanishing point, which is the common intersection of all the image lines



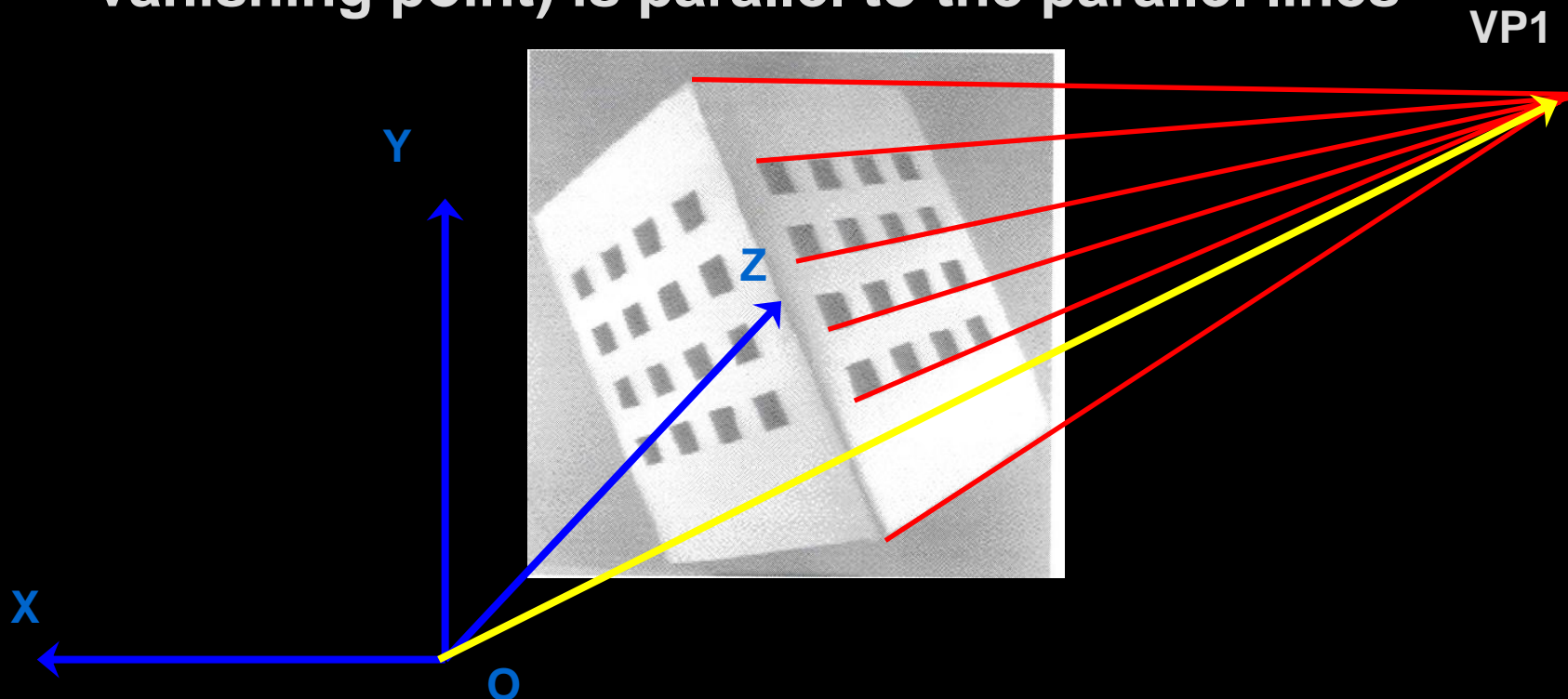
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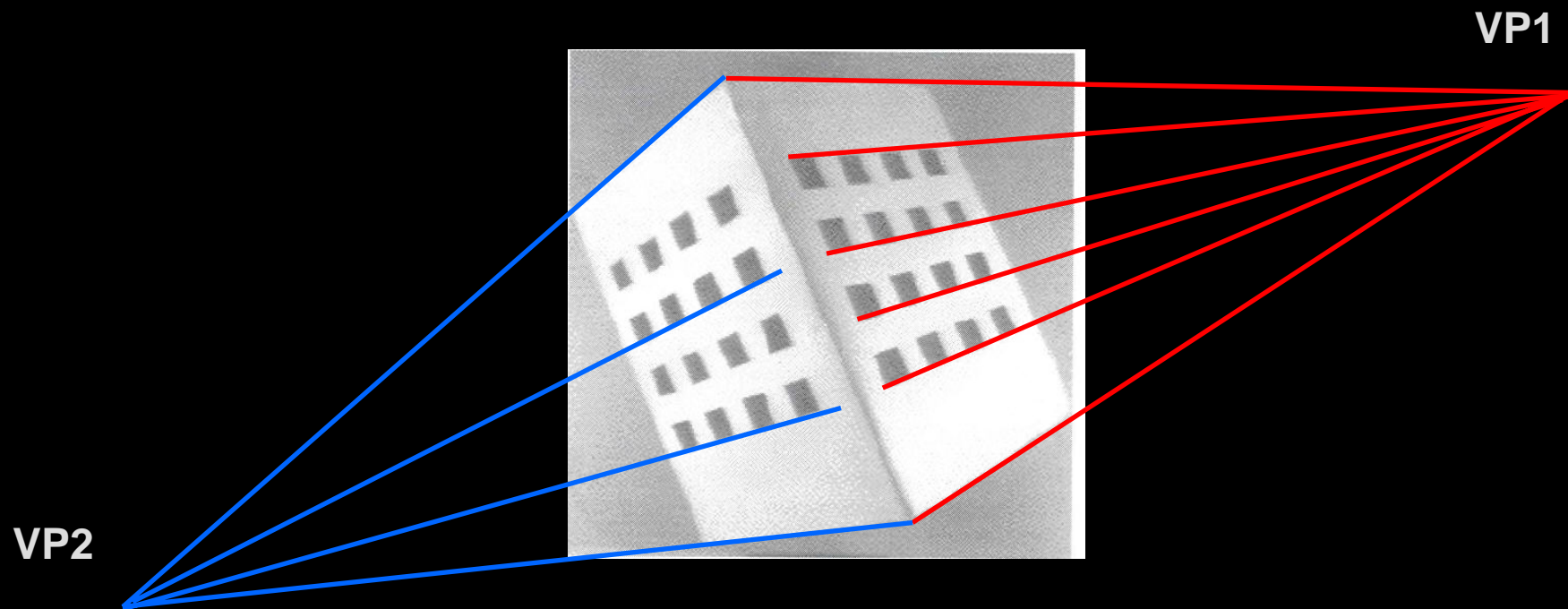
VP1

- Vanishing points:
 - Due to perspective, all parallel lines in 3D space appear to meet in a point on the image - the vanishing point, which is the common intersection of all the image lines
- Important property:
 - **Vector OV (from the center of projection to the vanishing point) is parallel to the parallel lines**



- Vanishing points:

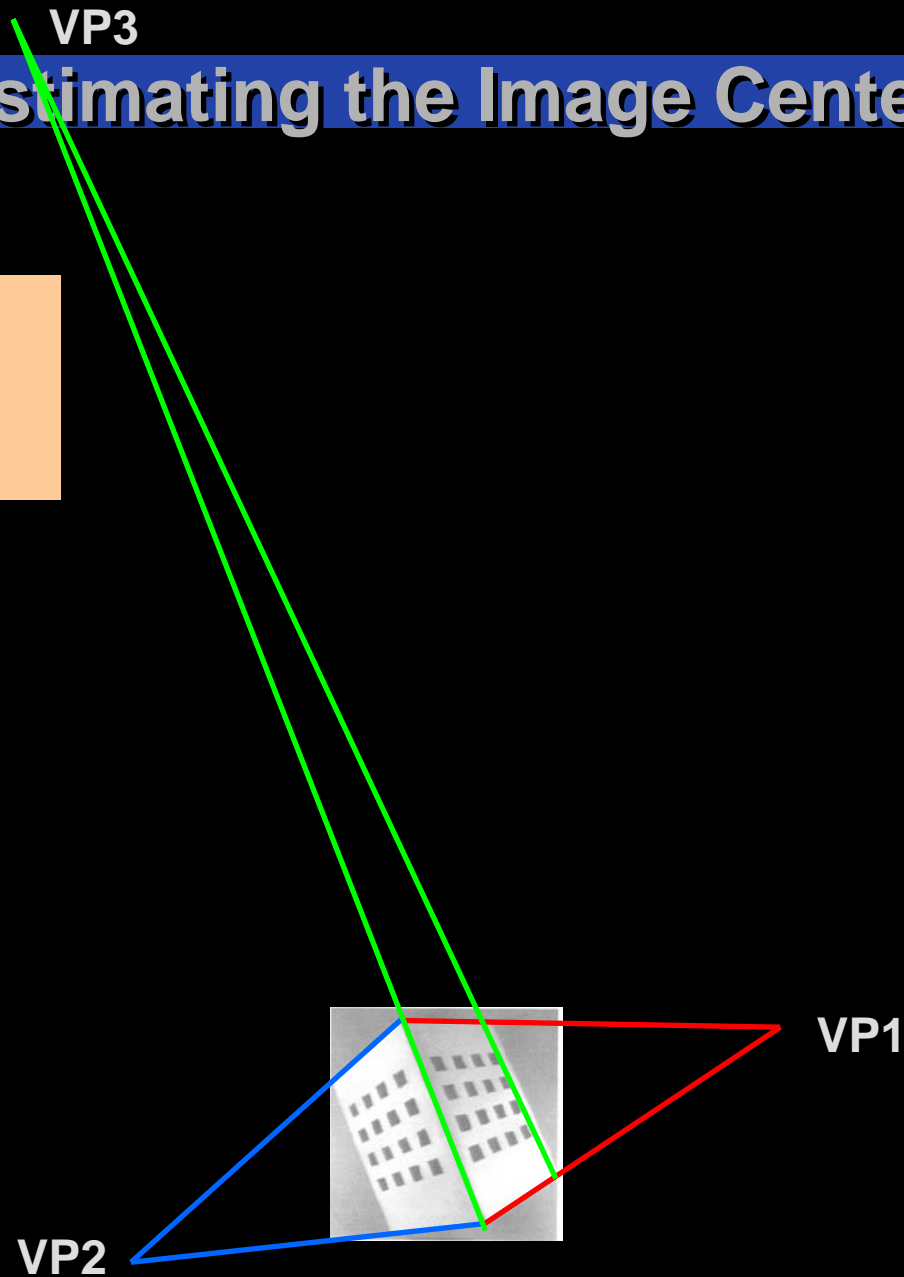
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Estimating the Image Center

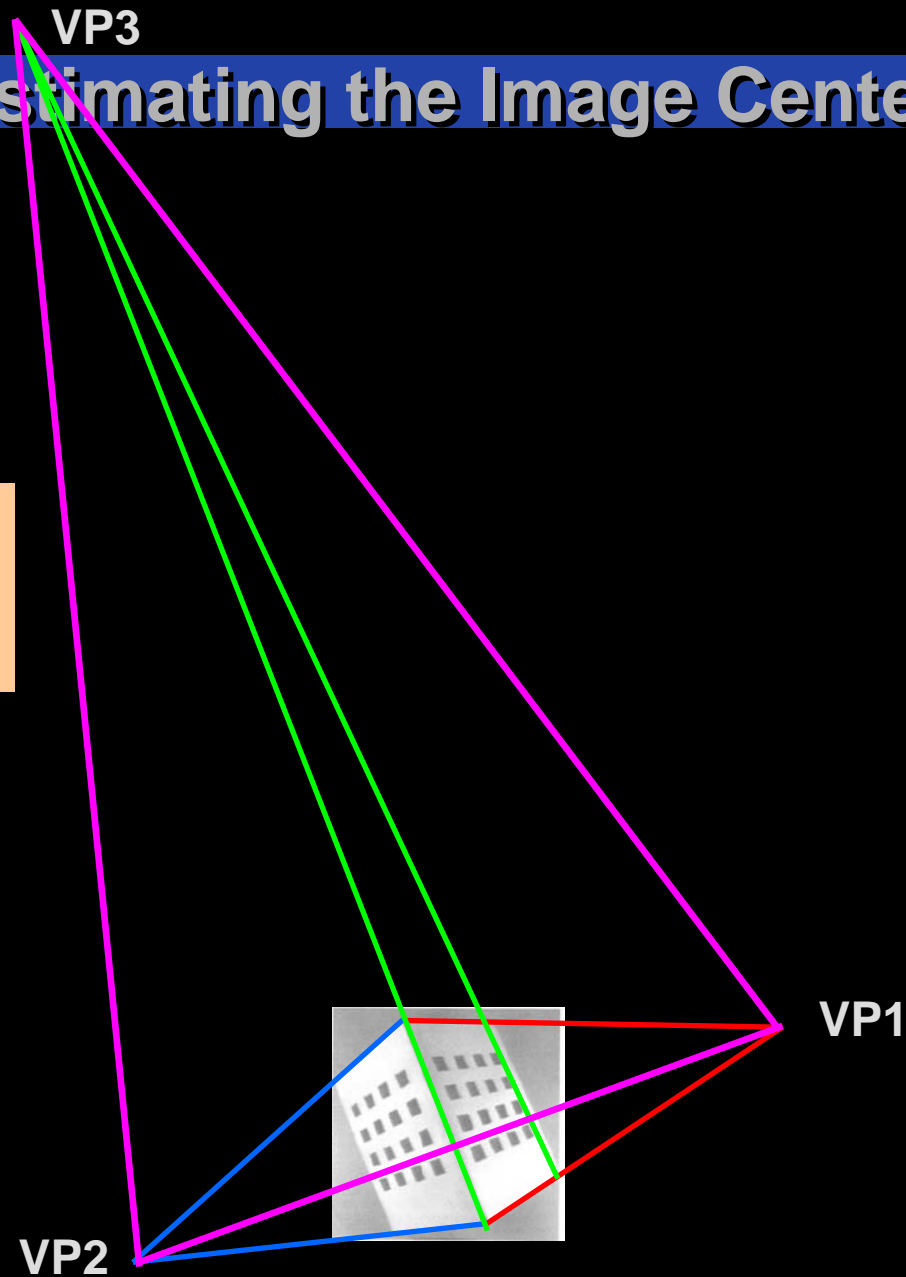
■ Orthocenter Theorem:

- Input: three mutually orthogonal sets of parallel lines in an image
- T: a triangle on the image plane defined by the three vanishing points
- Image center = orthocenter of triangle T
- Orthocenter of a triangle is the common intersection of the three altitudes



Estimating the Image Center

- Orthocenter Theorem:
 - Input: three mutually orthogonal sets of parallel lines in an image
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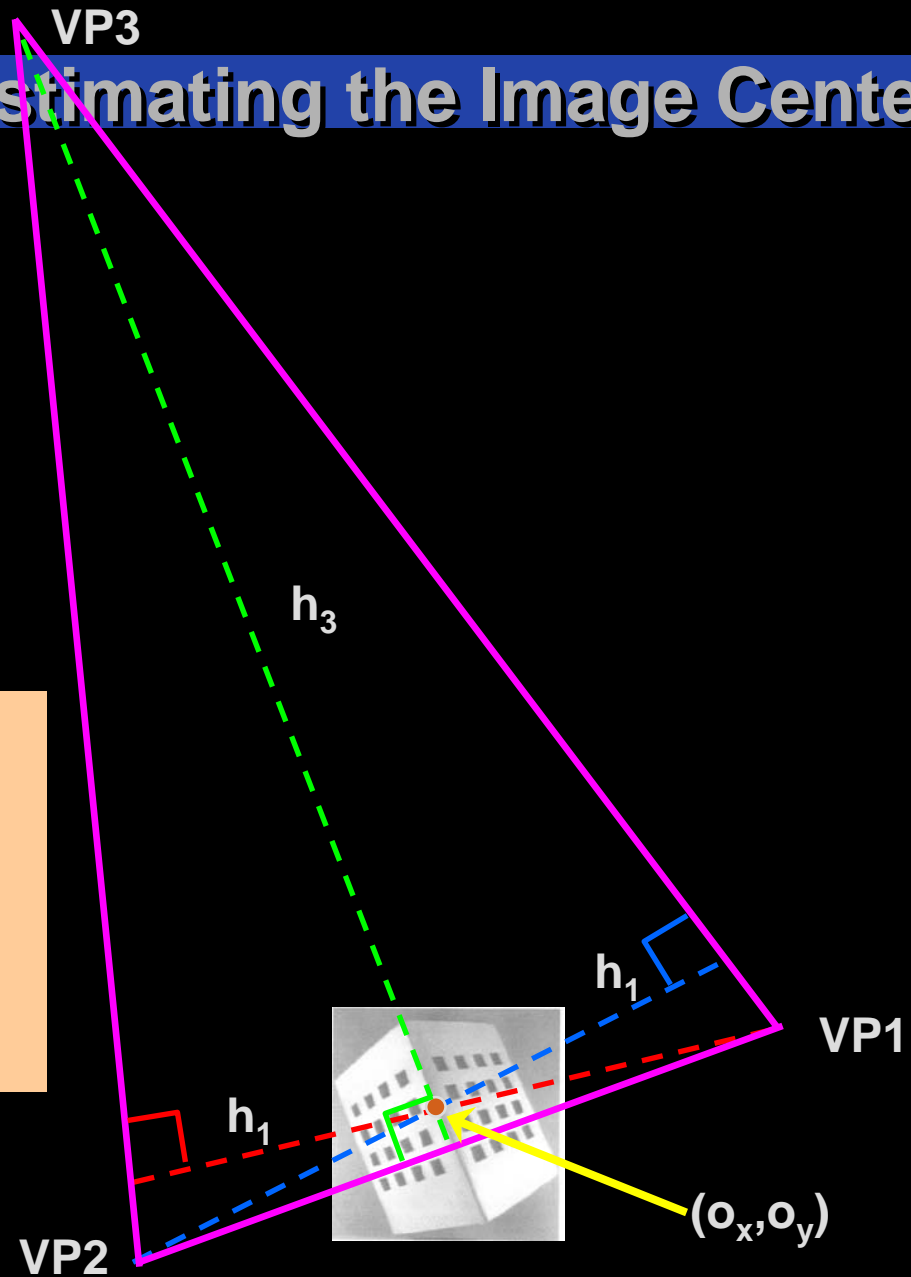


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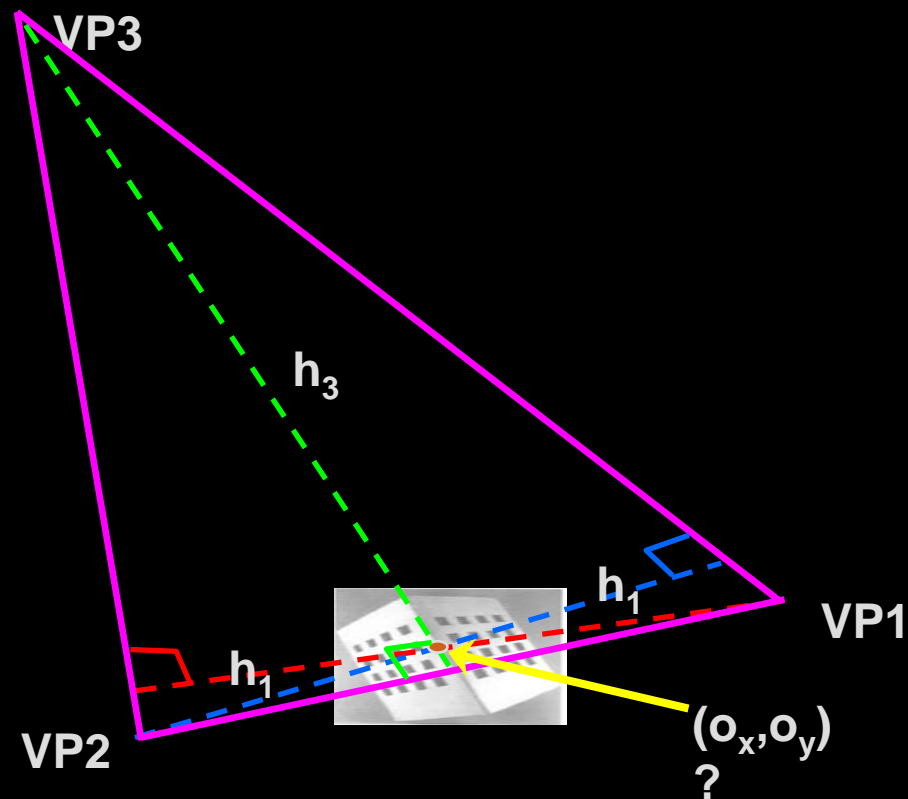
- Image center = orthocenter of triangle T
- Orthocenter of a triangle is the common intersection of the three altitudes

- Orthocenter Theorem:
 - WHY?





- Assumptions:
 - Known aspect ratio
 - Without lens distortions
- Questions:
 - Can we solve both aspect ratio and the image center?
 - How about with lens distortions?

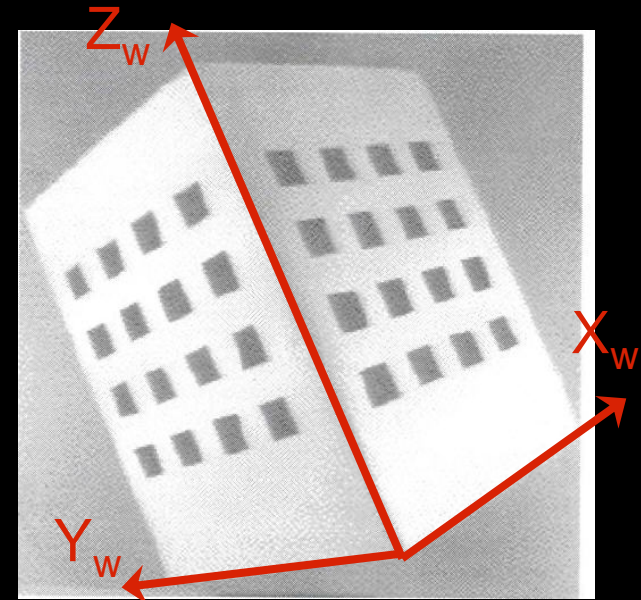


Direct parameter Calibration Summary

Algorithm (p130-131)

0. Estimate image center (and aspect ratio)

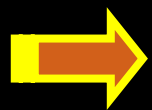
1. Measure N 3D coordinates (X_i, Y_i, Z_i)
2. Locate their corresponding image (x_i, y_i) - Edge, Corner, Hough
3. Build matrix A of a homogeneous system $Av = 0$
4. Compute **SVD** of A , solution v
5. Determine aspect ratio α and scale $|\gamma|$
6. Recover the first two rows of R and the first two components of T up to a sign
7. Determine sign s of γ by checking the projection equation
8. Compute the 3rd row of R by vector product, and enforce orthogonality constraint by **SVD**
9. Solve Tz and fx using Least Square and **SVD**, then $fy = fx / \alpha$



- Original assumptions:
 - Without lens distortions
 - Known aspect ratio when estimating image center
 - Known image center when estimating others including aspect ratio

- New Assumptions
 - Without lens distortion
 - Aspect ratio is approximately 1, or $\alpha = f_x/f_y = 4:3$; image center about $(M/2, N/2)$ given a $M \times N$ image

- Solution (?)
 1. Using $\alpha = 1$ to find image center (o_x, o_y)
 2. Using the estimated center to find others including α
 3. Refine image center using new α ; if change still significant, go to step 2; otherwise stop



Projection Matrix Approach

Linear Matrix Equation of perspective projection

Projective Space

- Add fourth coordinate
 - $P_w = (X_w, Y_w, Z_w, 1)^T$
- Define $(u, v, w)^T$ such that
 - $u/w = x_{im}, v/w = y_{im}$

$$\begin{pmatrix} x_{im} \\ y_{im} \end{pmatrix} = \begin{pmatrix} u/w \\ v/w \end{pmatrix}$$



$$\begin{pmatrix} u \\ v \\ w \end{pmatrix} = \mathbf{M}_{int} \mathbf{M}_{ext} \begin{pmatrix} X_w \\ Y_w \\ Z_w \\ 1 \end{pmatrix}$$

3x4 Matrix \mathbf{E}_{ext}

- Only extrinsic parameters
- World to camera

$$\mathbf{M}_{ext} = \begin{bmatrix} r_{11} & r_{12} & r_{13} & T_x \\ r_{21} & r_{22} & r_{23} & T_y \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix} = \begin{bmatrix} \mathbf{R}_1^T & T_x \\ \mathbf{R}_2^T & T_y \\ \mathbf{R}_3^T & T_z \end{bmatrix}$$

3x3 Matrix \mathbf{E}_{int}

- Only intrinsic parameters
- Camera to frame

$$\mathbf{M}_{int} = \begin{bmatrix} -f_x & 0 & o_x \\ 0 & -f_y & o_y \\ 0 & 0 & 1 \end{bmatrix}$$

Simple Matrix Product! Projective Matrix $\mathbf{M} = \mathbf{M}_{int} \mathbf{M}_{ext}$

- $(X_w, Y_w, Z_w)^T \rightarrow (x_{im}, y_{im})^T$
- Linear Transform from projective space to projective plane
- \mathbf{M} defined up to a scale factor – 11 independent entries

- World – Frame Transform
 - Drop “im” and “w”
 - N pairs $(x_i, y_i) \leftrightarrow (X_i, Y_i, Z_i)$
 - Linear equations of m

$$\begin{pmatrix} u_i \\ v_i \\ w_i \end{pmatrix} = \mathbf{M} \begin{pmatrix} X_i \\ Y_i \\ Z_i \\ 1 \end{pmatrix}$$

$$\mathbf{A}m = \mathbf{0}$$

$$x_i = \frac{u_i}{w_i} = \frac{m_{11}X_i + m_{12}Y_i + m_{13}Z_i + m_{14}}{m_{31}X_i + m_{32}Y_i + m_{33}Z_i + m_{34}}$$

$$y_i = \frac{v_i}{w_i} = \frac{m_{21}X_i + m_{22}Y_i + m_{23}Z_i + m_{24}}{m_{31}X_i + m_{32}Y_i + m_{33}Z_i + m_{34}}$$

- 3x4 Projection Matrix \mathbf{M}

- Both intrinsic (4) and extrinsic (6) – 10 parameters

$$\mathbf{M} = \begin{bmatrix} -f_x r_{11} + o_x r_{31} & -f_x r_{12} + o_x r_{32} & -f_x r_{13} + o_x r_{33} & -f_x T_x + o_x T_z \\ -f_y r_{21} + o_y r_{31} & -f_y r_{22} + o_y r_{32} & -f_y r_{23} + o_y r_{33} & -f_y T_y + o_y T_z \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix}$$

- World – Frame Transform
 - Drop “im” and “w”
 - N pairs $(x_i, y_i) \leftrightarrow (X_i, Y_i, Z_i)$

$$x_i = \frac{u_i}{w_i} = \frac{m_{11}X_i + m_{12}Y_i + m_{13}Z_i + m_{14}}{m_{31}X_i + m_{32}Y_i + m_{33}Z_i + m_{34}}$$

$$y_i = \frac{v_i}{w_i} = \frac{m_{21}X_i + m_{22}Y_i + m_{23}Z_i + m_{24}}{m_{31}X_i + m_{32}Y_i + m_{33}Z_i + m_{34}}$$

- Linear equations of m
 - 2N equations, 11 independent variables
 - $N \geq 6$, SVD \Rightarrow m up to a unknown scale

$$\mathbf{A}\mathbf{m} = \mathbf{0}$$

$$\mathbf{A} = \begin{bmatrix} X_1 & Y_1 & Z_1 & 1 & 0 & 0 & 0 & 0 & -x_1X_1 & -x_1Y_1 & -x_1Z_1 & -x_1 \\ 0 & 0 & 0 & 0 & X_1 & Y_1 & Z_1 & 1 & -y_1X_1 & -y_1Y_1 & -y_1Z_1 & -y_1 \\ \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot & \cdot \end{bmatrix}$$

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

Step 2: Computing camera parameters

- 3x4 Projection Matrix \mathbf{M}
 - Both intrinsic and extrinsic

$$\hat{\mathbf{M}} = \begin{bmatrix} \mathbf{q}_1 & q_{41} \\ \mathbf{q}_2 & q_{42} \\ \mathbf{q}_3 & q_{43} \end{bmatrix}$$

$$\mathbf{M} = \begin{bmatrix} -f_x r_{11} + o_x r_{31} & -f_x r_{12} + o_x r_{32} & -f_x r_{13} + o_x r_{33} & -f_x T_x + o_x T_z \\ -f_y r_{21} + o_y r_{31} & -f_y r_{22} + o_y r_{32} & -f_y r_{23} + o_y r_{33} & -f_y T_y + o_y T_z \\ r_{31} & r_{32} & r_{33} & T_z \end{bmatrix}$$

- From $\hat{\mathbf{M}}$ to parameters (p134-135)
 - Find scale $|\gamma|$ by using unit vector \mathbf{R}_3^\top
 - Determine T_z and sign of γ from m_{34} (i.e. q_{43})
 - Obtain \mathbf{R}_3^\top
 - Find (O_x, O_y) by dot products of Rows $\mathbf{q}_1, \mathbf{q}_3, \mathbf{q}_2, \mathbf{q}_3$, using the orthogonal constraints of \mathbf{R}
 - Determine f_x and f_y from \mathbf{q}_1 and \mathbf{q}_2 (Eq. 6.19) Wrong???)
 - All the remainings: $\mathbf{R}_1^\top, \mathbf{R}_2^\top, T_x, T_y$
 - Enforce orthogonality on \mathbf{R} ?

$$\hat{\mathbf{M}} = \gamma \mathbf{M}$$

- Direct parameter method and Projection Matrix method
- Properties in Common:
 - Linear system first, Parameter decomposition second
 - Results should be exactly the same
- Differences
 - Number of variables in homogeneous systems
 - Matrix method: All parameters at once, $2N$ Equations of 12 variables
 - Direct method in three steps: N Equations of 8 variables, N equations of 2 Variables, Image Center – **maybe more stable**
 - Assumptions
 - Matrix method: simpler, and more general; **sometime projection matrix is sufficient so no need for parameter decomposition**
 - Direct method: Assume known image center in the first two steps, and known aspect ratio in estimating image center

- Pick up a well-known technique or a few
- Design and construct calibration patterns (with known 3D)
- Make sure what parameters you want to find for your camera
- Run algorithms on **ideal** simulated data
 - You can either use the data of the real calibration pattern or using computer generated data
 - Define a virtual camera with known intrinsic and extrinsic parameters
 - Generate 2D points from the 3D data using the virtual camera
 - Run algorithms on the 2D-3D data set
- Add **noises** in the simulated data to test the robustness
- Run algorithms on the **real data** (images of calibration target)
- If successful, you are all set
- Otherwise:
 - Check how you select the **distribution** of control points
 - Check the **accuracy** in 3D and 2D localization
 - Check the **robustness** of your algorithms again
 - Develop your own algorithms → **NEW METHODS?**

- 3D reconstruction using two cameras

Stereo Vision

- Homework #3 online, Due April 13 before midnight