



Chapter 1

PRINCIPLES OF PHOTOGRAMMETRIC MAPPING

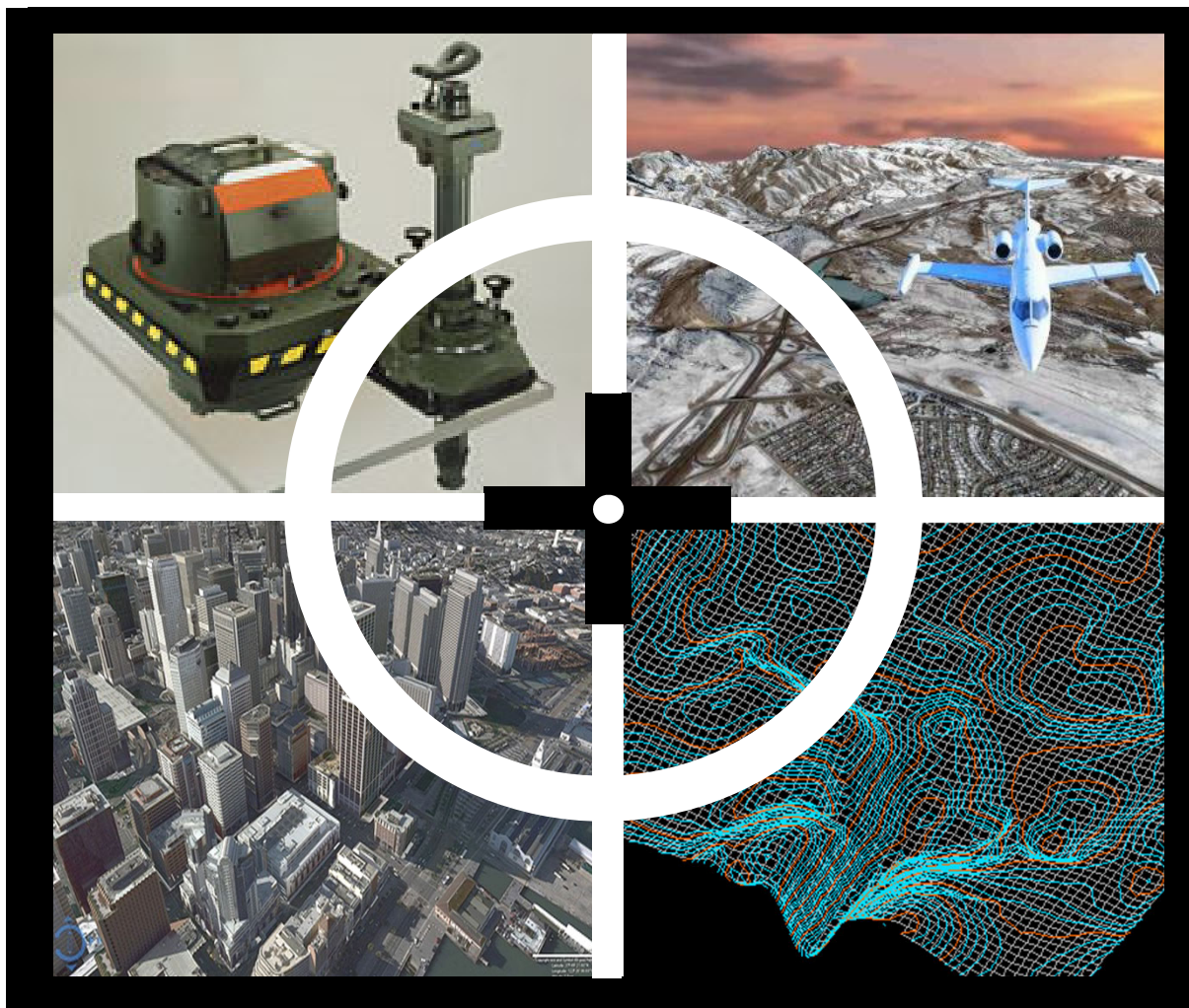


Overview

- Photogrammetry: Definition and applications
- Photogrammetric tools:
 - Rotation matrices
 - Photogrammetric orientation: interior and exterior orientation
- Photogrammetric point positioning
 - Collinearity equations/conditions (single camera systems)
 - GNSS/INS-assisted photogrammetric systems
 - Multi-camera photogrammetric systems
- Photogrammetric bundle adjustment
 - Structure of the design and normal matrices

Photogrammetry

- Objective: Derive the positions and shapes of objects from imagery





Photogrammetry

- **Classical Definitions:**
 - The art and science of determining the position and shape of objects from photography
 - The process of reconstructing objects without touching them
 - Non-contact positioning method
- **Contemporary Definition:**
 - The art and science of tool development for automatic generation of spatial and descriptive information from multi-sensory data and/or systems

Data Acquisition Systems



Traditional Mapping Cameras

Large Format Imaging Systems



Low-Cost Digital Cameras



Medium and Small Format Digital Imaging Systems

Data Acquisition Systems



Traditional mapping cameras

- ↑ accurate lab calibration
- ↑ large image format
- ↑ low distortion lens system
- ↑ stable IOP
- ↑ extremely-high geometric image quality
- ↓ high initial procurement cost
- ↓ not easy to integrate with other systems on the same platform (e.g., LiDAR)

Medium-format digital cameras

- ↑ low-cost/off-the-shelf
- ↑ easy to integrate with other systems on the same platform (e.g., LiDAR)
- ↑ convenient for small area coverage & UAV systems
- ↓ should be calibrated by the end user
- ↓ inferior geometric quality and lens system
- ↓ stability of IOP is not guaranteed
- ↓ limited array size

Data Acquisition Systems



WILD RC10

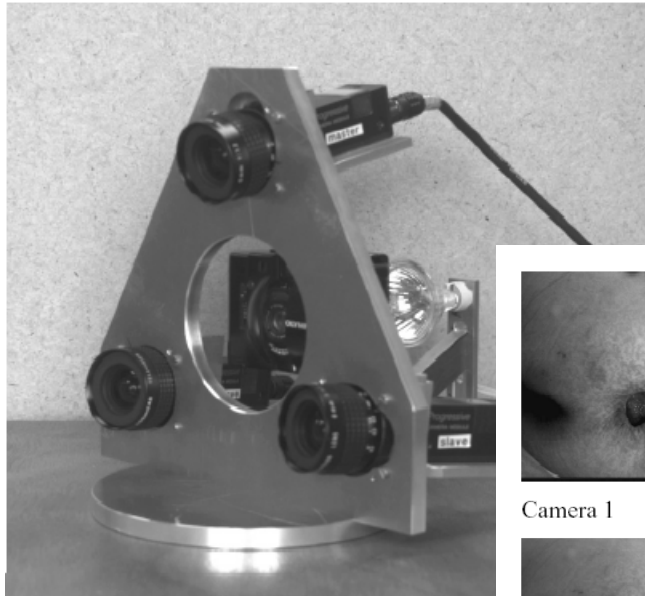
Laser Scanning

Data Acquisition Systems



SONY DSC F717

Terrestrial (Close Range) Imagery



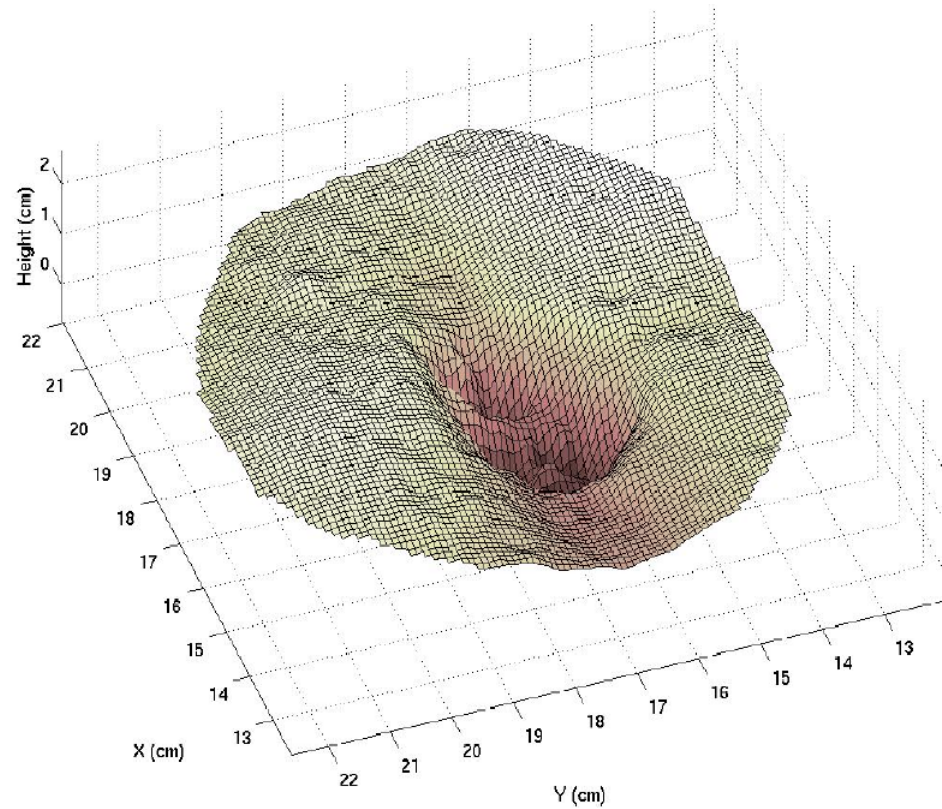
Camera 1



Camera 2



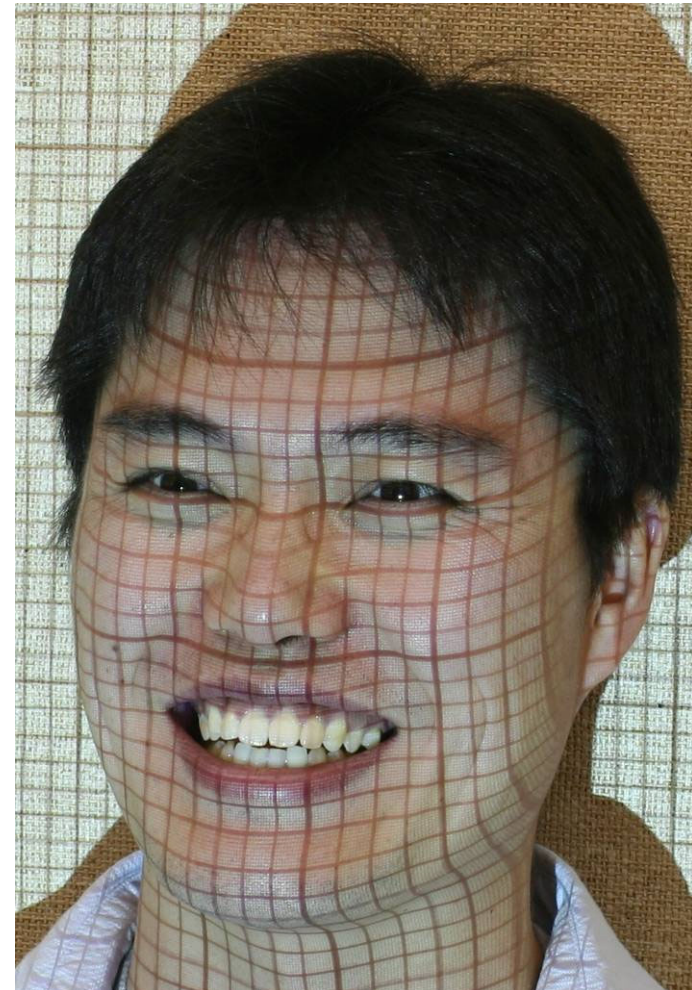
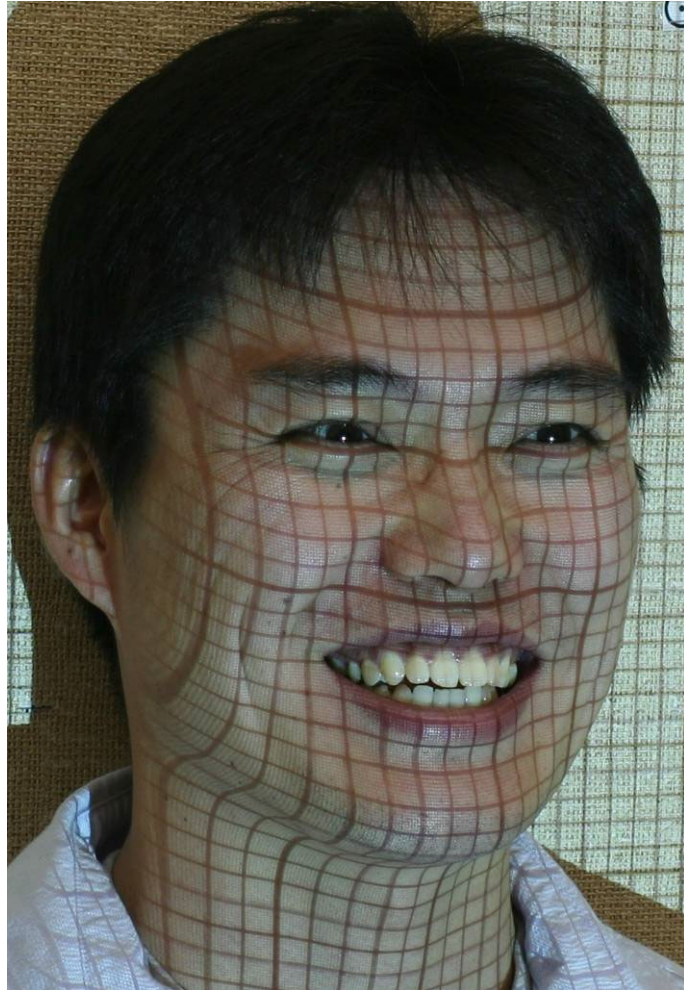
Camera 3



Terrestrial (Close Range) Imagery

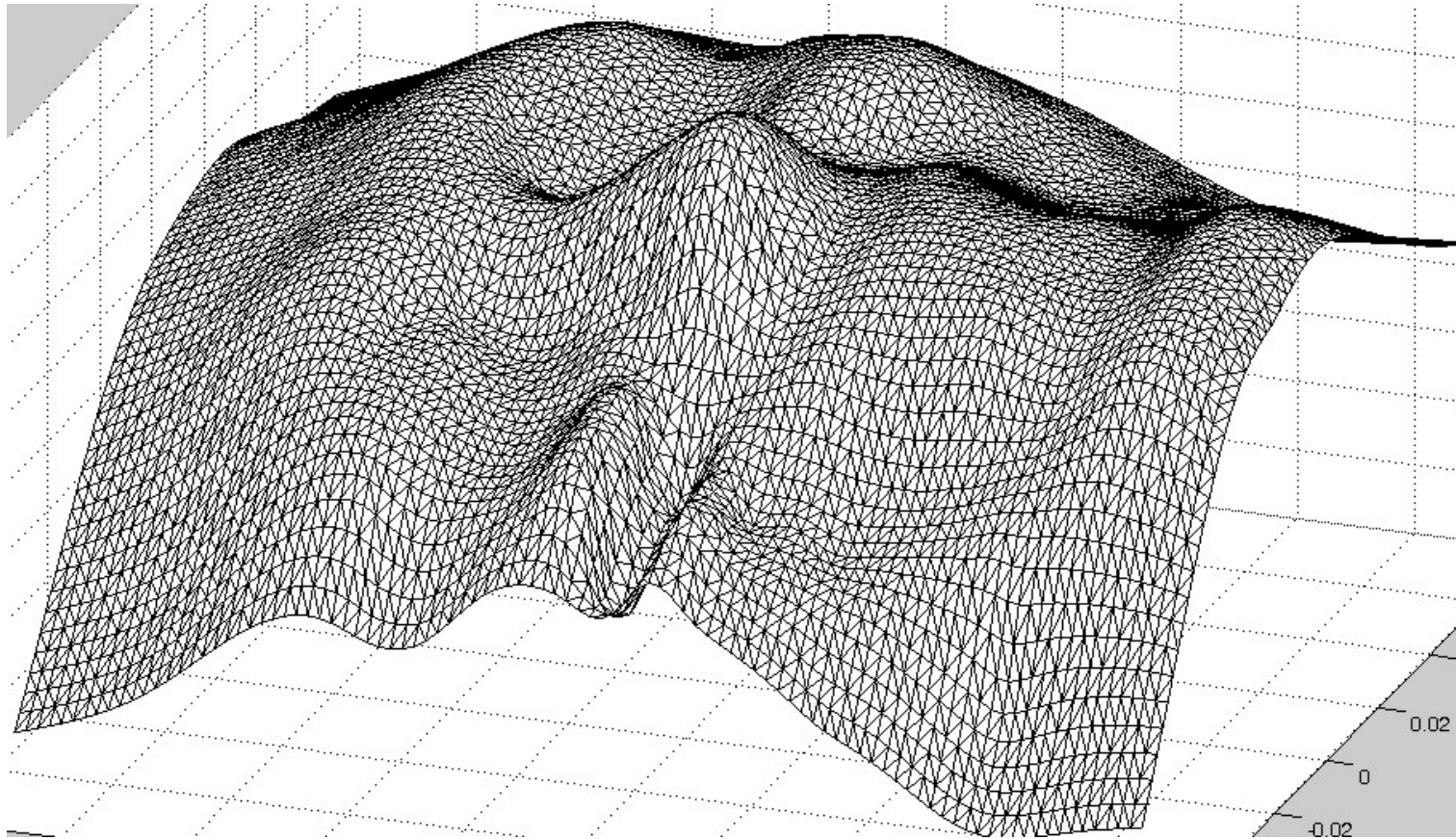


Terrestrial (Close Range) Imagery



Input Stereo-Imagery

Terrestrial (Close Range) Imagery



Output Three-Dimensional Model

Terrestrial (Close Range) Imagery



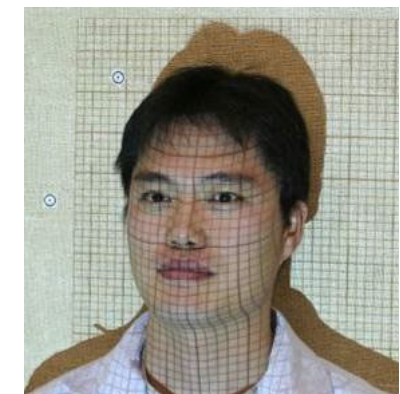
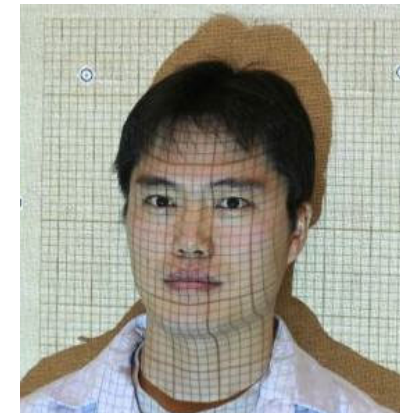
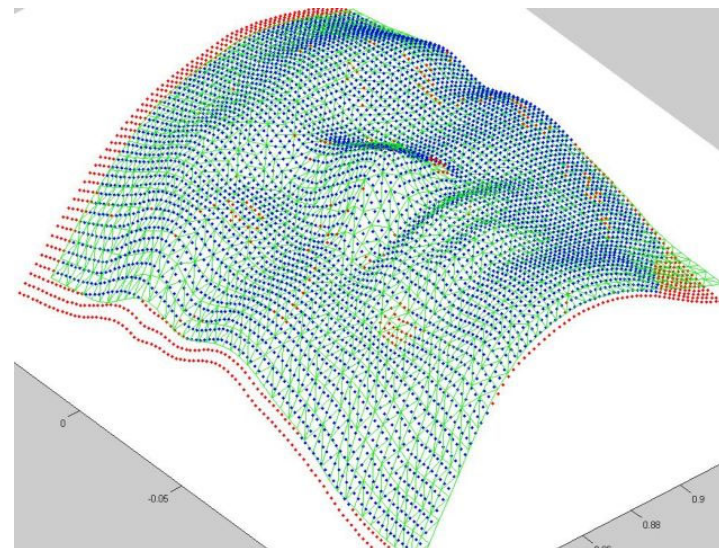
- Experiments

Test	Descriptions
1	Subject 1: Time 1 & Time 2
2	Subject 1: No Smile & Smile
3	Subject 2 & Subject 3

- Results:

Test 1

Green: Reference
Blue: Matches
Red: Non-matches



Terrestrial (Close Range) Imagery



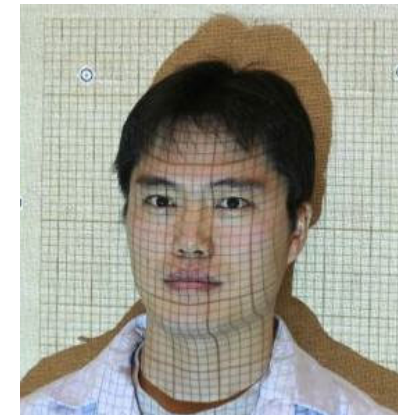
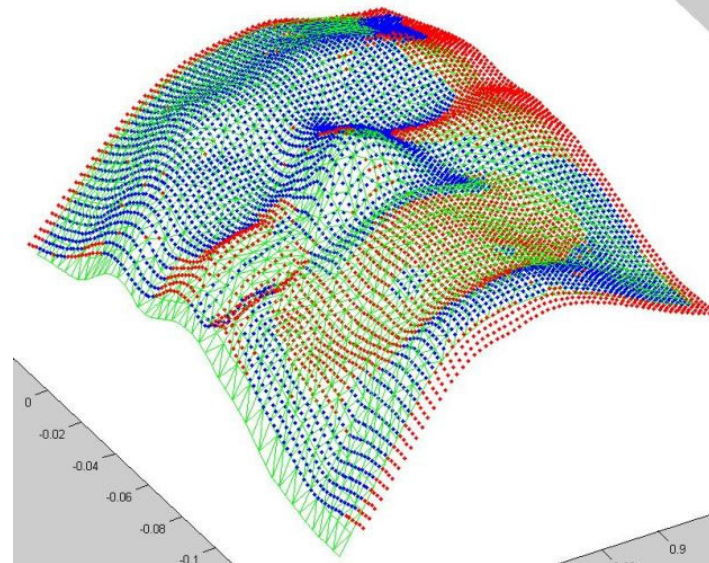
- Experiments

Test	Descriptions
1	Subject 1: Time 1 & Time 2
2	Subject 1: No Smile & Smile
3	Subject 2 & Subject 3

- Results:

Test 2

Green: Reference
Blue: Matches
Red: Non-matches



Terrestrial (Close Range) Imagery



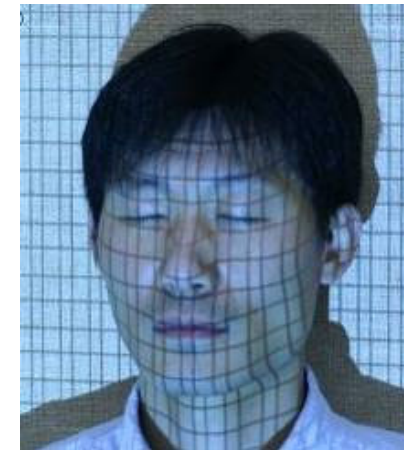
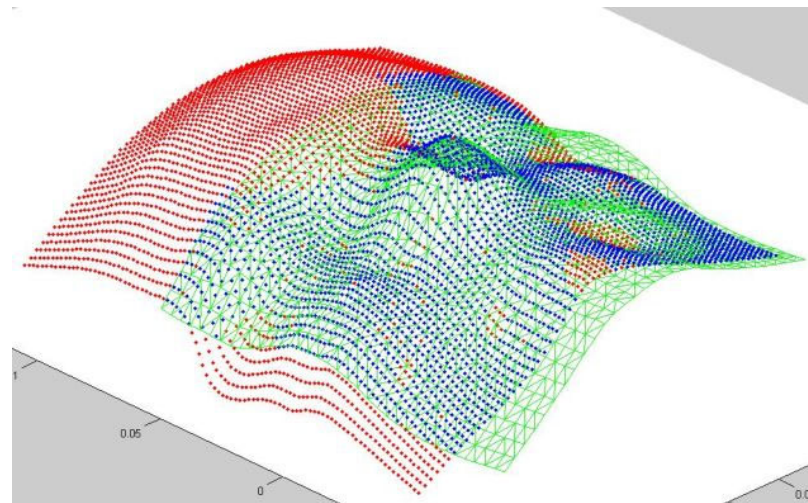
- Experiments

Test	Descriptions
1	Subject 1: Time 1 & Time 2
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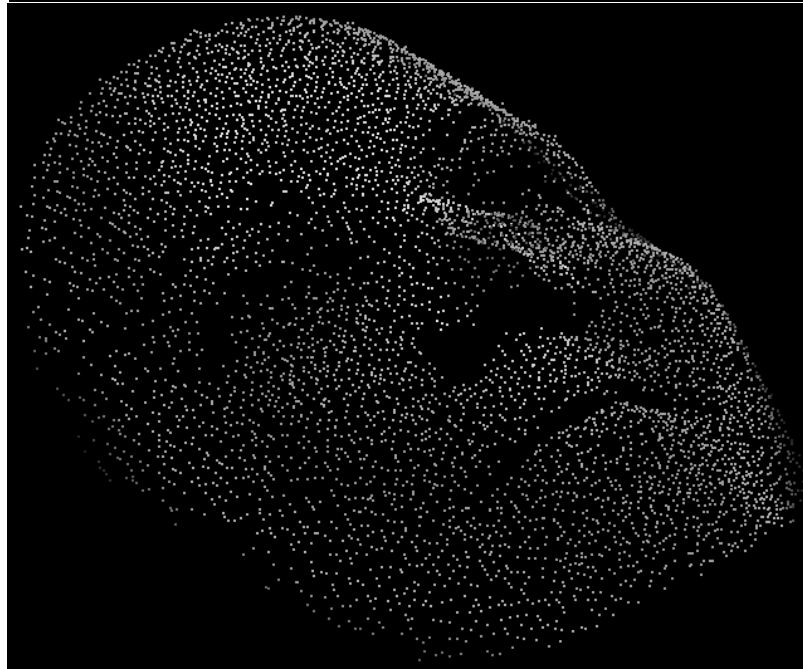
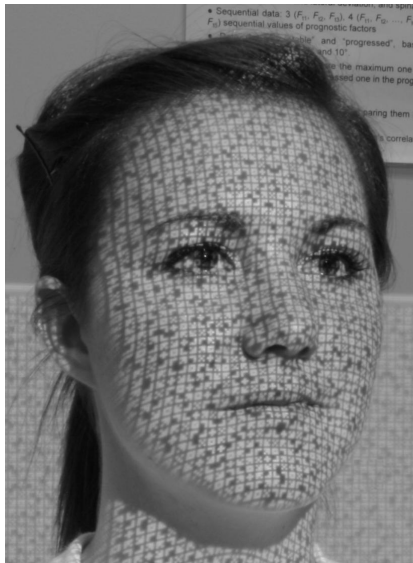
- Results:

Test 3

Green: Reference
Blue: Matches
Red: Non-matches



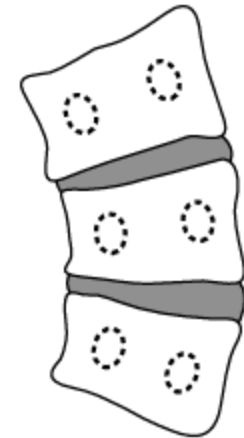
Terrestrial (Close Range) Imagery



Terrestrial (Close Range) Imagery



- **Scoliosis**
- 3D deformity of the human spine
- Affects 2-3% of the population
- Impacts the quality of life
- Early detection is vital



www.rad.washington.edu/mskbook/scoliosis.html

Signs of scoliosis



ADAM.

www.nlm.nih.gov/MEDLINEPLUS/ency/images/ency/fullsize/19466.jpg

Terrestrial (Close Range) Imagery



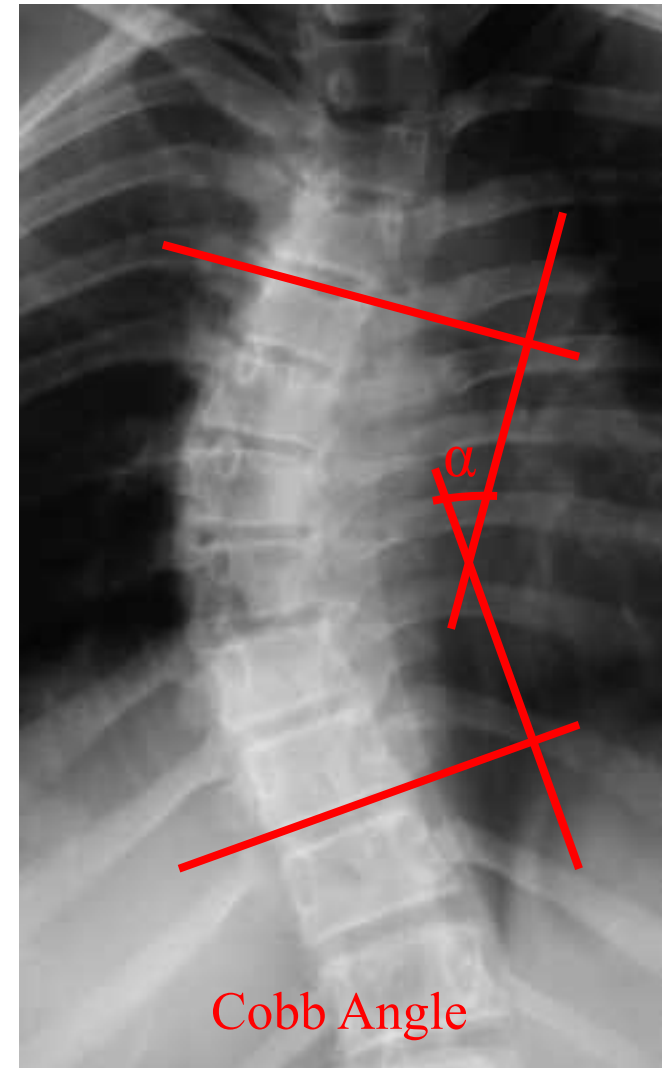
- **Scoliosis Detection & Monitoring**

- Traditional method:

- Full-length spinal x-ray in a standing position

- Consequences:

- Frequent exposure to radiation (4-5 times a year, for 3-5 years)
- Increased risk of cancer

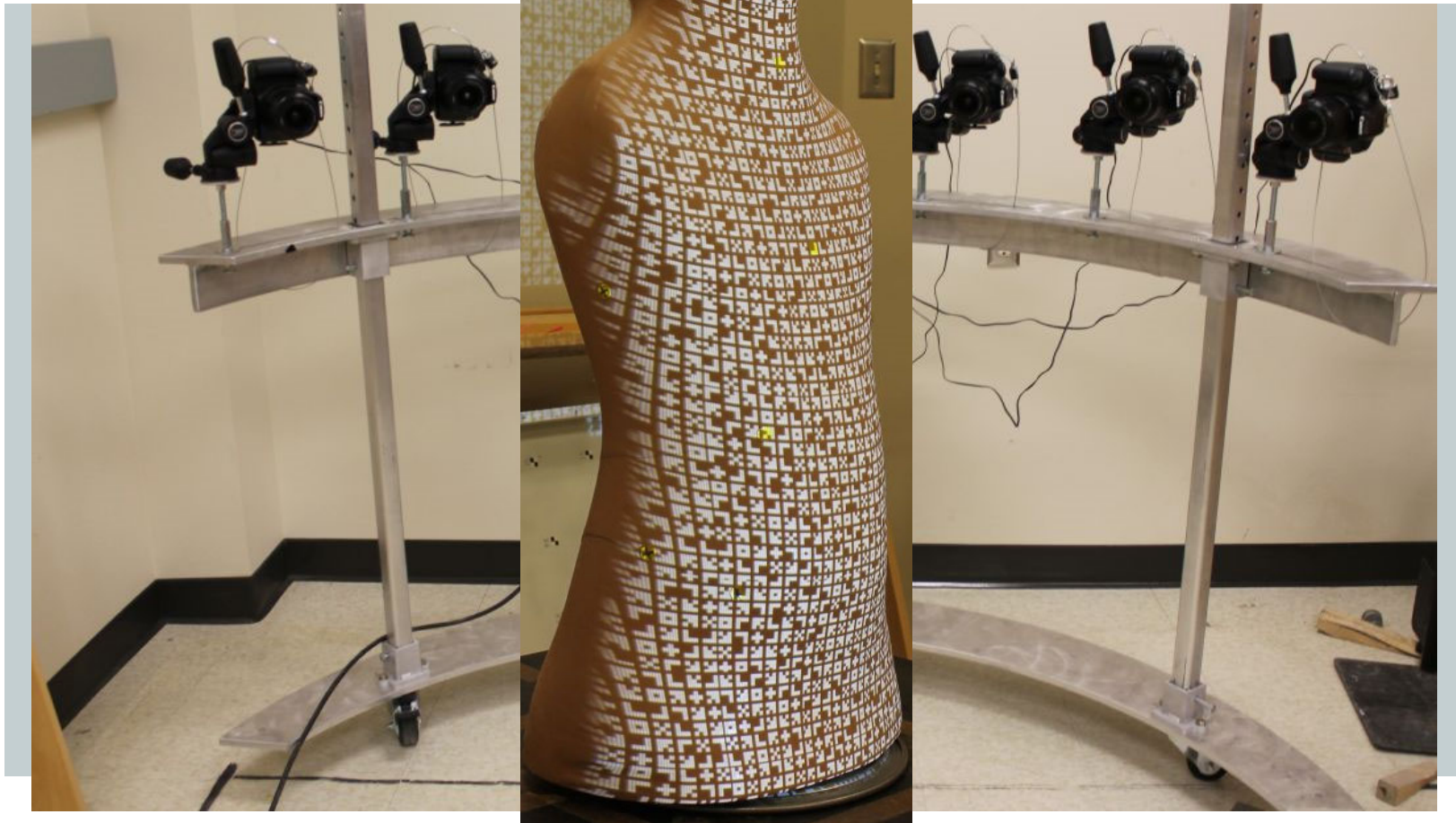


<http://www.e-radiography.net/radpath/c/cobb-angle.jpg>

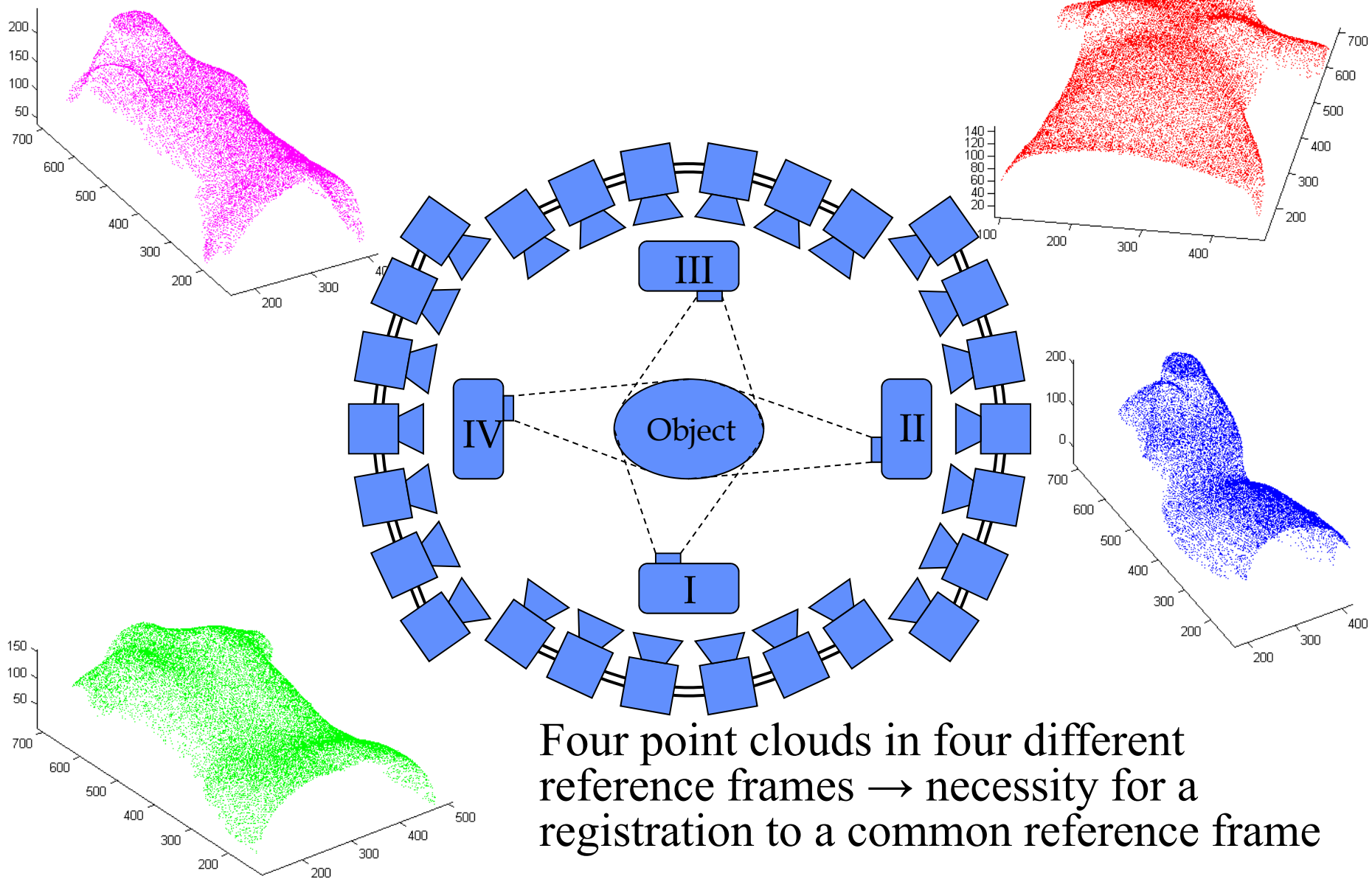
Terrestrial (Close Range) Imagery



Cameras, projectors, frame, target board, computer(s), remote control



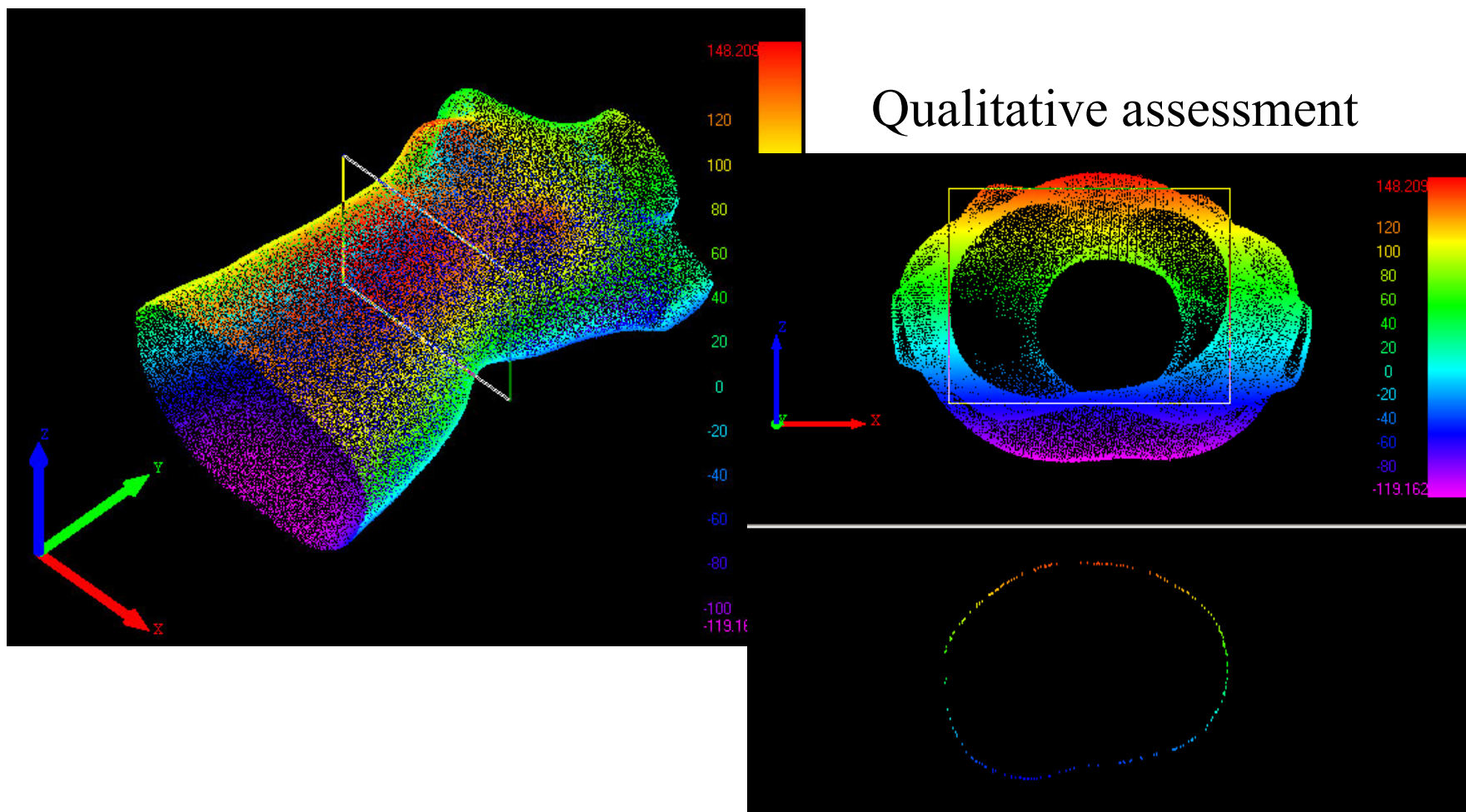
Terrestrial (Close Range) Imagery



Four point clouds in four different reference frames → necessity for a registration to a common reference frame

Terrestrial (Close Range) Imagery

- Multiple surface registration: complete 3D torso model



Terrestrial (Close Range) Imagery



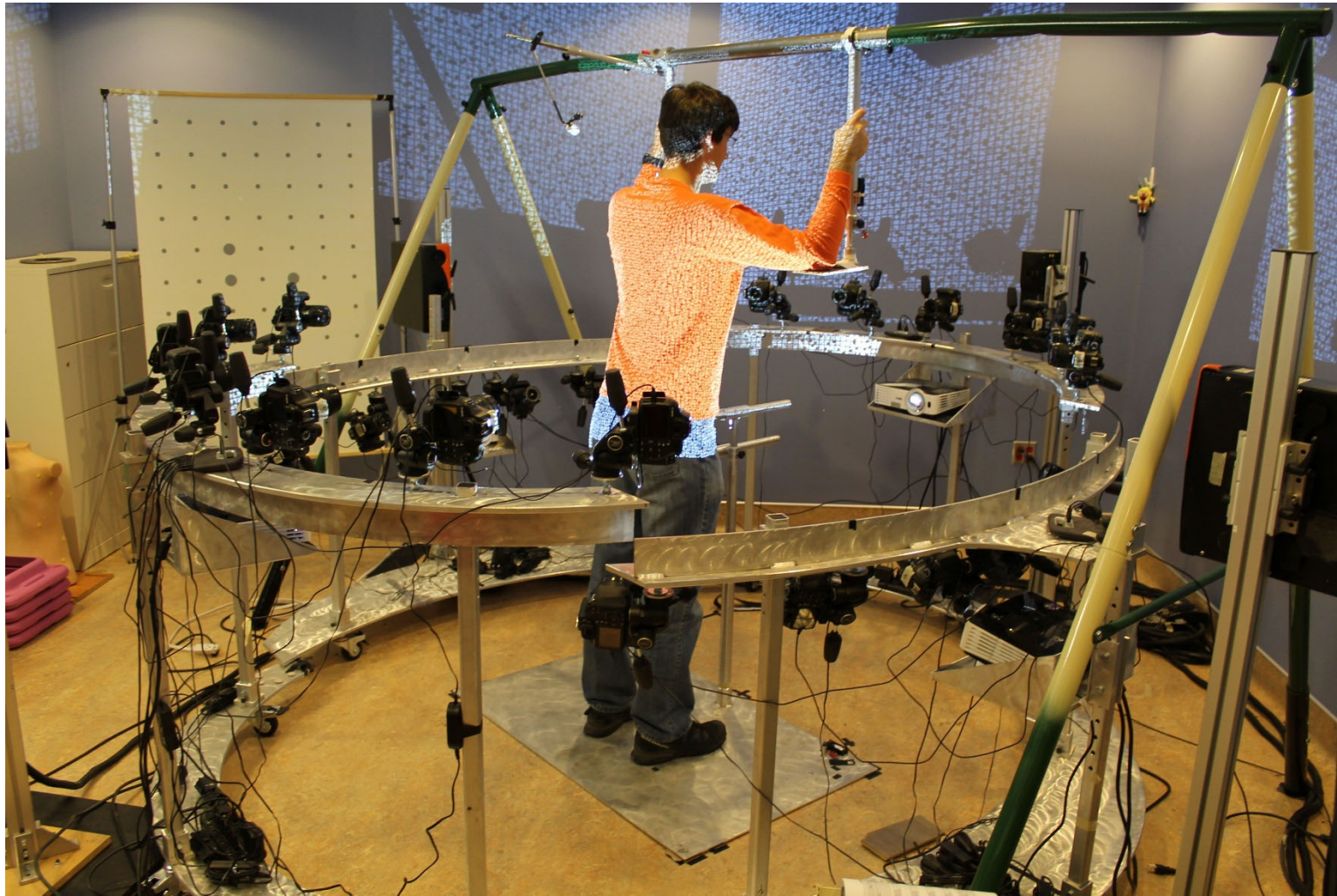
Terrestrial (Close Range) Imagery



Terrestrial (Close Range) Imagery



Terrestrial (Close Range) Imagery



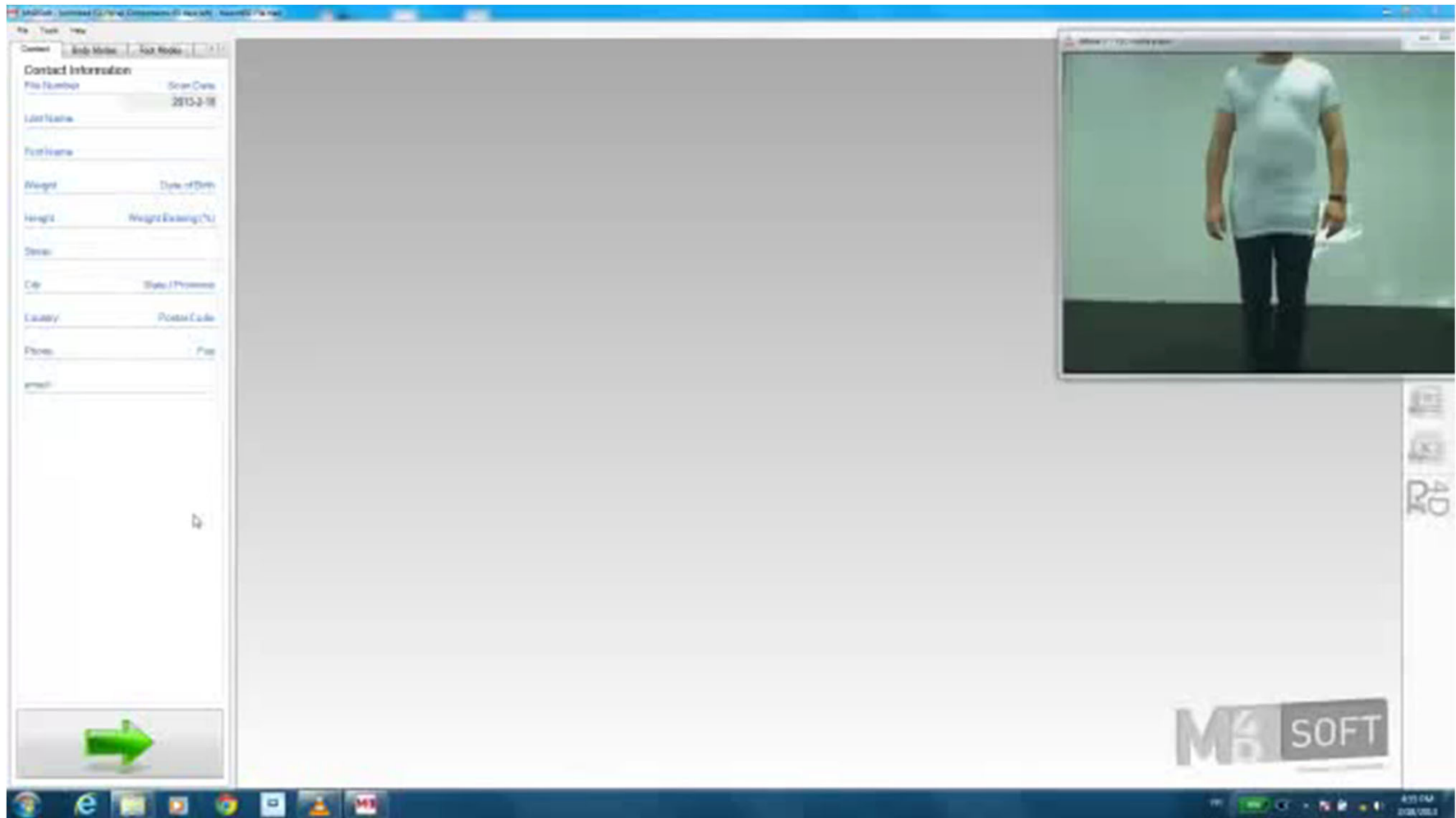
Terrestrial (Close Range) Imagery



Terrestrial (Close Range) Imagery



Laser-Based Torso Reconstruction



Terrestrial (Close Range) Imagery

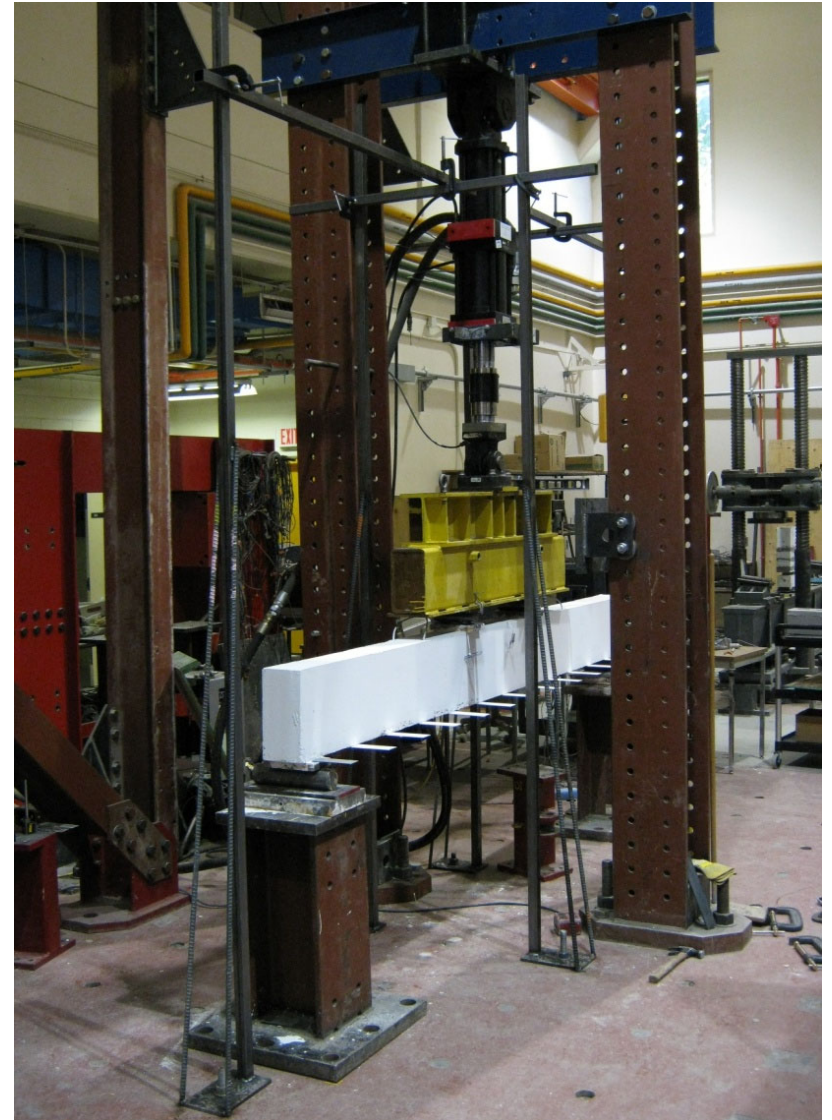


➤ Objective:

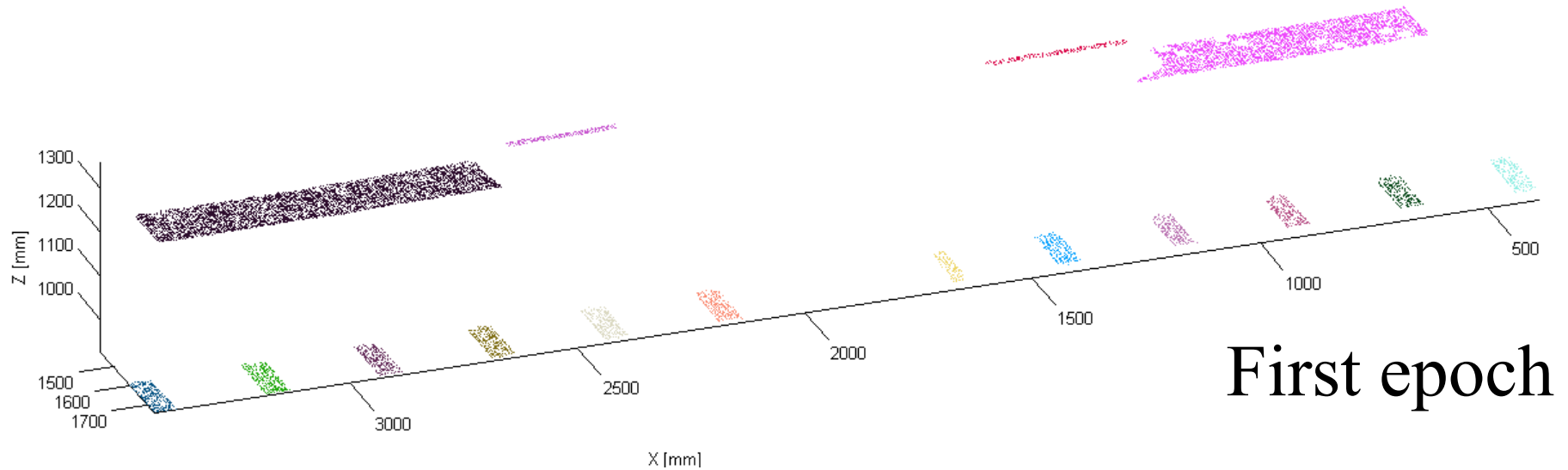
- Develop a system that can evaluate the deflection along the beam under static and dynamic loading conditions

➤ Design target function:

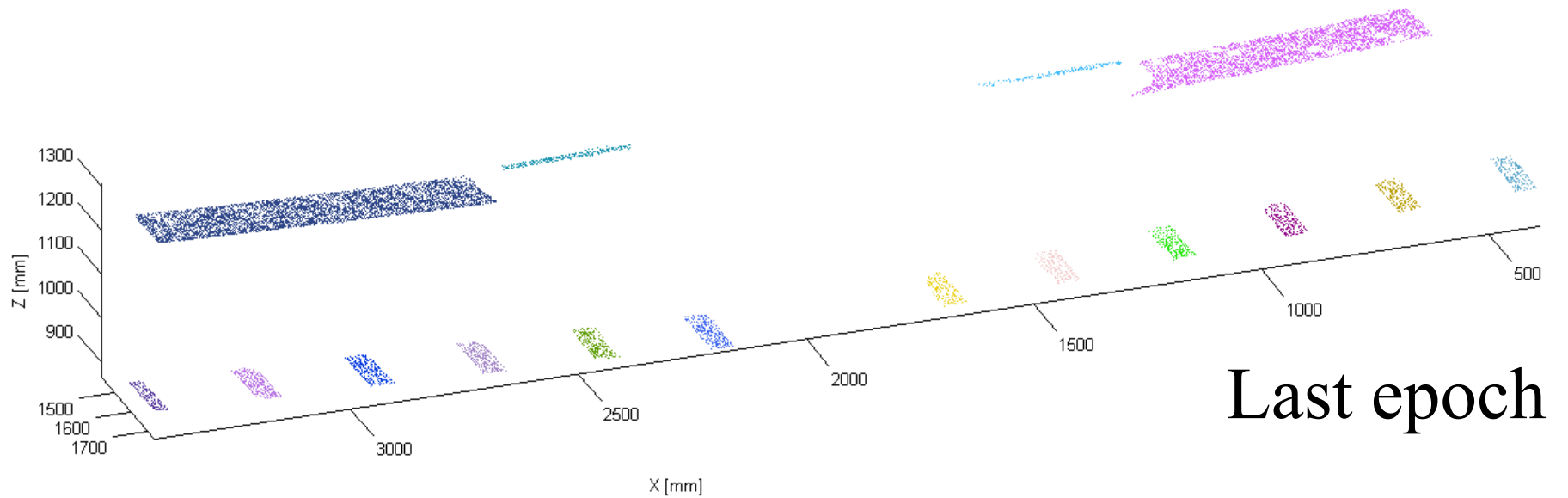
- Low cost
- Non-contact
- Accurate
- Reusable
- Continuous evaluation of the deflection along the beam



Terrestrial (Close Range) Imagery

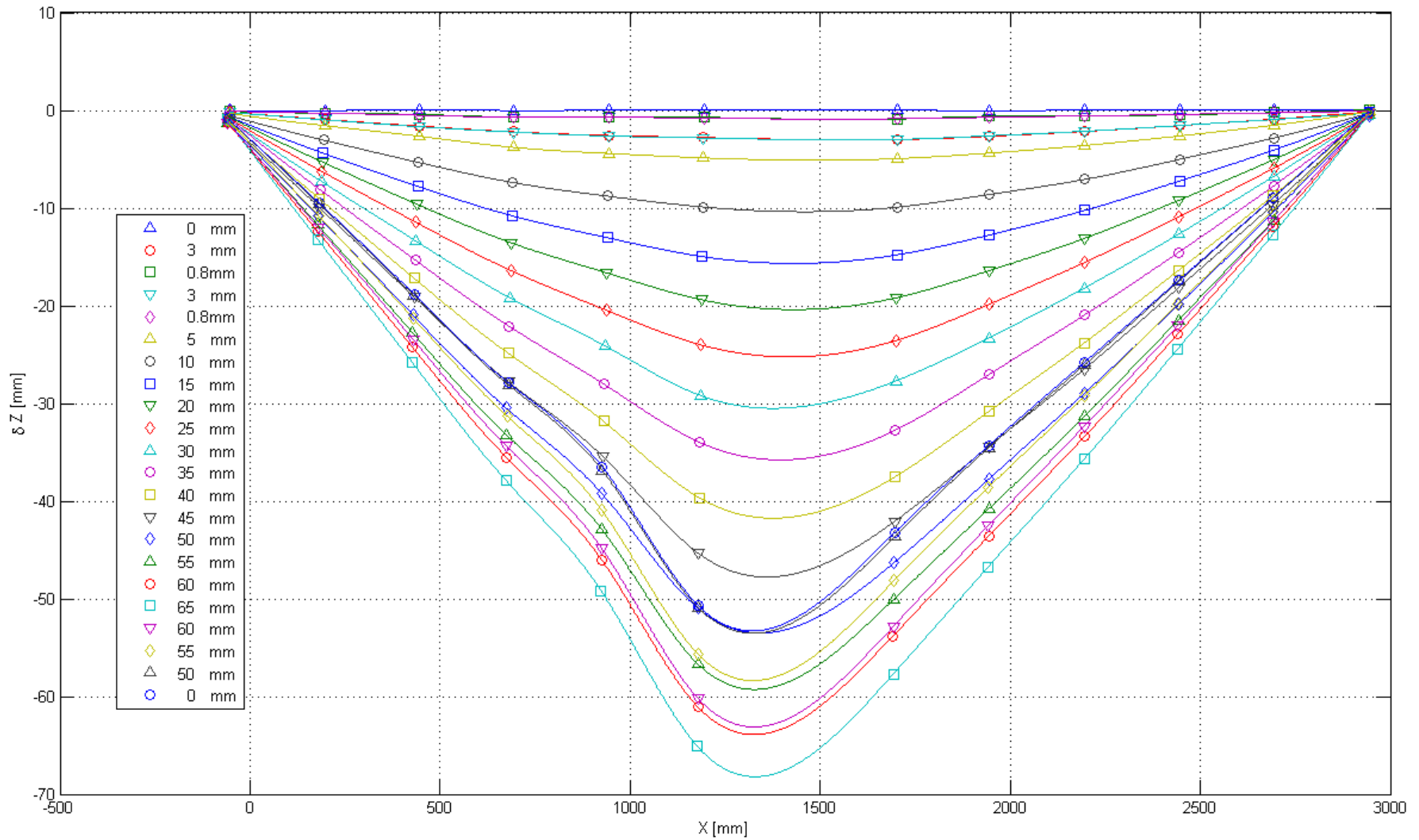


First epoch



Last epoch

Terrestrial (Close Range) Imagery



Terrestrial (Close Range) Imagery



Mobile Mapping Systems (MMS)



Terrestrial (Close Range) Imagery



Mobile Mapping Systems (MMS)



Terrestrial (Close Range) Imagery



Mobile Mapping Systems (MMS)



University of Calgary

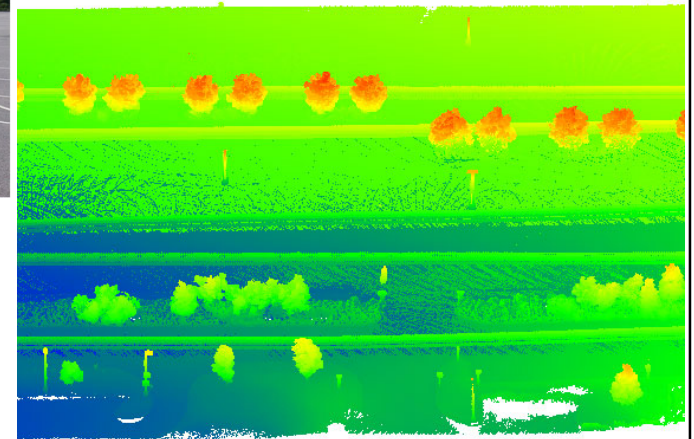
Terrestrial Mobile Mapping Systems



Platform: Truck



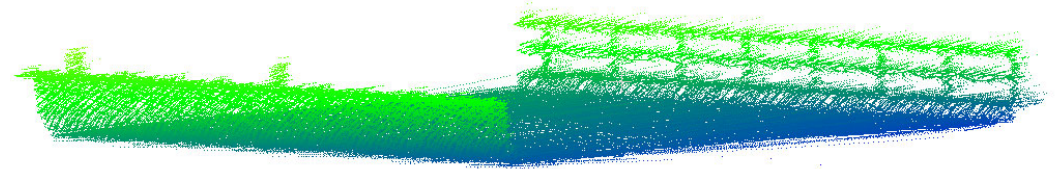
Test Area: Stadium



Collected point cloud
(Colored by height)

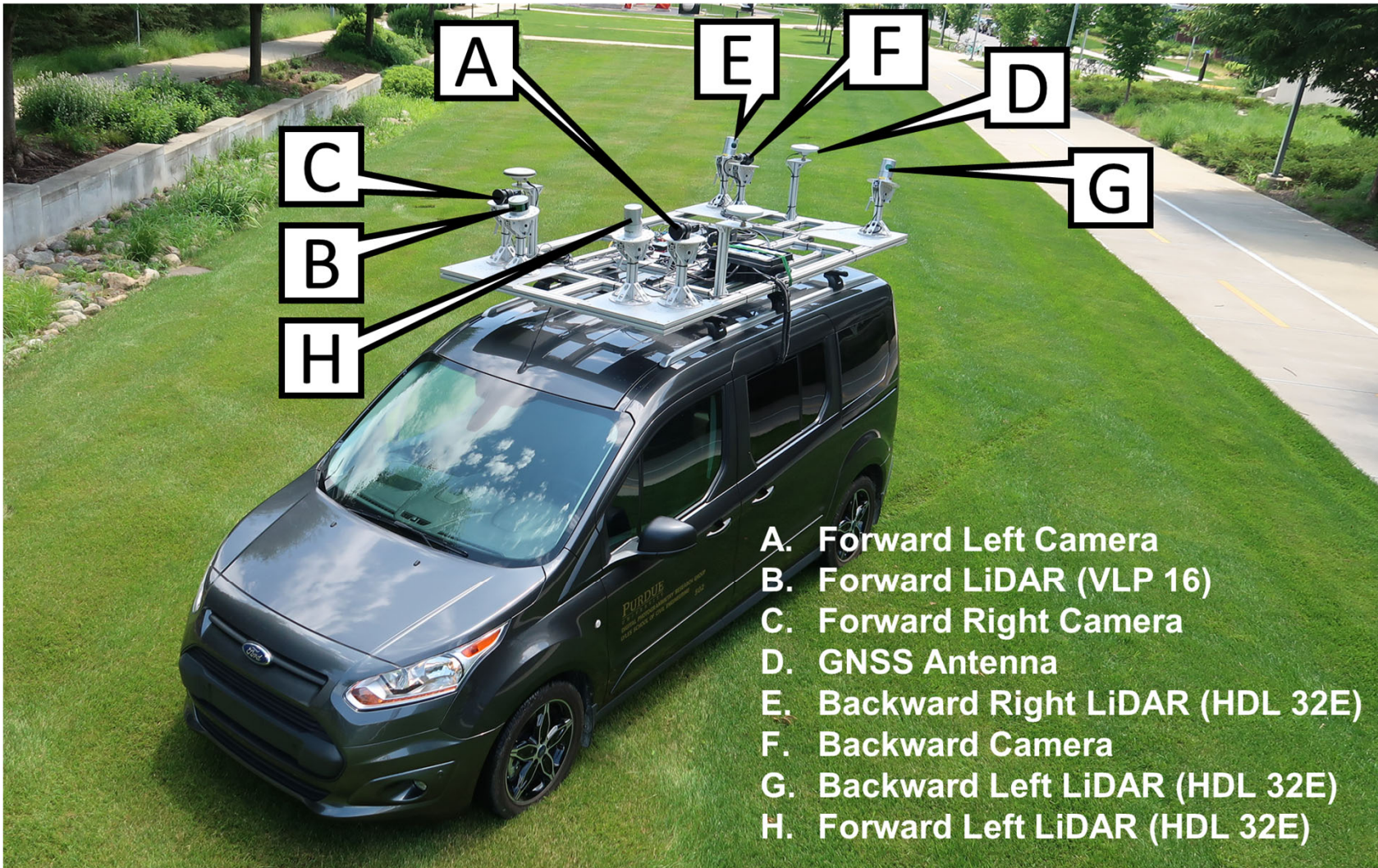
Purdue University

Terrestrial Mobile Mapping Systems



Purdue University

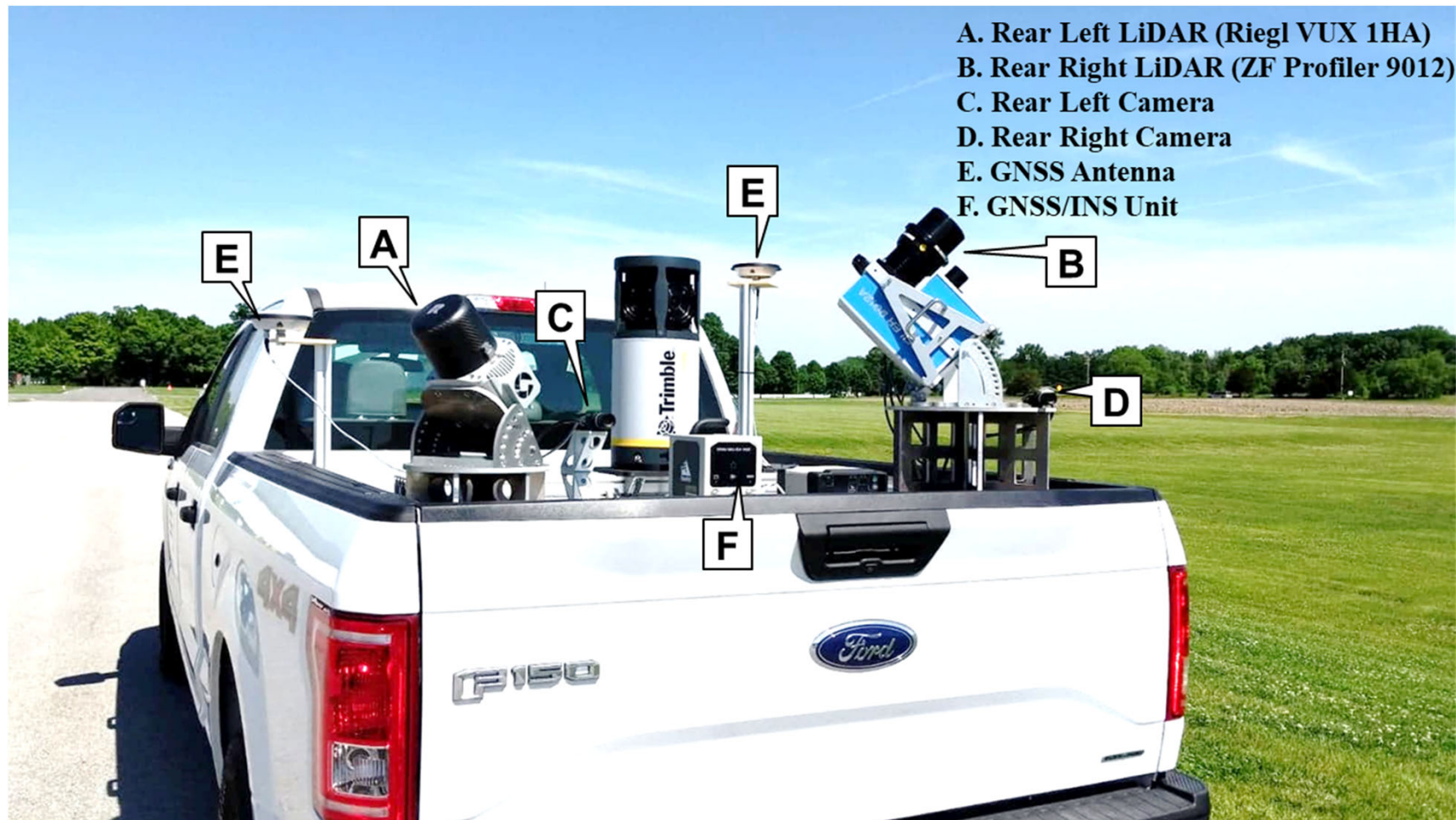
Terrestrial Mobile Mapping Systems



- A. Forward Left Camera
- B. Forward LiDAR (VLP 16)
- C. Forward Right Camera
- D. GNSS Antenna
- E. Backward Right LiDAR (HDL 32E)
- F. Backward Camera
- G. Backward Left LiDAR (HDL 32E)
- H. Forward Left LiDAR (HDL 32E)

Purdue University

Terrestrial Mobile Mapping Systems

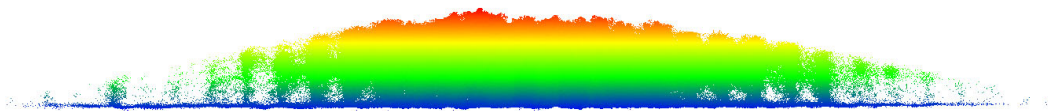


Purdue University

Terrestrial Mobile Mapping Systems



Phenomobile: RGB, Hyperspectral, and LiDAR



Purdue University

Terrestrial (Close Range) Imagery



Mobile Mapping Systems (MMS)



Terrestrial (Close Range) Imagery



Mobile Mapping Systems (MMS)

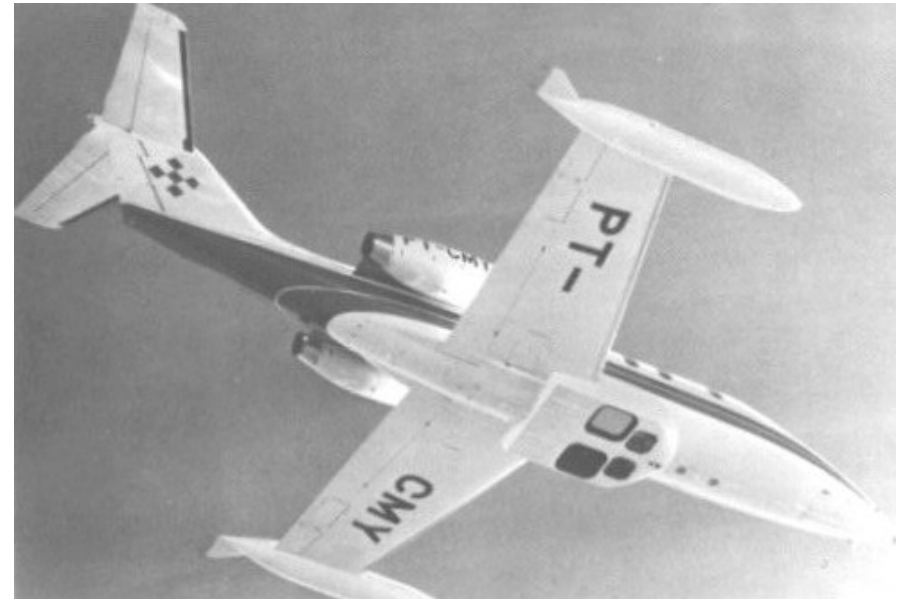


Collecting Inventories

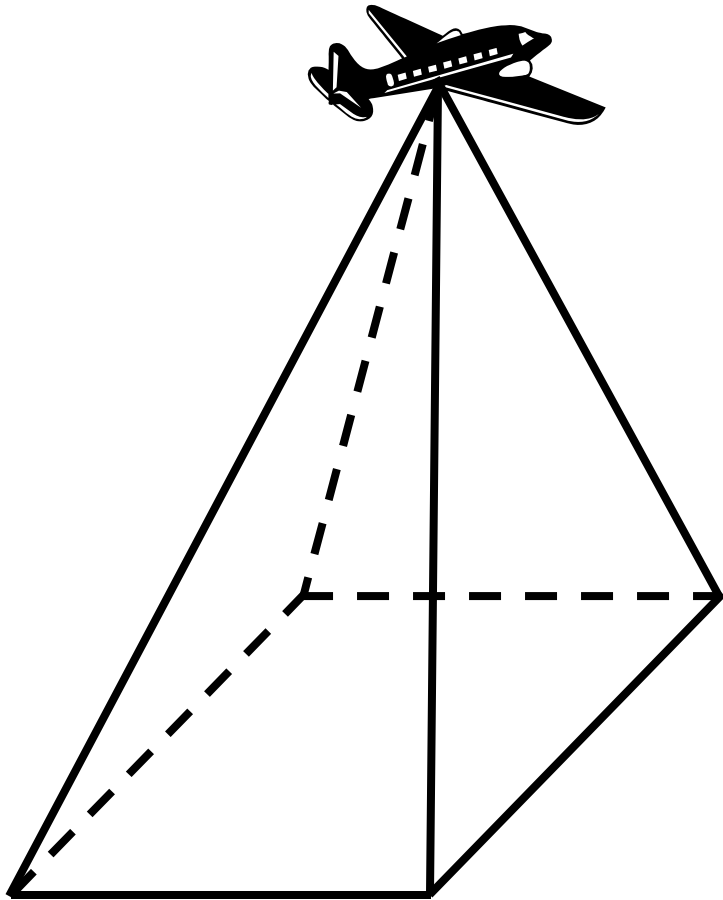
Database Integration

On-going Maintenance

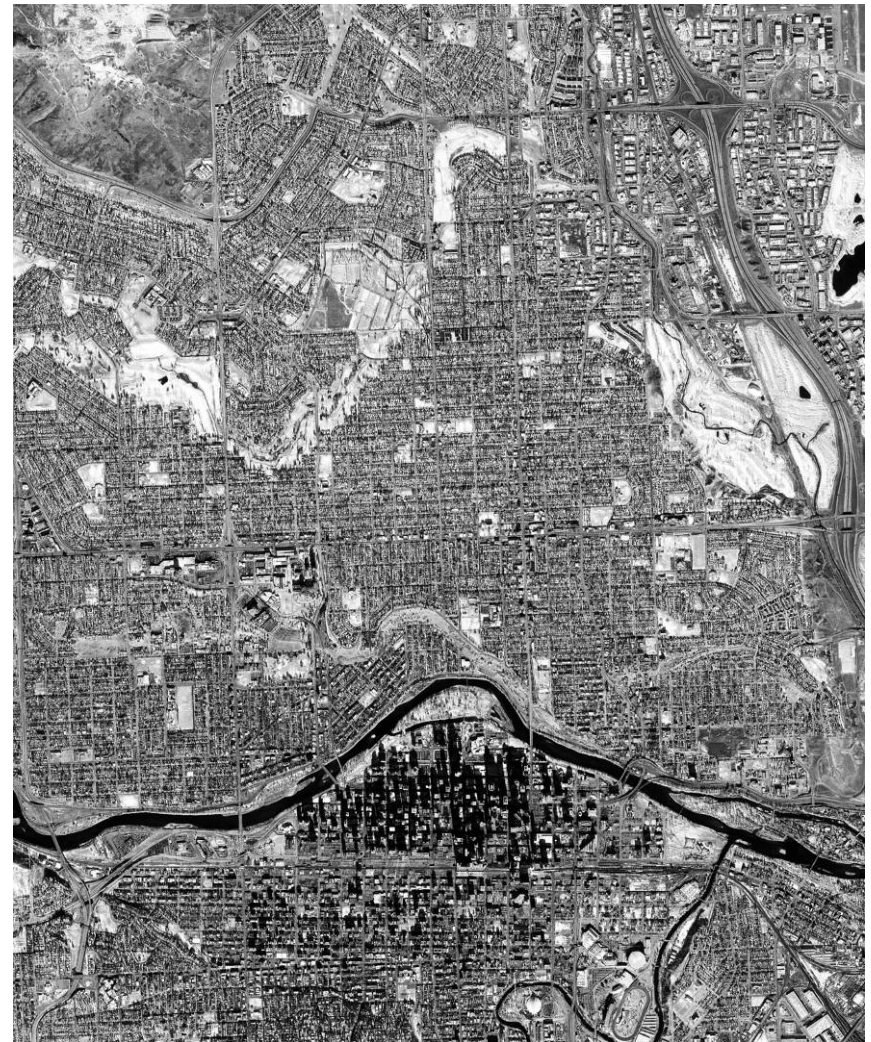
Aerial Imagery



Aerial Imagery



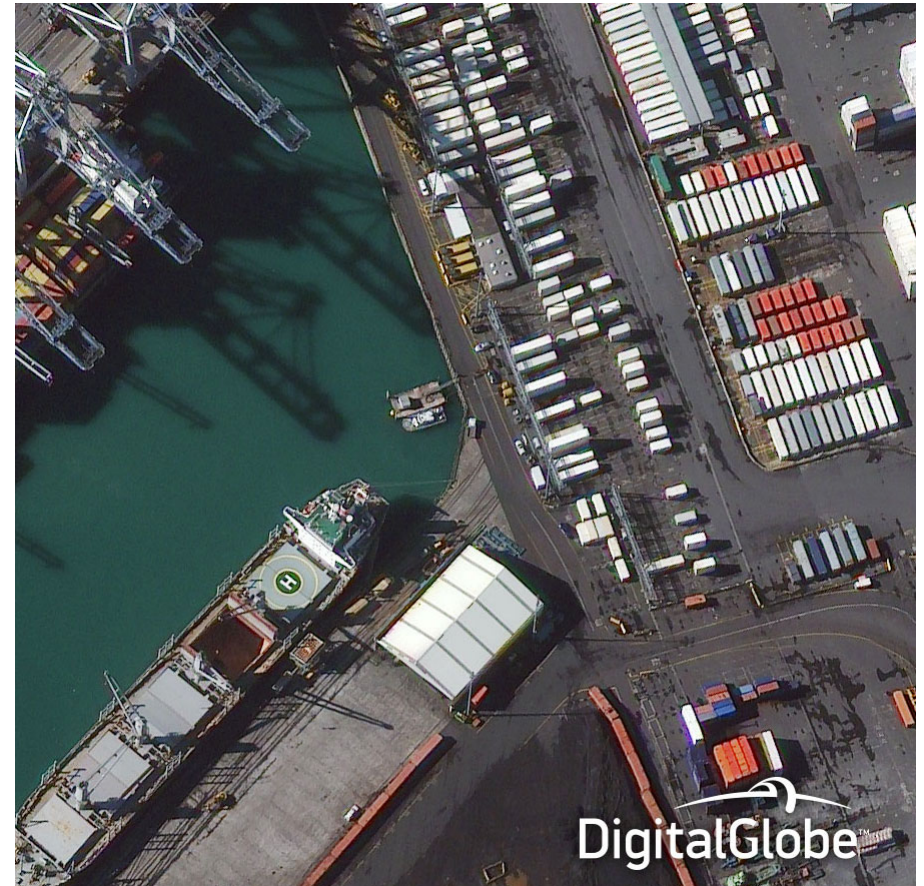
Satellite Imagery



Satellite Imagery



IKONOS



Digital Globe – WorldView 3 (30cm GSD)

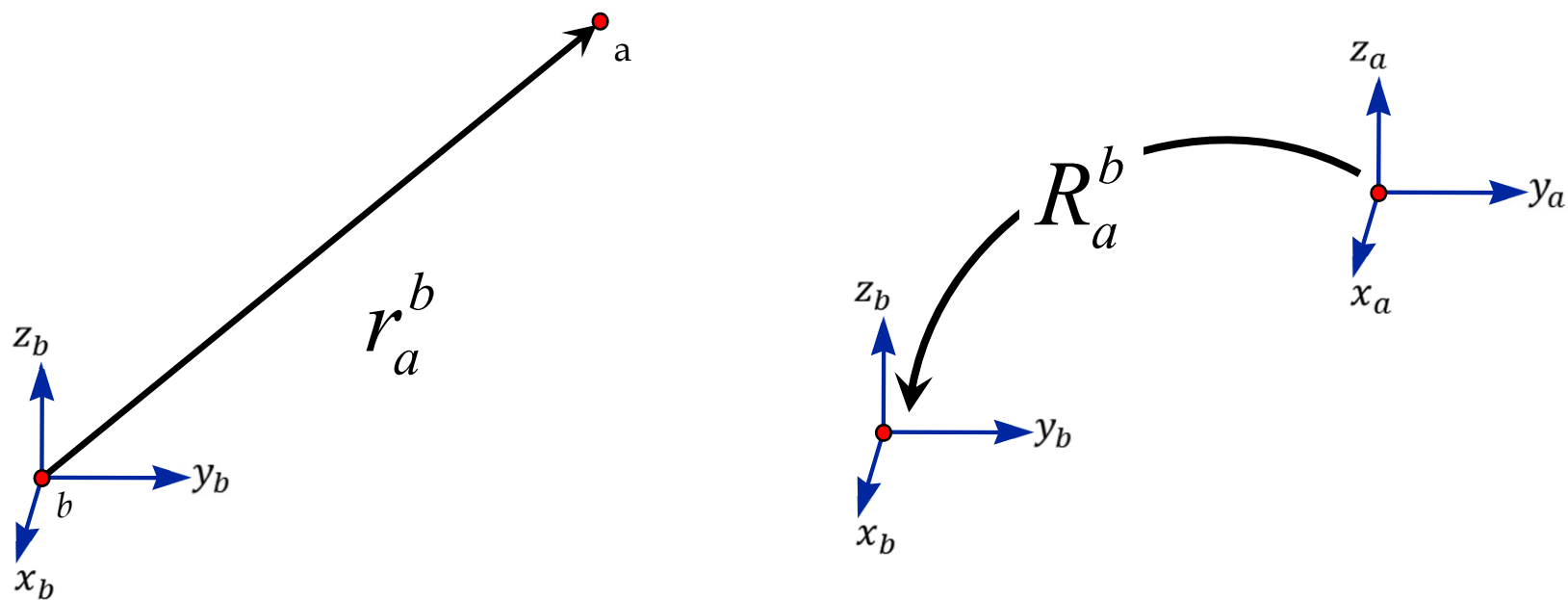


Notations

- r_a^b Stands for the coordinates of point ***a*** relative to point ***b*** – this vector is defined relative to the coordinate system associated with point ***b***.

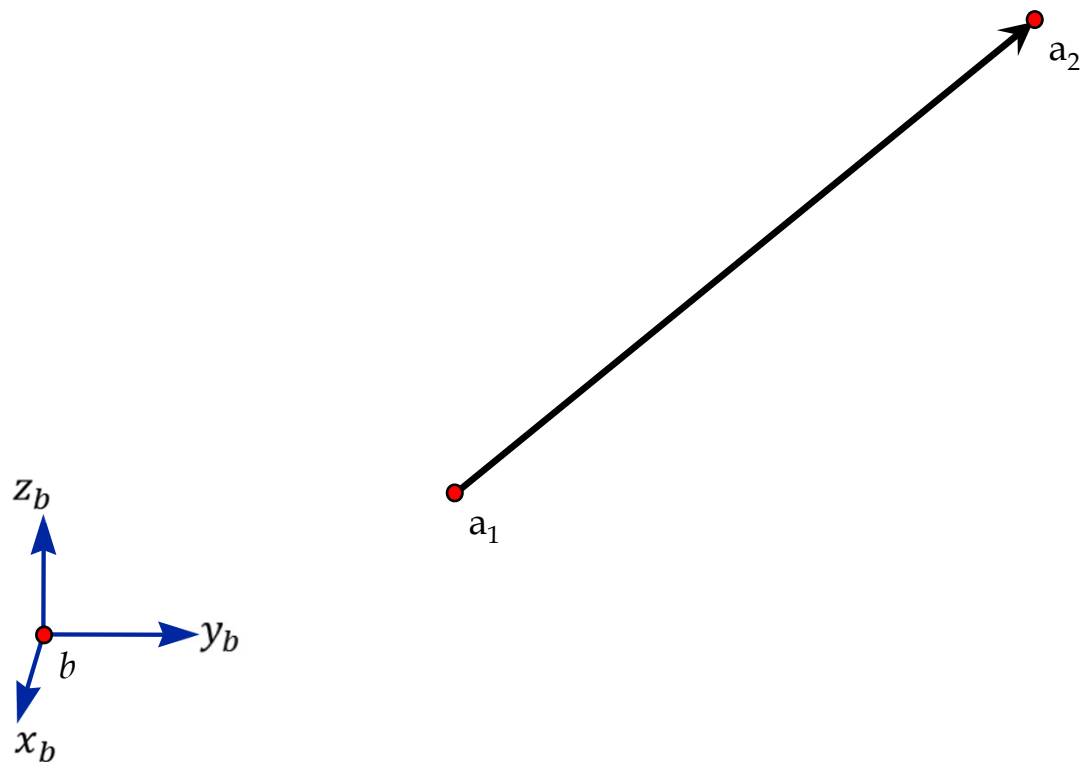
- R_a^b Stands for the rotation matrix that transforms a vector defined relative to the coordinate system denoted by ***a*** into a vector defined relative to the coordinate system denoted by ***b***.

Notations



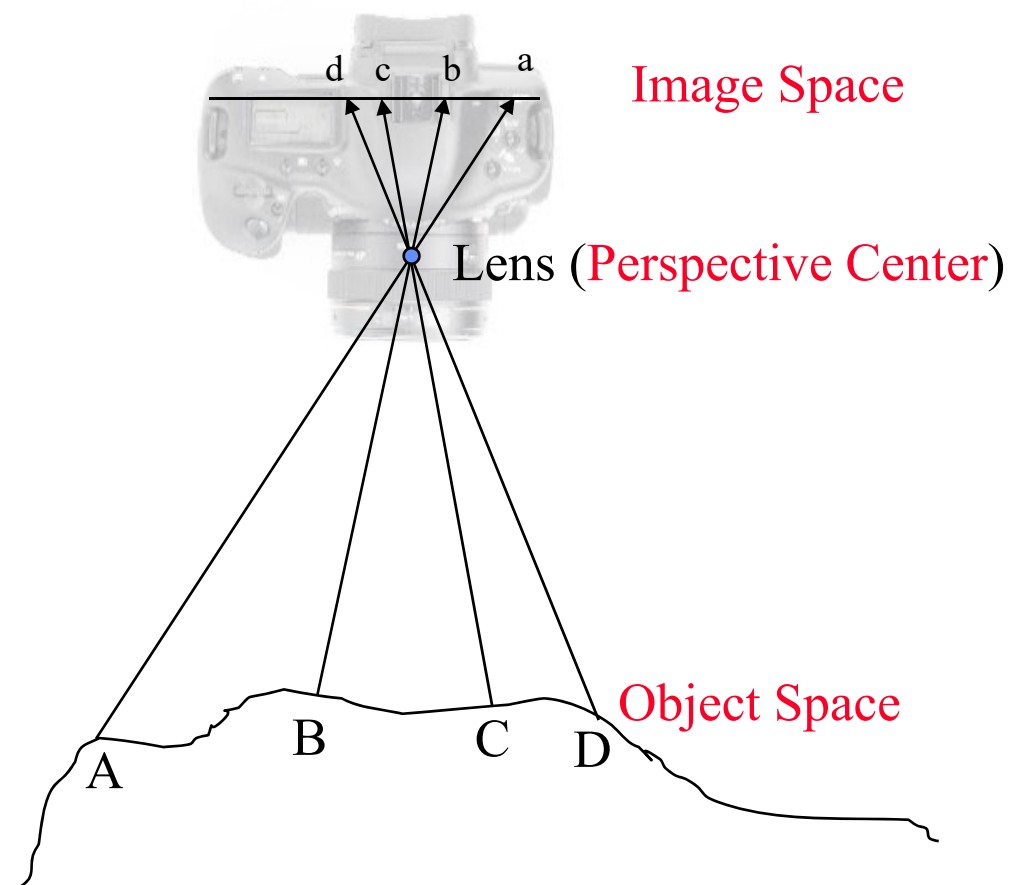


Notations



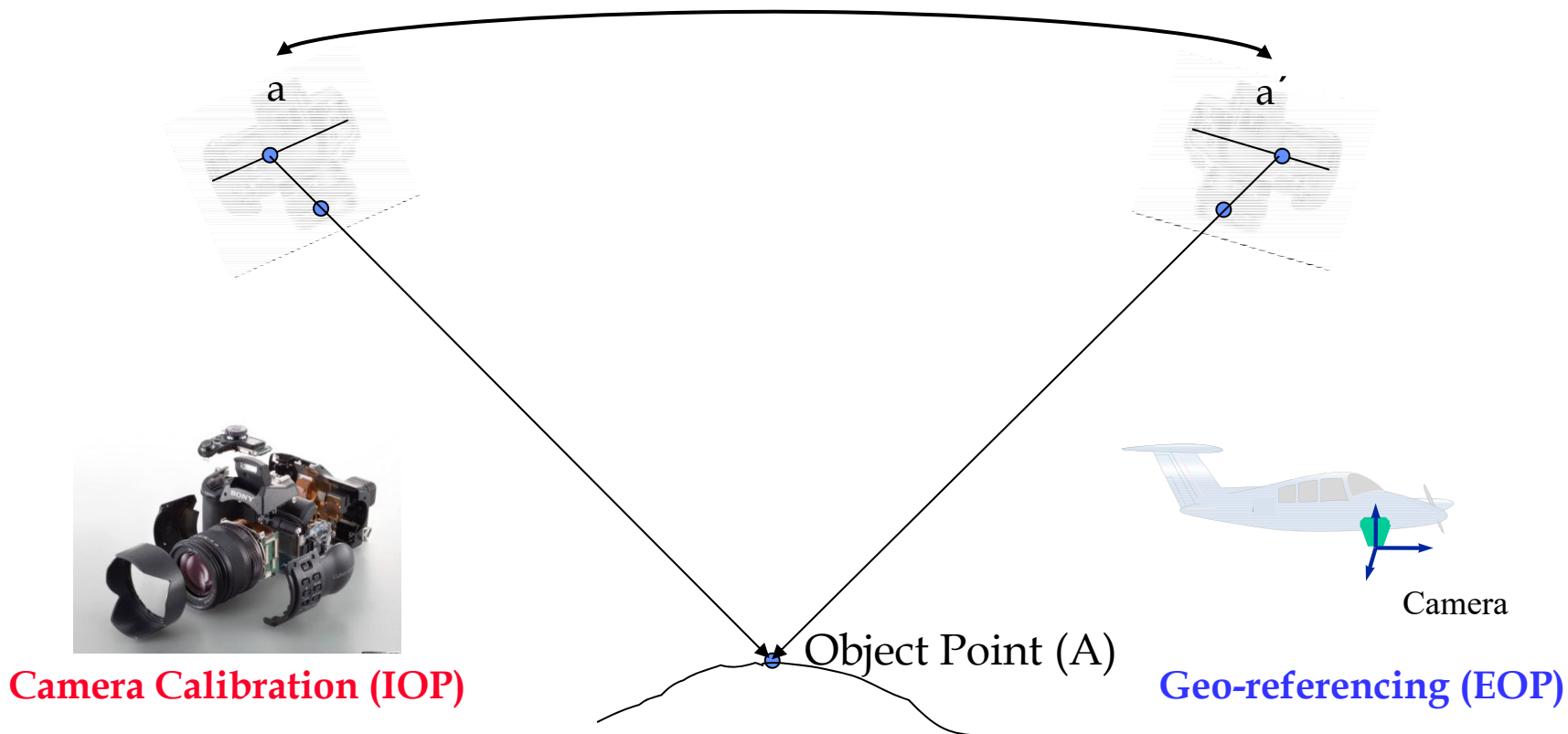
$$r_{a_1 a_2}^b$$

Photography



Photogrammetry

Conjugate Points



- The interior orientation parameters of the involved cameras have to be known.
- The position and the orientation of the camera stations have to be known.

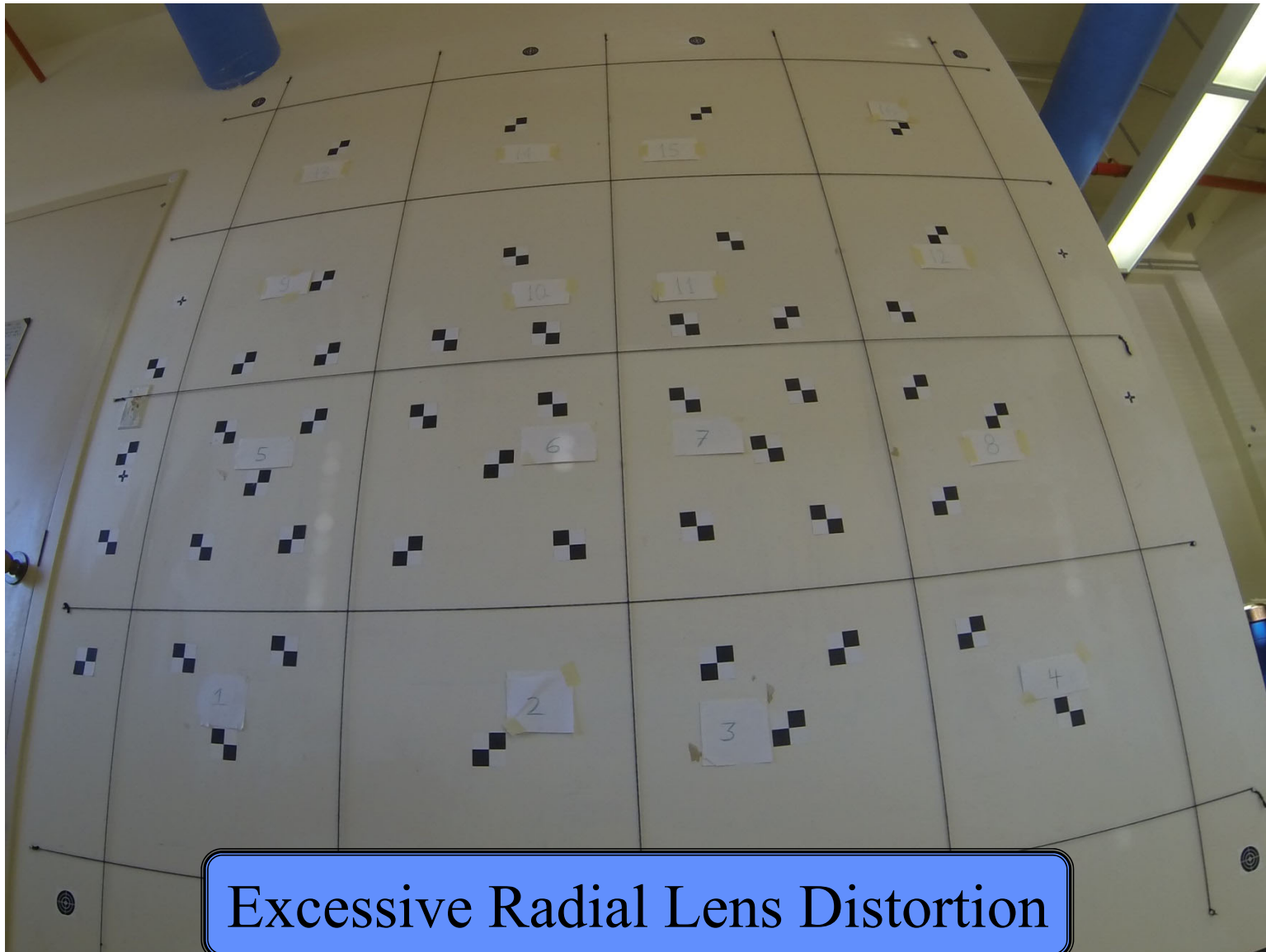
Camera Calibration

- Alternative procedures for camera calibration are well established.
 - Laboratory camera calibration (Multi-collimators)
 - Indoor camera calibration
 - In-situ camera calibration

Analytical camera calibration

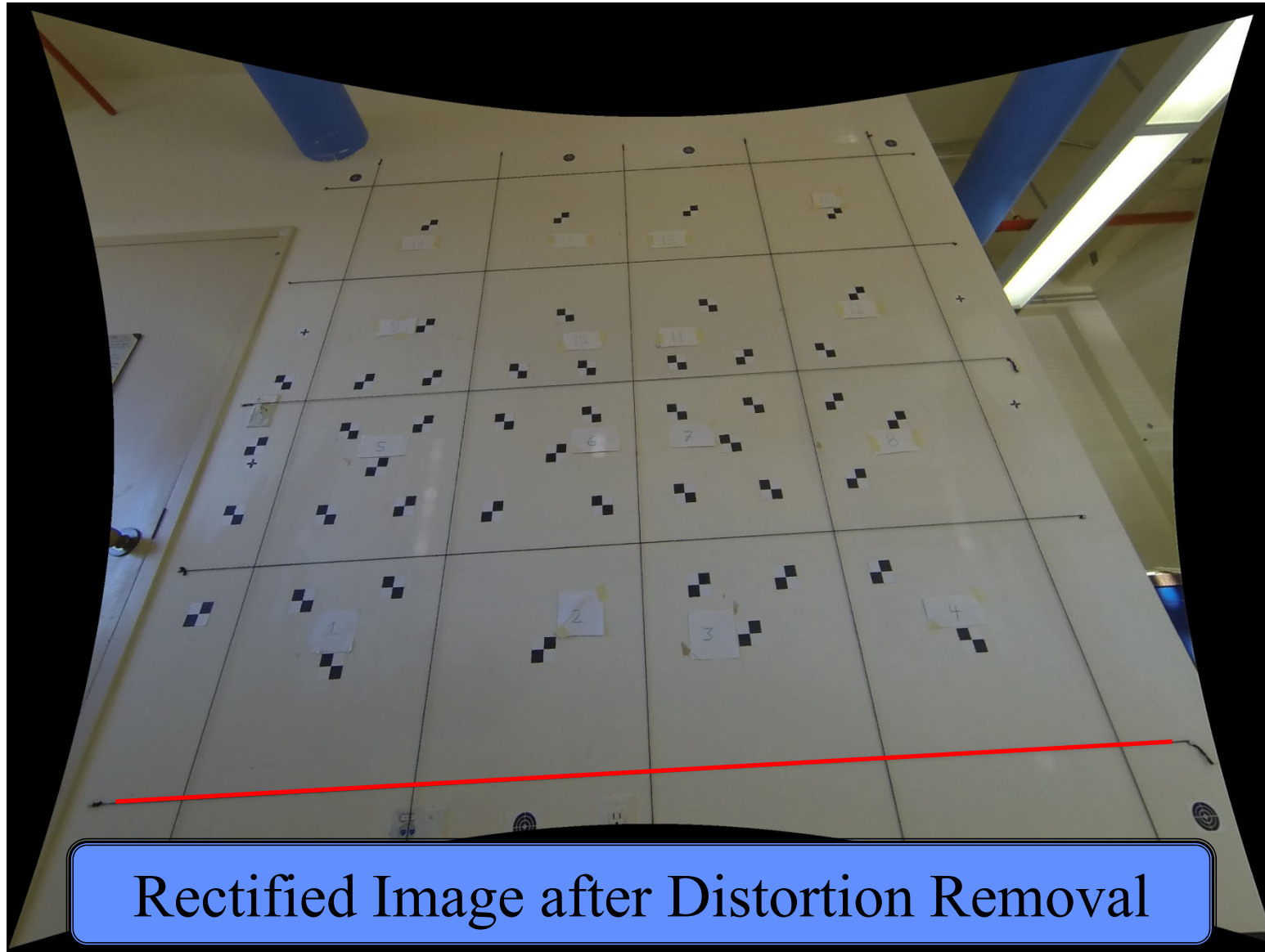


Camera Calibration



Excessive Radial Lens Distortion

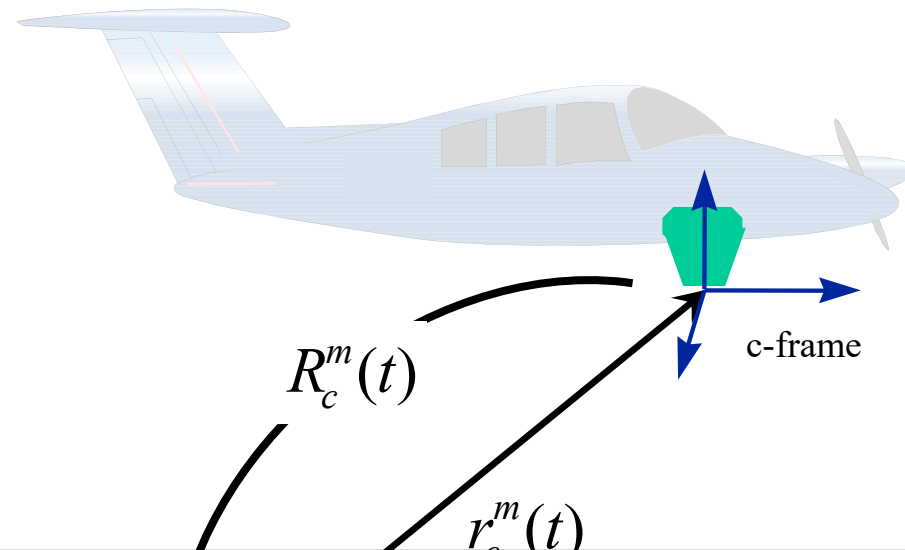
Camera Calibration



Rectified Image after Distortion Removal

Georeferencing

- Exterior Orientation Parameters (EOPs) define the position, $r_c^m(t)$, and orientation $R_c^m(t)$, of the camera coordinate system relative to the mapping reference frame at the moment of exposure.

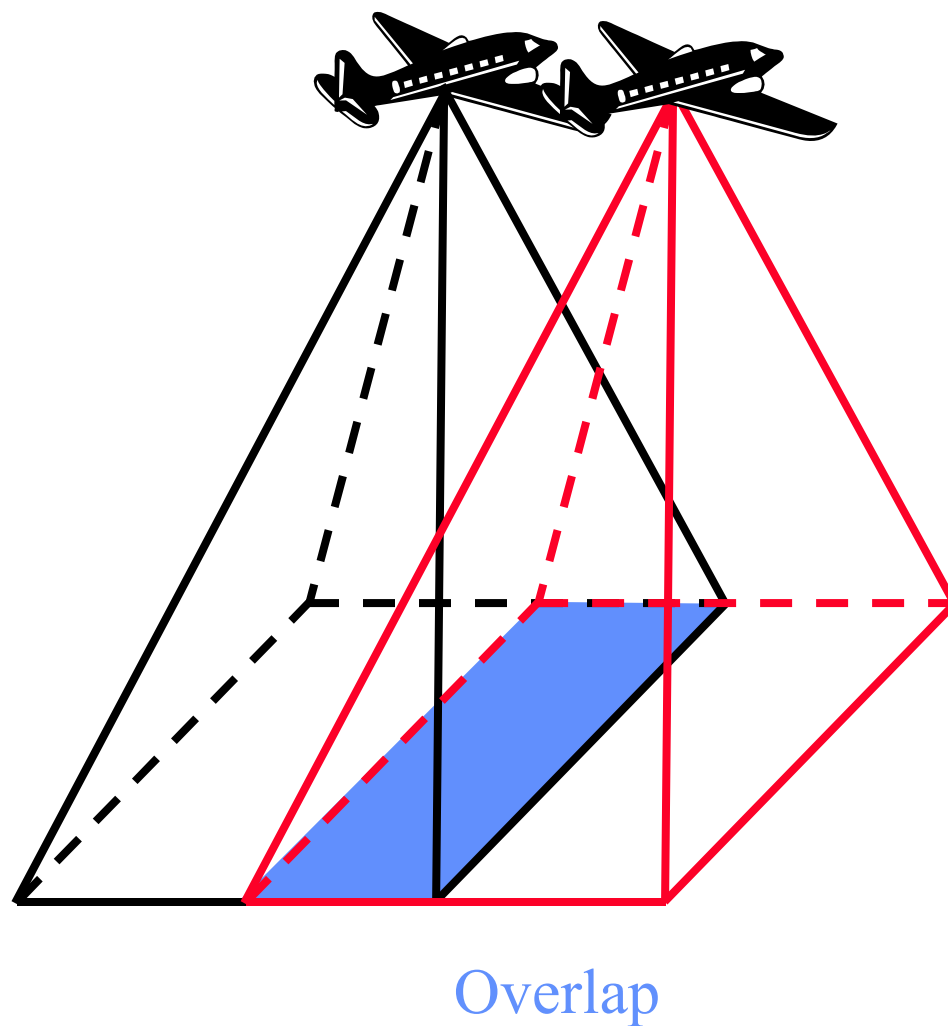


EOPs can be either:

- Indirectly estimated using Ground Control Points (GCPs), or
- Directly derived using GNSS/INS units onboard the imaging platform.



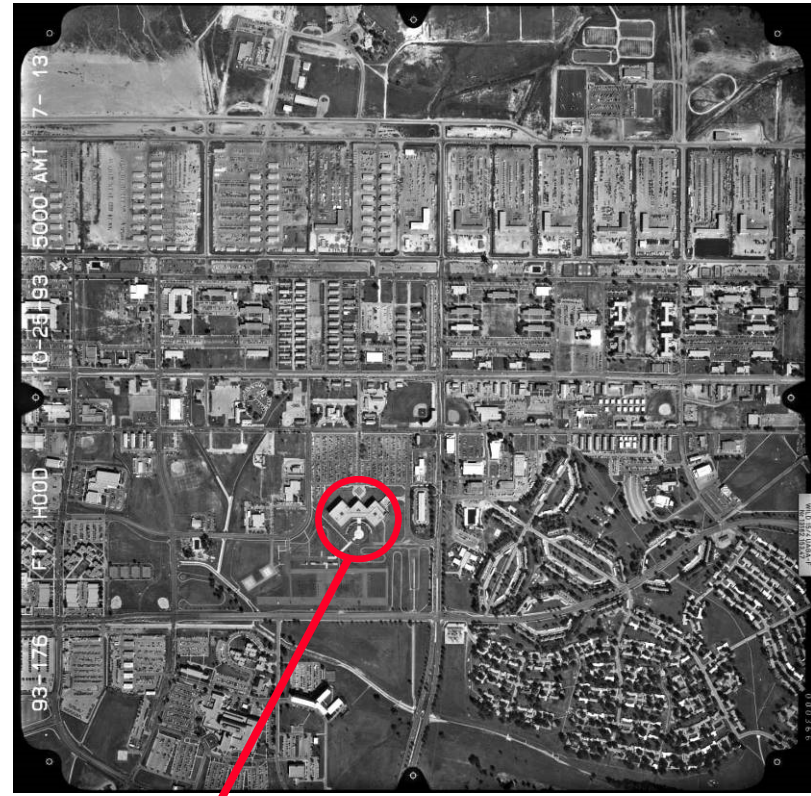
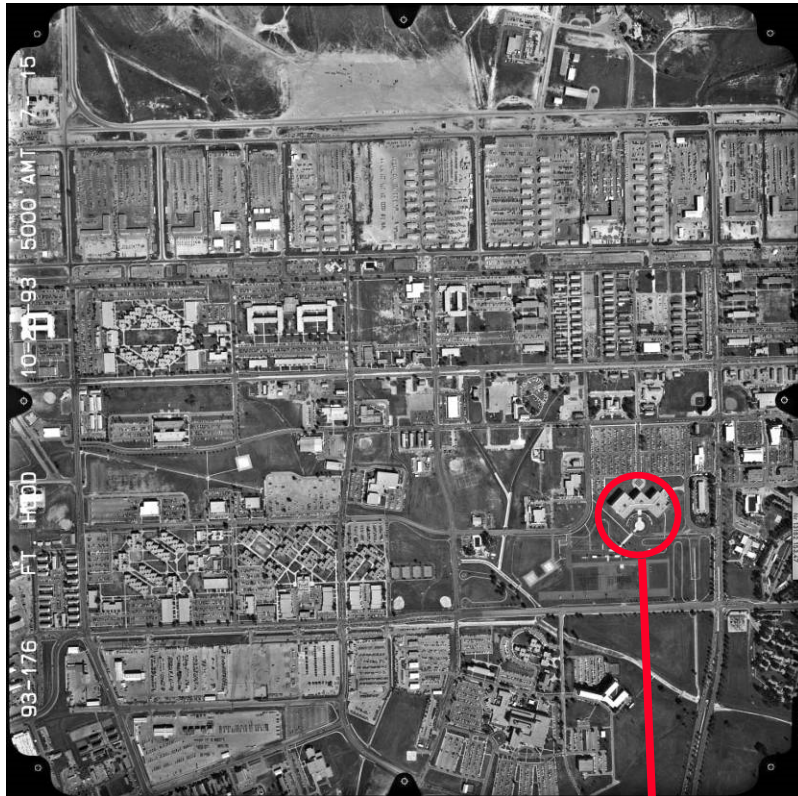
Photogrammetry



Photogrammetry



Photogrammetry



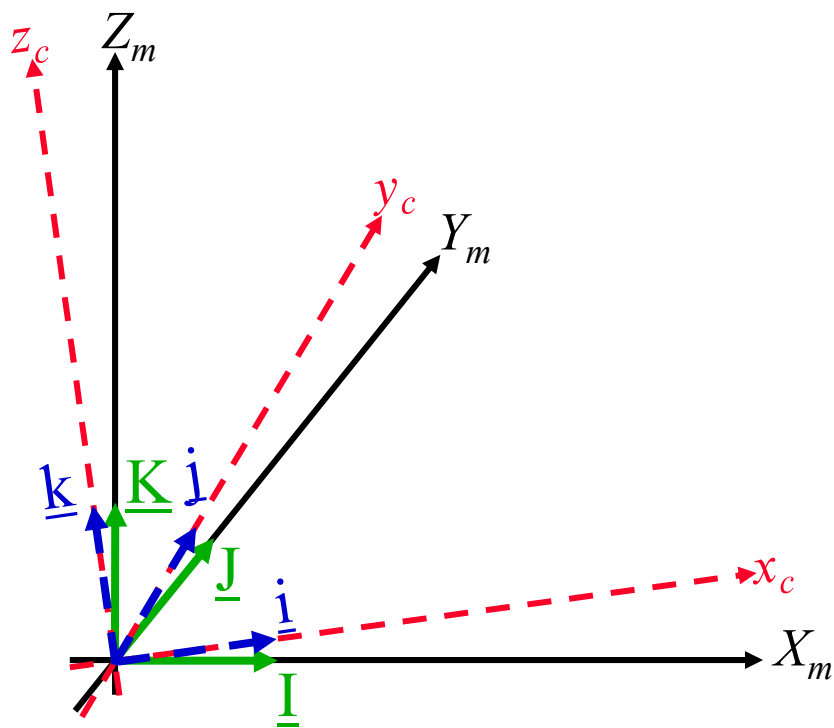
Photogrammetry: Necessary Tools



- Rotation matrices:
 - Express the mathematical relationship between two coordinate systems
 - In a three-dimensional space, a rotation matrix involves at most three independent rotation angles.
- Photogrammetric orientation:
 - Internal characteristics: Interior Orientation Parameters (IOPs)
 - External characteristics: Exterior Orientation Parameters (EOPs)
- Collinearity conditions:
 - The general mathematical model relating the image and ground coordinates of corresponding points

Rotation Matrix

- A rotation matrix transforms a vector from one coordinate system to another.



$$r_a^m = R_c^m r_a^c$$

$$R_c^m = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$$



Rotation Matrix

- Let's consider the transformation of a unit vector along the x-axis of the camera coordinate system

$$r_a^m = R_c^m r_a^c$$
$$\begin{bmatrix} r_{11} \\ r_{21} \\ r_{31} \end{bmatrix} = R_c^m \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

- The first column of the rotation matrix represents the components of a unit vector along the x-axis of the camera coordinate system w.r.t. the mapping reference frame.
- The norm of the first column is unity.

$$r_{11}^2 + r_{21}^2 + r_{31}^2 = 1 \quad \mathbf{1}$$



Rotation Matrix

- Let's consider the transformation of a unit vector along the y-axis of the camera coordinate system

$$r_a^m = R_c^m r_a^c$$
$$\begin{bmatrix} r_{12} \\ r_{22} \\ r_{32} \end{bmatrix} = R_c^m \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

- The second column of the rotation matrix represents the components of a unit vector along the y-axis of the camera coordinate system w.r.t. the mapping reference frame.
- The norm of the second column is unity.

$$r_{12}^2 + r_{22}^2 + r_{32}^2 = 1$$

2



Rotation Matrix

- Let's consider the transformation of a unit vector along the z-axis of the camera coordinate system

$$r_a^m = R_c^m r_a^c$$
$$\begin{bmatrix} r_{13} \\ r_{23} \\ r_{33} \end{bmatrix} = R_c^m \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

- The third column of the rotation matrix represents the components of a unit vector along the z-axis of the camera coordinate system w.r.t. the mapping reference frame.
- The norm of the third column is unity.

$$r_{13}^2 + r_{23}^2 + r_{33}^2 = 1$$

3



Rotation Matrix

- Since the x and y axes of the camera coordinate system are orthogonal to each other, then

$$\begin{bmatrix} r_{11} \\ r_{21} \\ r_{31} \end{bmatrix} \cdot \begin{bmatrix} r_{12} \\ r_{22} \\ r_{32} \end{bmatrix} = 0 \quad r_{11} r_{12} + r_{21} r_{22} + r_{31} r_{32} = 0 \quad 4$$

- Since the x and z axes of the camera coordinate system are orthogonal to each other, then

$$\begin{bmatrix} r_{11} \\ r_{21} \\ r_{31} \end{bmatrix} \cdot \begin{bmatrix} r_{13} \\ r_{23} \\ r_{33} \end{bmatrix} = 0 \quad r_{11} r_{13} + r_{21} r_{23} + r_{31} r_{33} = 0 \quad 5$$



Rotation Matrix

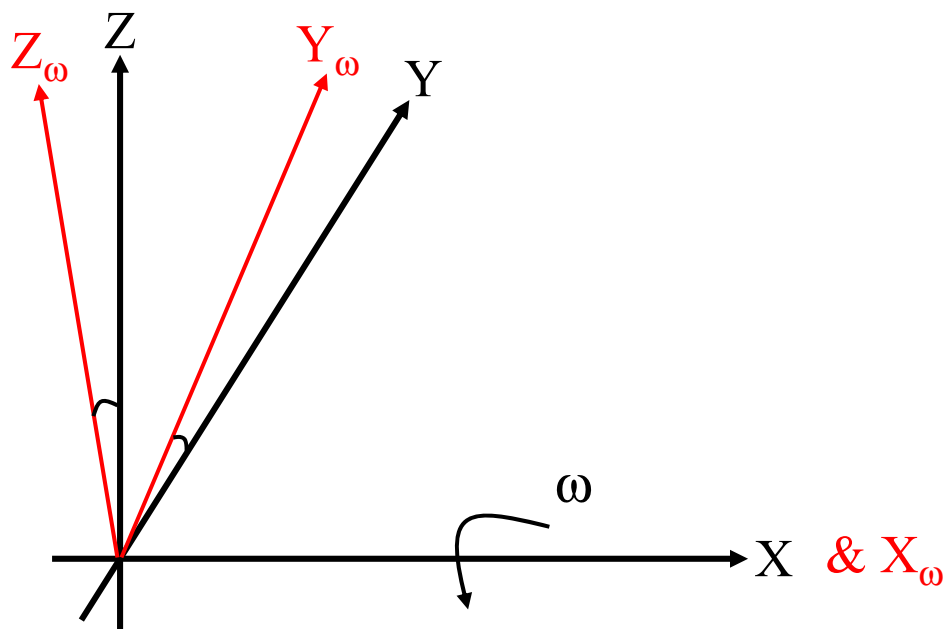
- Since the y and z axes of the camera coordinate system are orthogonal to each other, then

$$\begin{bmatrix} r_{12} \\ r_{22} \\ r_{32} \end{bmatrix} \cdot \begin{bmatrix} r_{13} \\ r_{23} \\ r_{33} \end{bmatrix} = 0 \qquad r_{12} r_{13} + r_{22} r_{23} + r_{32} r_{33} = 0 \quad \mathbf{6}$$

- Since the nine elements of a rotation matrix must satisfy six constraints (orthogonality constraints), **a 3D rotation matrix is defined by a maximum of three independent parameters/rotation angles.**
- In photogrammetry, the rotation matrix is defined by the angles (ω , ϕ , and κ).



Primary Rotation (ω)





Primary Rotation (ω)

$$\begin{bmatrix} x_\omega \\ y_\omega \\ z_\omega \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \omega & \sin \omega \\ 0 & -\sin \omega & \cos \omega \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

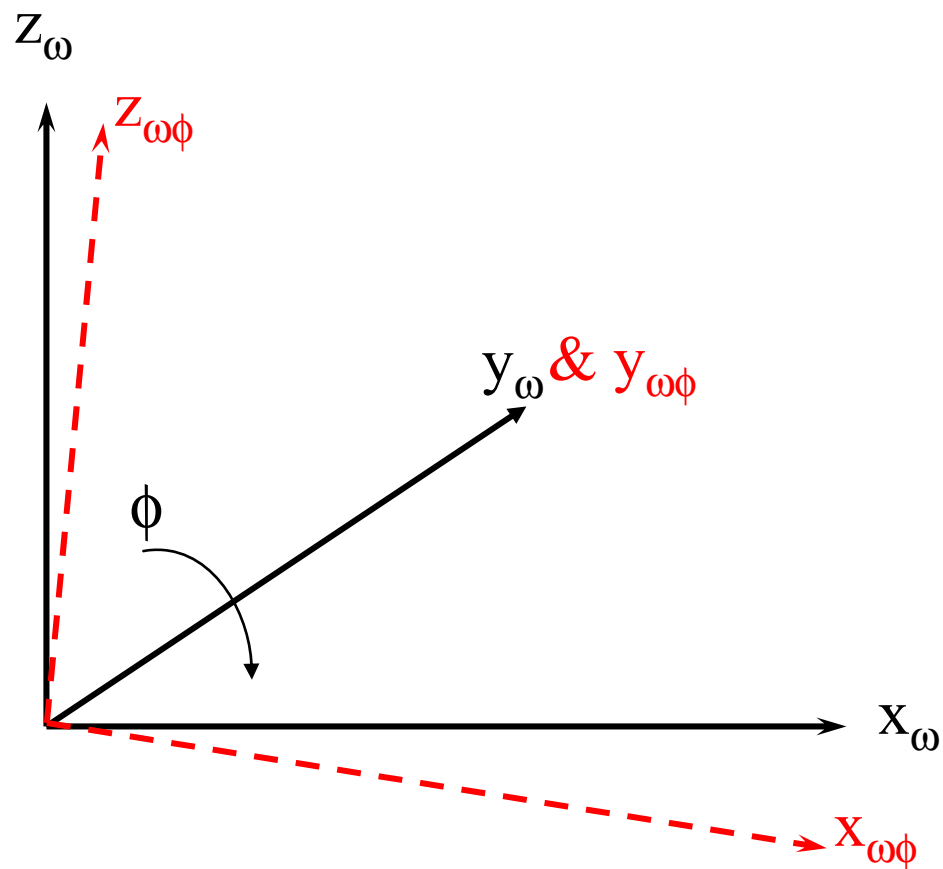
$$\begin{bmatrix} x_\omega \\ y_\omega \\ z_\omega \end{bmatrix} = M_\omega \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \omega & -\sin \omega \\ 0 & \sin \omega & \cos \omega \end{bmatrix} \begin{bmatrix} x_\omega \\ y_\omega \\ z_\omega \end{bmatrix}$$

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = R_\omega \begin{bmatrix} x_\omega \\ y_\omega \\ z_\omega \end{bmatrix}$$



Secondary Rotation (ϕ)





Secondary Rotation (ϕ)

$$\begin{bmatrix} x_{\omega\phi} \\ y_{\omega\phi} \\ z_{\omega\phi} \end{bmatrix} = \begin{bmatrix} \cos \phi & 0 & -\sin \phi \\ 0 & 1 & 0 \\ \sin \phi & 0 & \cos \phi \end{bmatrix} \begin{bmatrix} x_{\omega} \\ y_{\omega} \\ z_{\omega} \end{bmatrix}$$

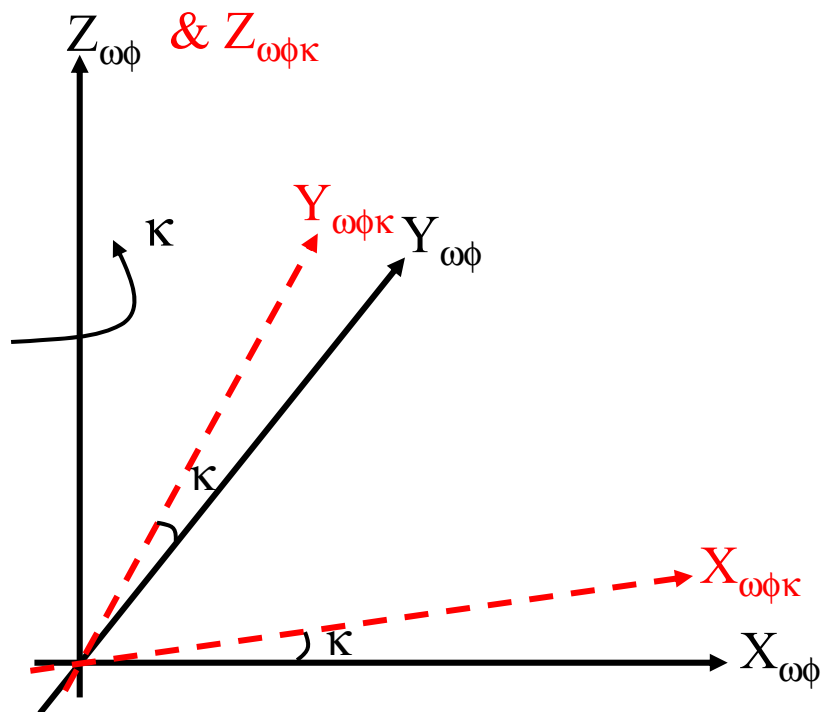
$$\begin{bmatrix} x_{\omega\phi} \\ y_{\omega\phi} \\ z_{\omega\phi} \end{bmatrix} = M_{\phi} \begin{bmatrix} x_{\omega} \\ y_{\omega} \\ z_{\omega} \end{bmatrix}$$

$$\begin{bmatrix} x_{\omega} \\ y_{\omega} \\ z_{\omega} \end{bmatrix} = \begin{bmatrix} \cos \phi & 0 & \sin \phi \\ 0 & 1 & 0 \\ -\sin \phi & 0 & \cos \phi \end{bmatrix} \begin{bmatrix} x_{\omega\phi} \\ y_{\omega\phi} \\ z_{\omega\phi} \end{bmatrix}$$

$$\begin{bmatrix} x_{\omega} \\ y_{\omega} \\ z_{\omega} \end{bmatrix} = R_{\phi} \begin{bmatrix} x_{\omega\phi} \\ y_{\omega\phi} \\ z_{\omega\phi} \end{bmatrix}$$



Tertiary Rotation (κ)





Tertiary Rotation (κ)

$$\begin{bmatrix} x_{\omega\phi\kappa} \\ y_{\omega\phi\kappa} \\ z_{\omega\phi\kappa} \end{bmatrix} = \begin{bmatrix} \cos \kappa & \sin \kappa & 0 \\ -\sin \kappa & \cos \kappa & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_{\omega\phi} \\ y_{\omega\phi} \\ z_{\omega\phi} \end{bmatrix}$$

$$\begin{bmatrix} x_{\omega\phi\kappa} \\ y_{\omega\phi\kappa} \\ z_{\omega\phi\kappa} \end{bmatrix} = M_{\kappa} \begin{bmatrix} x_{\omega\phi} \\ y_{\omega\phi} \\ z_{\omega\phi} \end{bmatrix}$$

$$\begin{bmatrix} x_{\omega\phi} \\ y_{\omega\phi} \\ z_{\omega\phi} \end{bmatrix} = \begin{bmatrix} \cos \kappa & -\sin \kappa & 0 \\ \sin \kappa & \cos \kappa & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_{\omega\phi\kappa} \\ y_{\omega\phi\kappa} \\ z_{\omega\phi\kappa} \end{bmatrix}$$

$$\begin{bmatrix} x_{\omega\phi} \\ y_{\omega\phi} \\ z_{\omega\phi} \end{bmatrix} = R_{\kappa} \begin{bmatrix} x_{\omega\phi\kappa} \\ y_{\omega\phi\kappa} \\ z_{\omega\phi\kappa} \end{bmatrix}$$



Rotation in Space

$$\begin{bmatrix} x_{\omega\phi\kappa} \\ y_{\omega\phi\kappa} \\ z_{\omega\phi\kappa} \end{bmatrix} = M_{\kappa} M_{\phi} M_{\omega} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

// to the image coordinate system

// to the ground coordinate system

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = R_{\omega} R_{\phi} R_{\kappa} \begin{bmatrix} x_{\omega\phi\kappa} \\ y_{\omega\phi\kappa} \\ z_{\omega\phi\kappa} \end{bmatrix}$$

// to the ground coordinate system

// to the image coordinate system



Rotation in Space

$$M_{\kappa} M_{\phi} M_{\omega} = M = \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix}$$

where :

$$m_{11} = \cos \phi \cos \kappa$$

$$m_{12} = \cos \omega \sin \kappa + \sin \omega \sin \phi \cos \kappa$$

$$m_{13} = \sin \omega \sin \kappa - \cos \omega \sin \phi \cos \kappa$$

$$m_{21} = -\cos \phi \sin \kappa$$

$$m_{22} = \cos \omega \cos \kappa - \sin \omega \sin \phi \sin \kappa$$

$$m_{23} = \sin \omega \cos \kappa + \cos \omega \sin \phi \sin \kappa$$

$$m_{31} = \sin \phi$$

$$m_{32} = -\sin \omega \cos \phi$$

$$m_{33} = \cos \omega \cos \phi$$



Rotation in Space

$$R_{\omega} R_{\phi} R_{\kappa} = R = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$$

where :

$$r_{11} = \cos \phi \cos \kappa$$

$$r_{12} = -\cos \phi \sin \kappa$$

$$r_{13} = \sin \phi$$

$$r_{21} = \cos \omega \sin \kappa + \sin \omega \sin \phi \cos \kappa$$

$$r_{22} = \cos \omega \cos \kappa - \sin \omega \sin \phi \sin \kappa$$

$$r_{23} = -\sin \omega \cos \phi$$

$$r_{31} = \sin \omega \sin \kappa - \cos \omega \sin \phi \cos \kappa$$

$$r_{32} = \sin \omega \cos \kappa + \cos \omega \sin \phi \sin \kappa$$

$$r_{33} = \cos \omega \cos \phi$$



Orthogonality Conditions

$$R = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$$

$$r_{11}^2 + r_{21}^2 + r_{31}^2 = 1$$

$$r_{12}^2 + r_{22}^2 + r_{32}^2 = 1$$

$$r_{13}^2 + r_{23}^2 + r_{33}^2 = 1$$

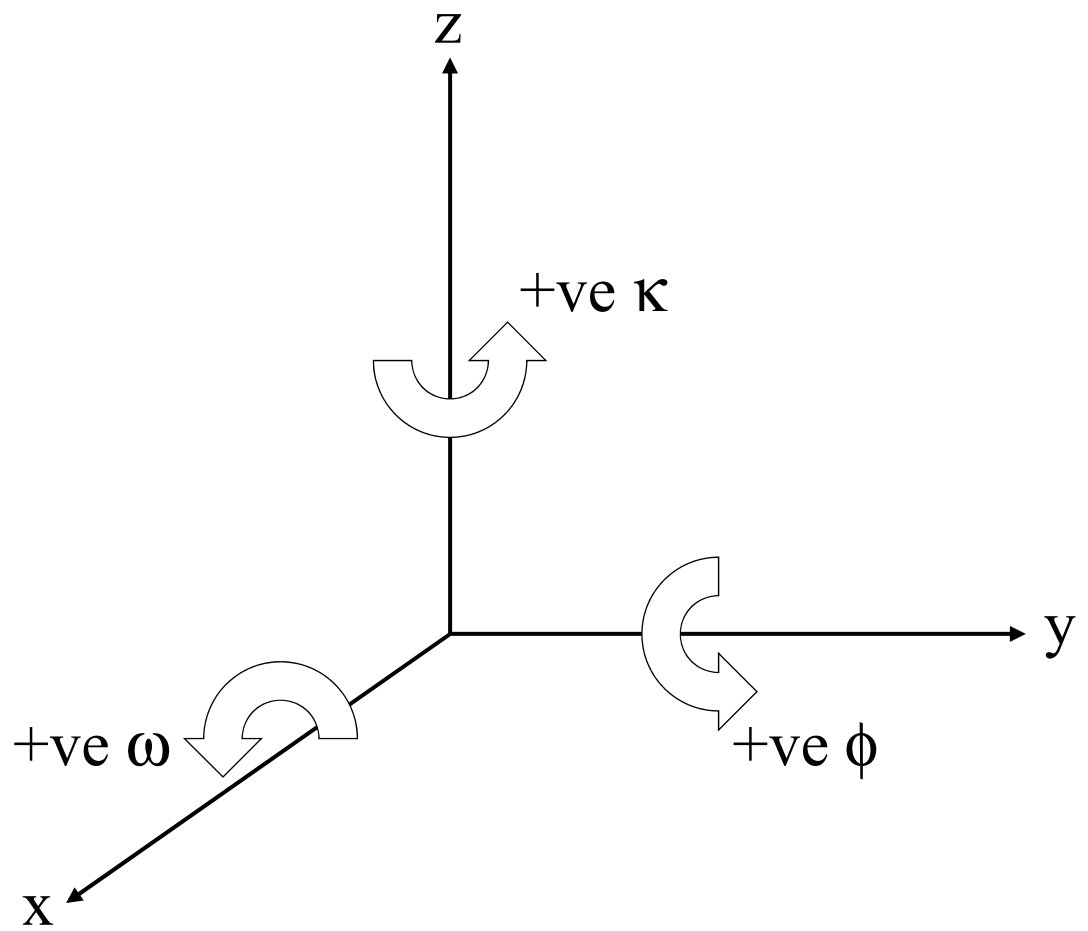
$$r_{11} r_{12} + r_{21} r_{22} + r_{31} r_{32} = 0$$

$$r_{11} r_{13} + r_{21} r_{23} + r_{31} r_{33} = 0$$

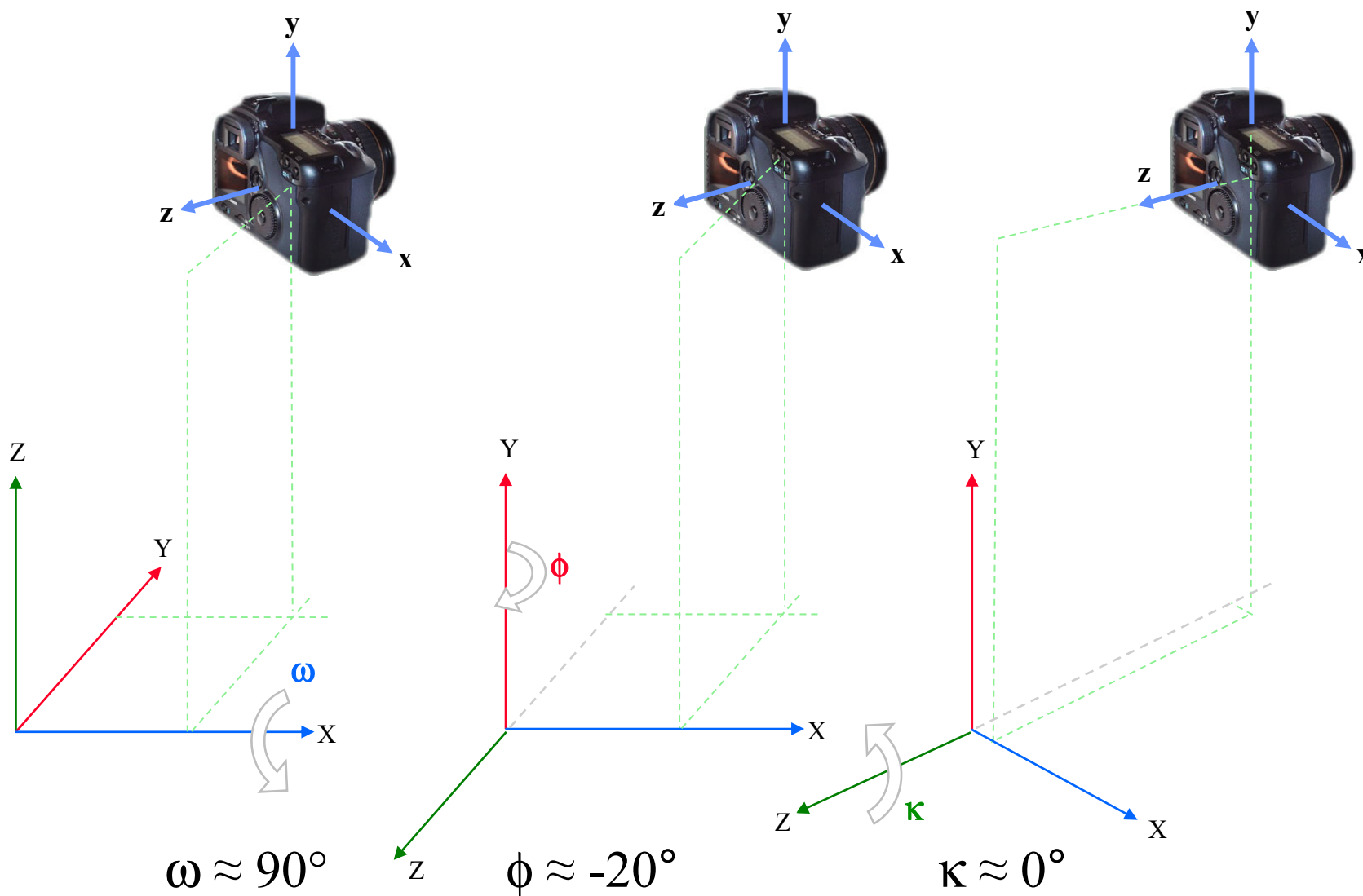
$$r_{12} r_{13} + r_{22} r_{23} + r_{32} r_{33} = 0$$



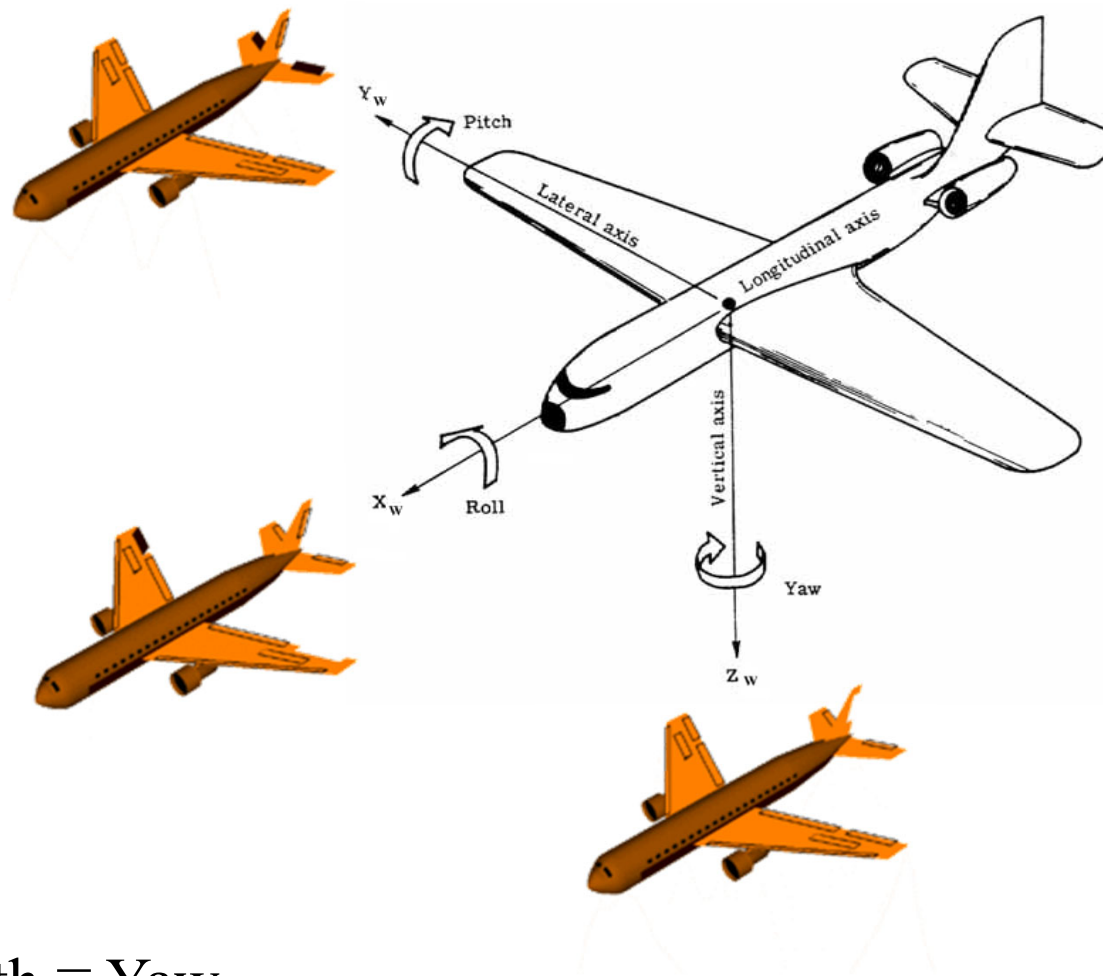
Positive Rotation Angles: (Right Handed System)



Rotation Angles (ω , ϕ , κ)



Rotation Angles (Azimuth, Pitch, Roll)



Azimuth \equiv Yaw



Photogrammetric Orientation

Interior Orientation

Interior Orientation Parameters

- Interior Orientation Parameters (IOPs) describe the internal characteristics of the implemented camera.
 - IOPs include the principal distance, principal point coordinates, and distortion parameters.
 - IOPs are determined using a calibration procedure.



Interior Orientation Parameters

- Alternative procedures for camera calibration are well established.
 - Laboratory camera calibration (Multi-collimators)
 - Indoor camera calibration
 - In-situ camera calibration

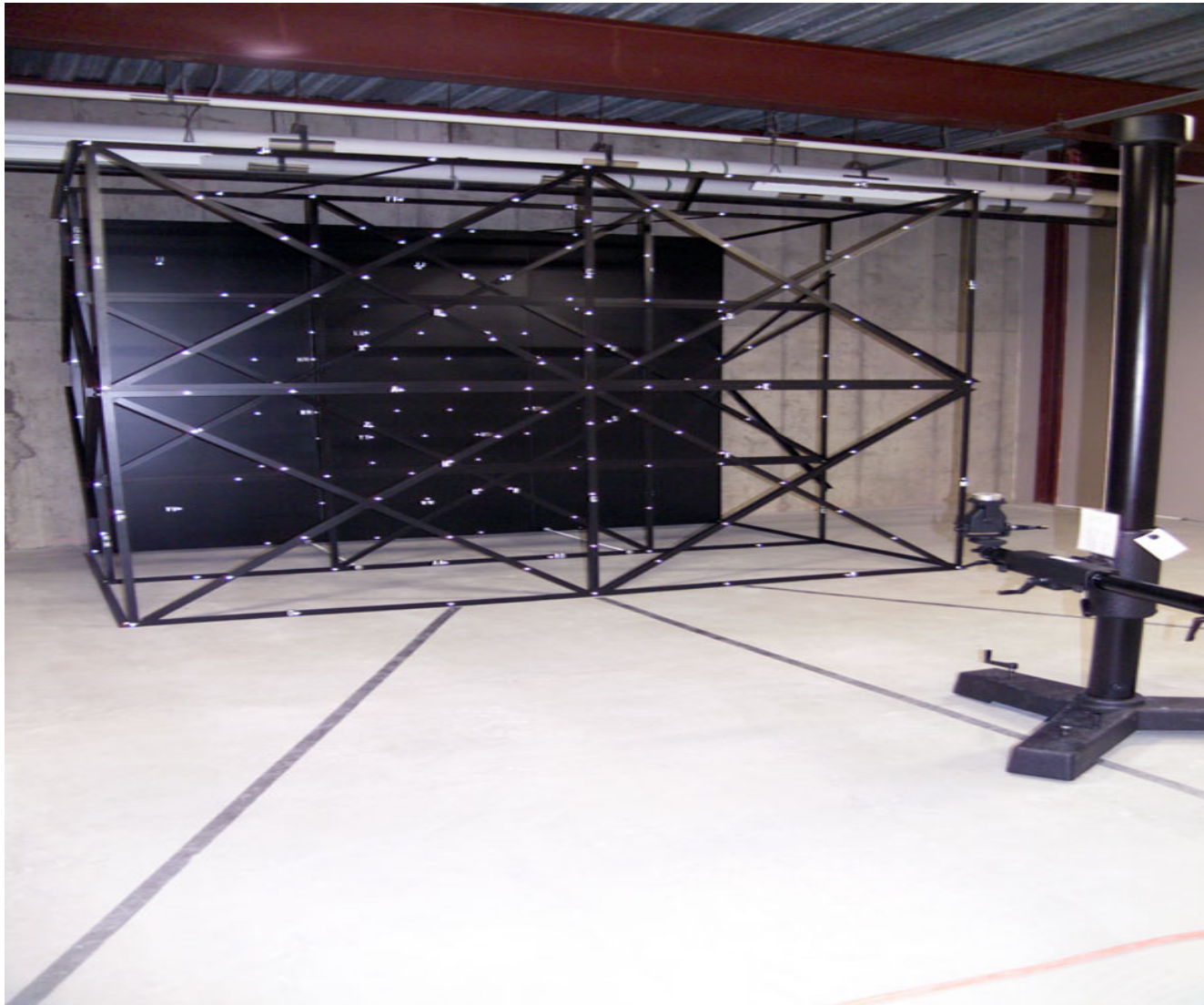
Analytical camera calibration



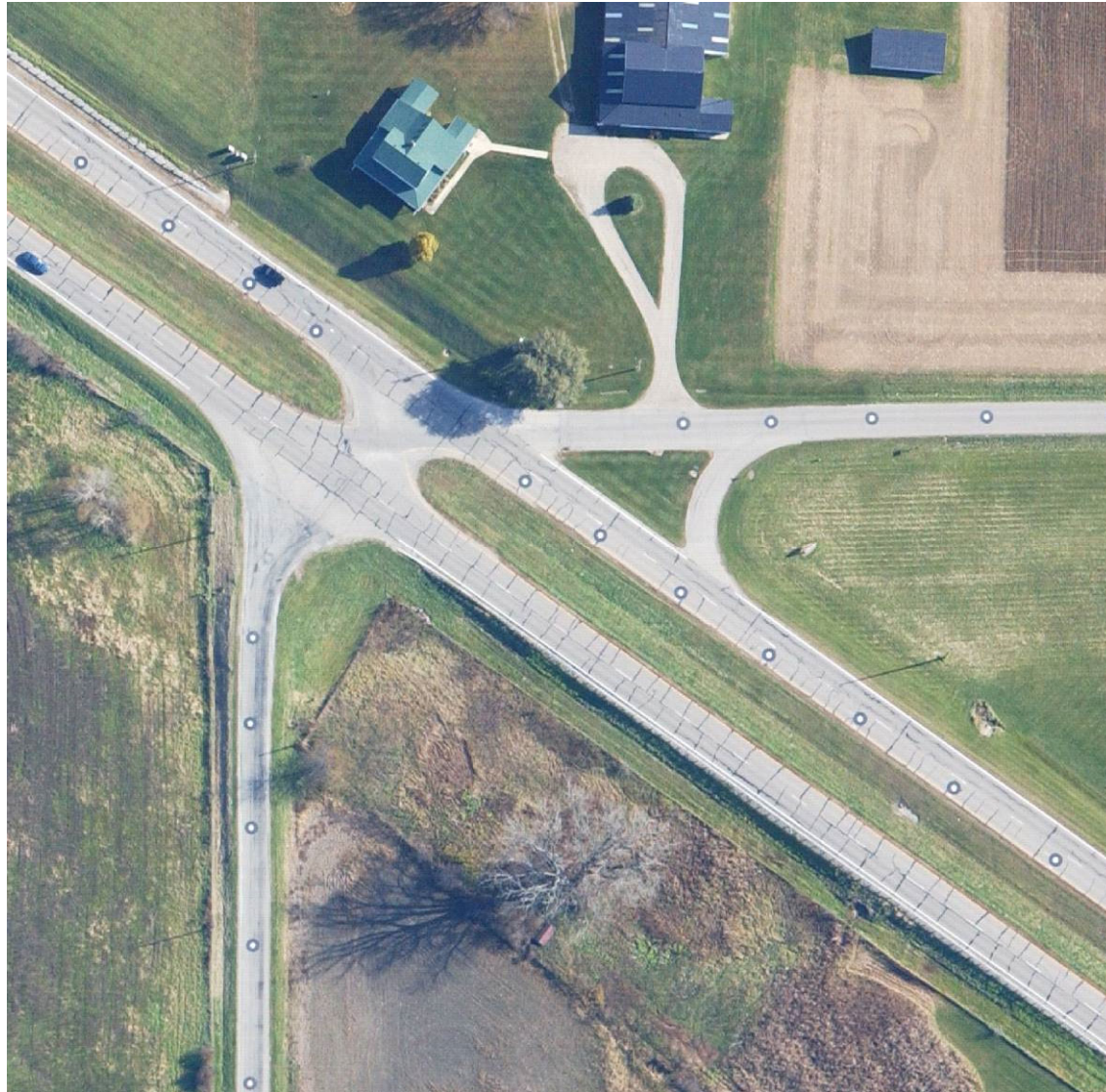
Laboratory Calibration: Multi-Collimators



Indoor Camera Calibration

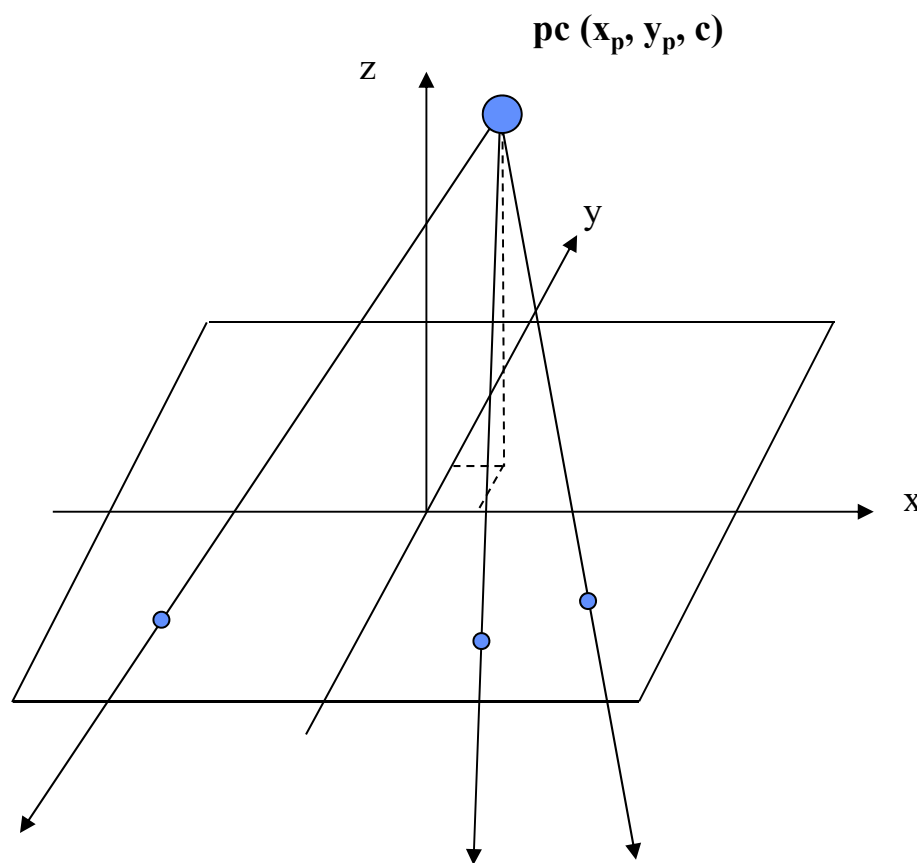


In-Situ Camera Calibration



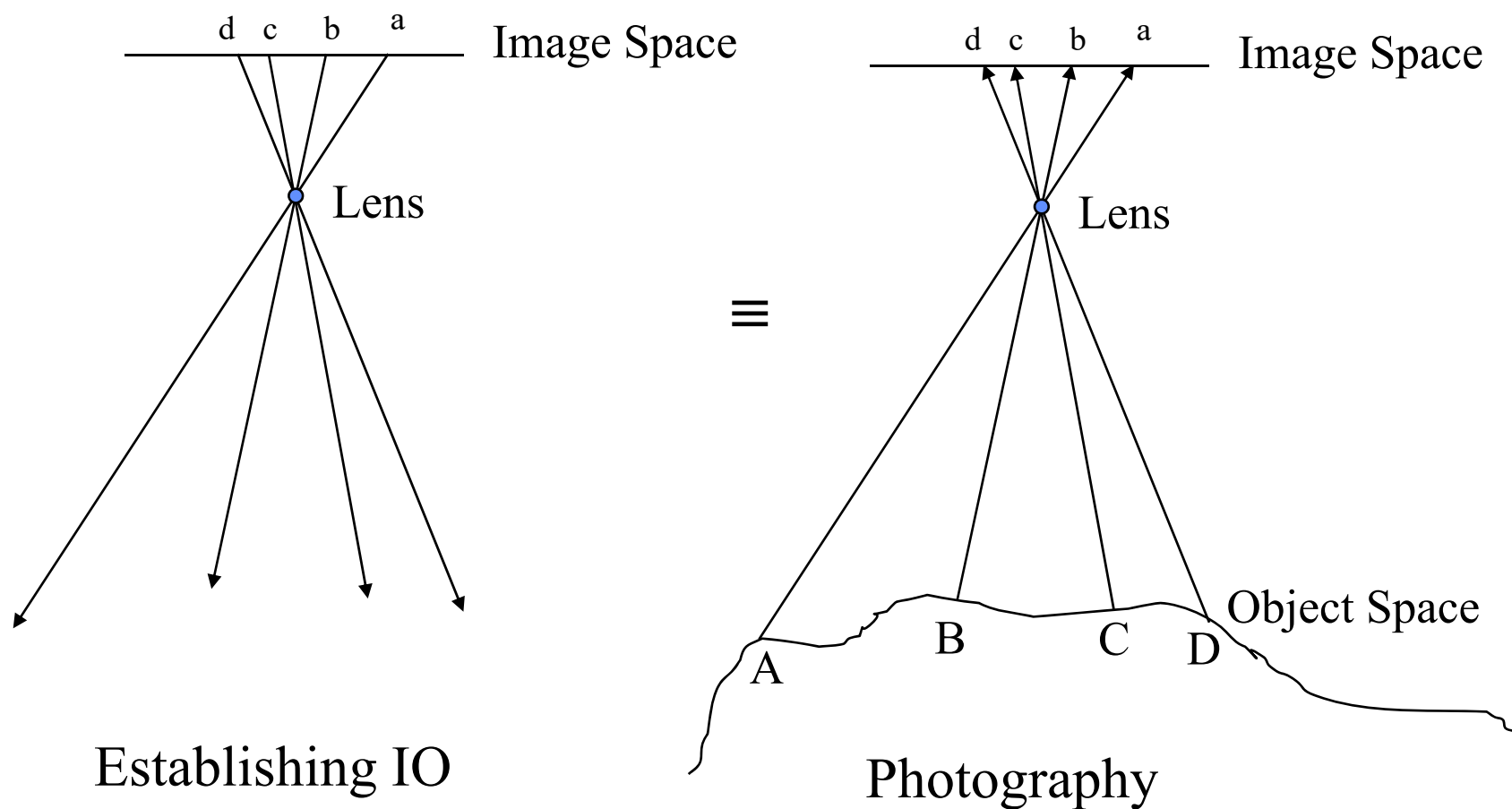
Interior Orientation Parameters

- IOPs together with the image coordinates of selected features define a bundle of light rays (image bundle).



Interior Orientation Parameters

- **Target function of the Interior Orientation:**
 - The defined bundle by the IOPs should be as similar as possible to the incident bundle onto the camera at the moment of exposure.





Photogrammetric Orientation

Exterior Orientation

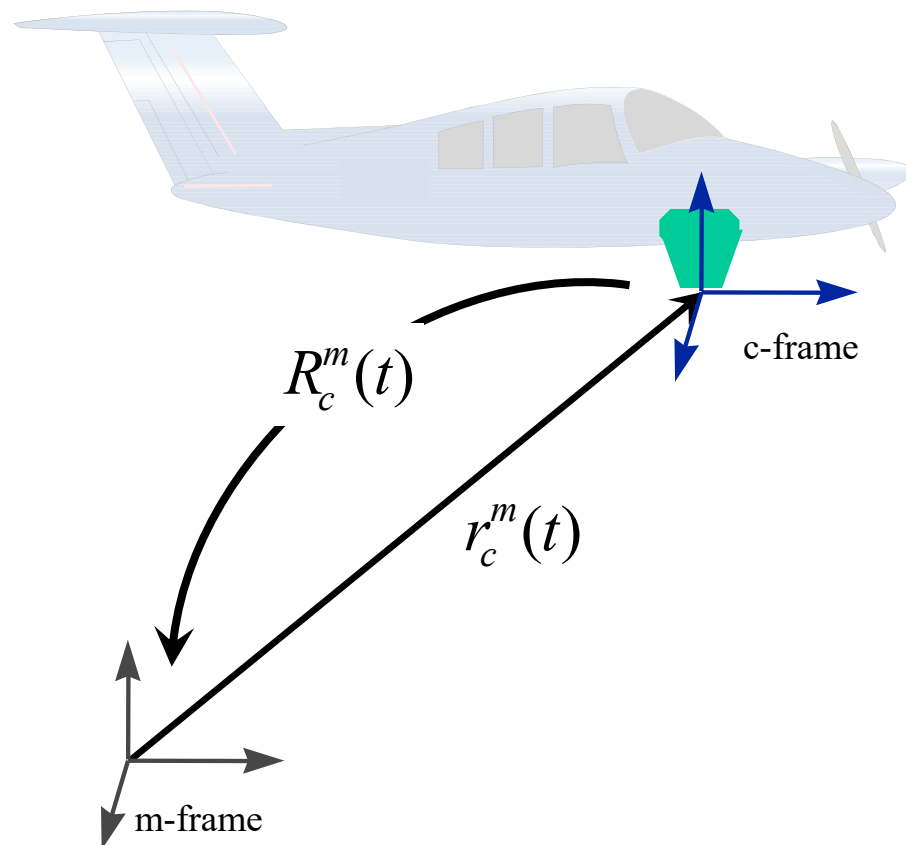


Exterior Orientation Parameters

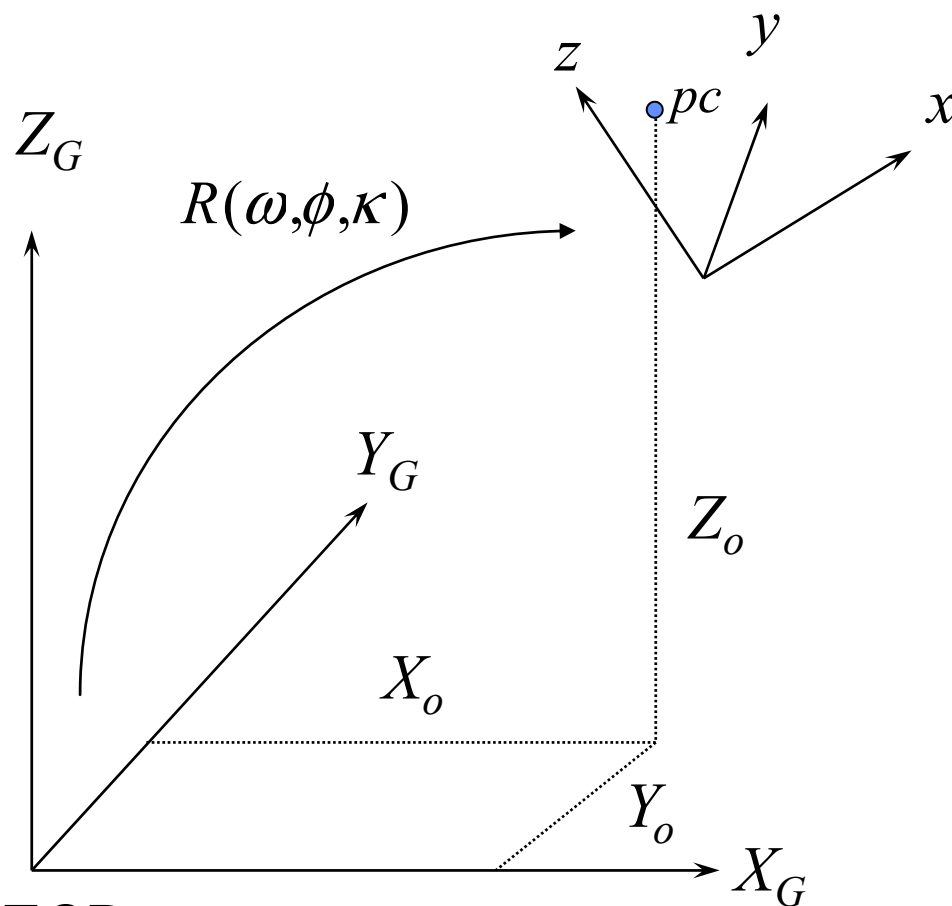
- Exterior Orientation Parameters (EOPs) – **georeferencing parameters** – define the position and the attitude of the image bundle relative to the ground coordinate system.
 - The position of the bundle is defined by (X_o, Y_o, Z_o) .
 - The attitude of the bundle (camera/image coordinate system) relative to the ground coordinate system is defined by the rotation angles (ω, ϕ, κ) .
- EOPs can be either:
 - Indirectly estimated using Ground Control Points (GCPs), or
 - Directly derived using GNSS/INS units onboard the imaging platform.

Exterior Orientation Parameters

- Exterior Orientation Parameters (EOPs) define the position, $r_c^m(t)$, and orientation $R_c^m(t)$, of the camera coordinate system relative to the mapping reference frame at the moment of exposure.



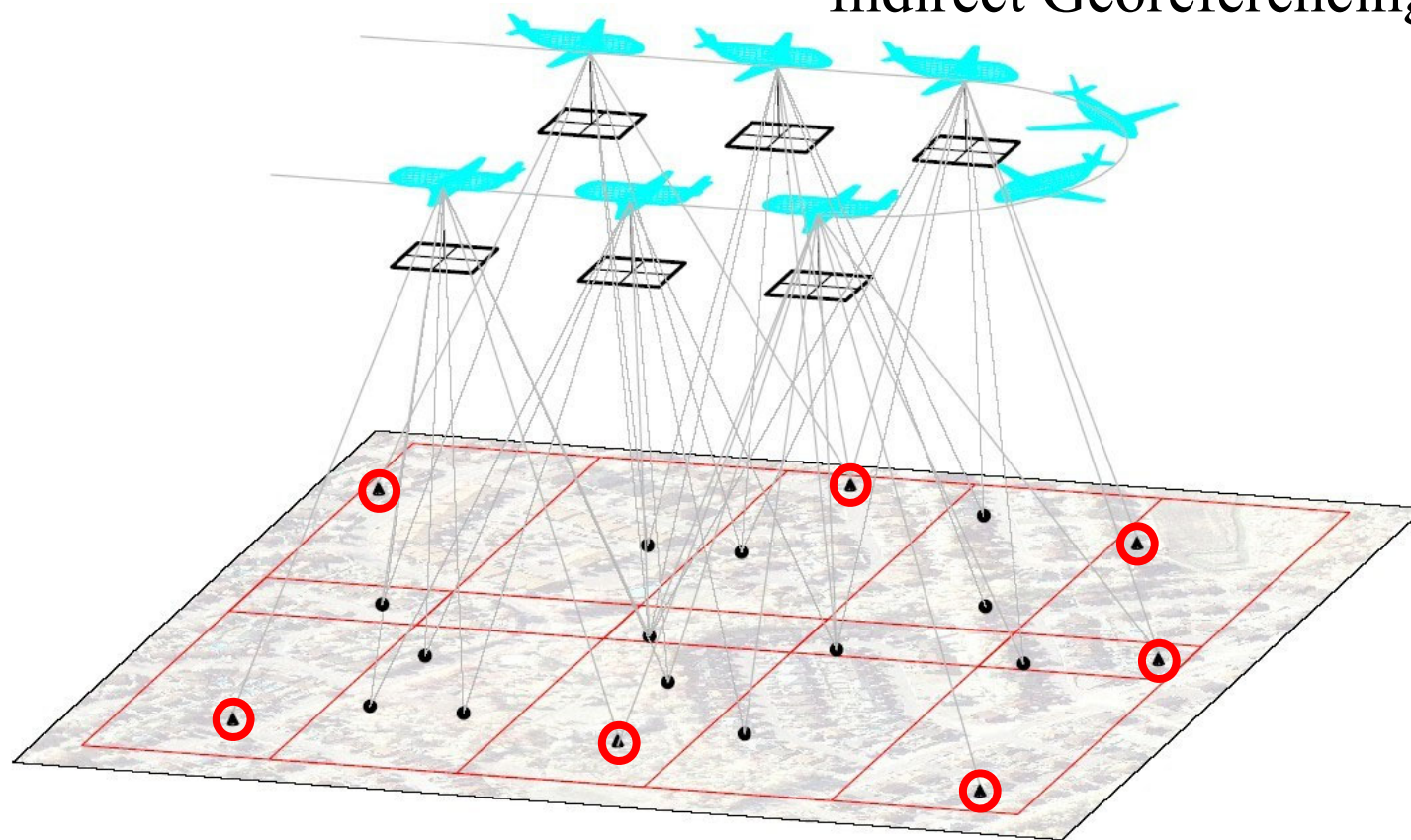
Exterior Orientation Parameters



- EOPs:
 - Indirectly estimated (indirect georeferencing), or
 - Directly derived (direct georeferencing)

Exterior Orientation Parameters

Indirect Georeferencing



-  Ground Control Points
-  Tie Points

Exterior Orientation Parameters



Indirect Georeferencing



Signalized Targets

Exterior Orientation Parameters



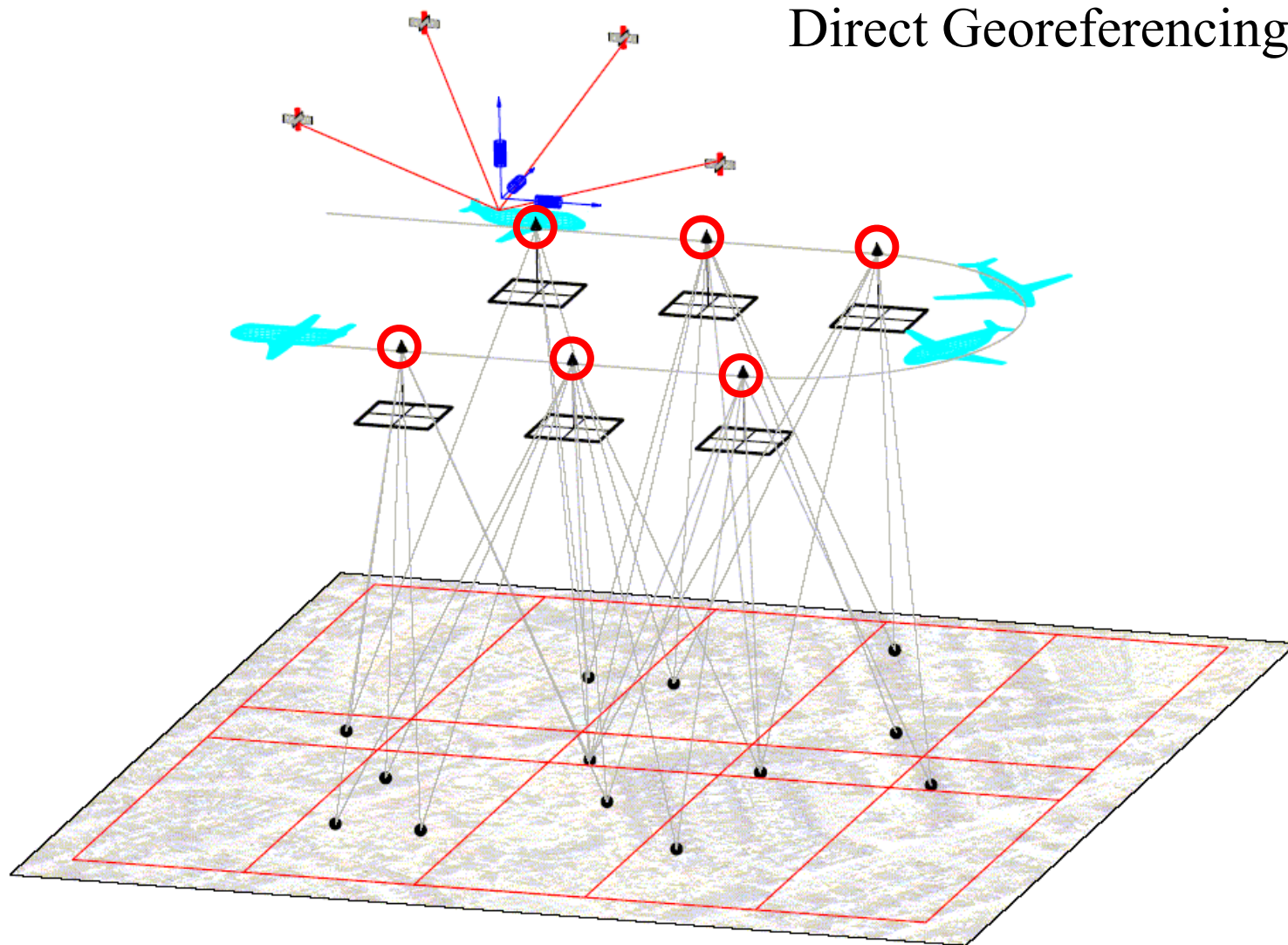
Indirect Georeferencing



Natural Targets

Exterior Orientation Parameters

Direct Georeferencing



Exterior Orientation Parameters

Direct Georeferencing



GNSS Antenna

INS

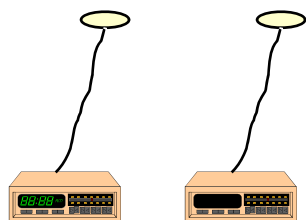
PC



Two Base Stations

Camera





GNSS Receiver






Direct georeferencing in practice

Exterior Orientation Parameters

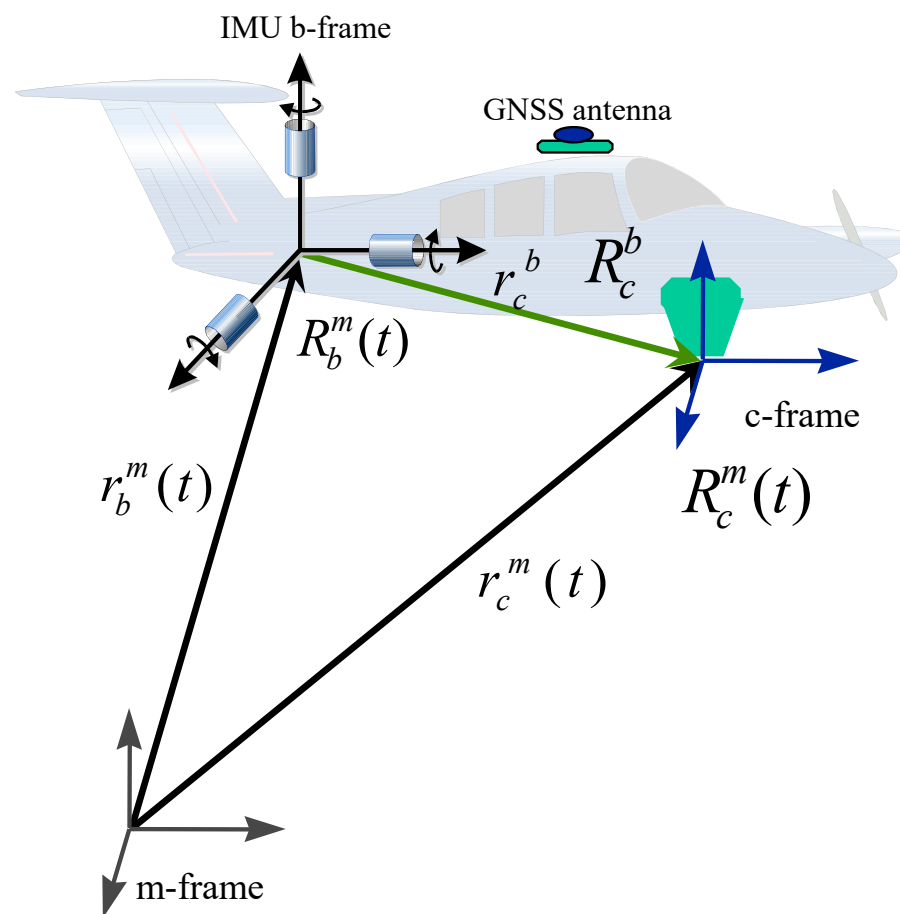
$$r_c^m(t) = r_b^m(t) + R_b^m(t) r_c^b$$

 Camera position
  GNSS/INS position
  GNSS/INS attitude
  Calibration

$$R_c^m(t) = R_b^m(t) R_c^b$$

 Camera attitude
  GNSS/INS attitude
  Calibration

Direct Georeferencing





Photogrammetric Mathematical Model

Collinearity Equations

Vector Summation Based Point Positioning



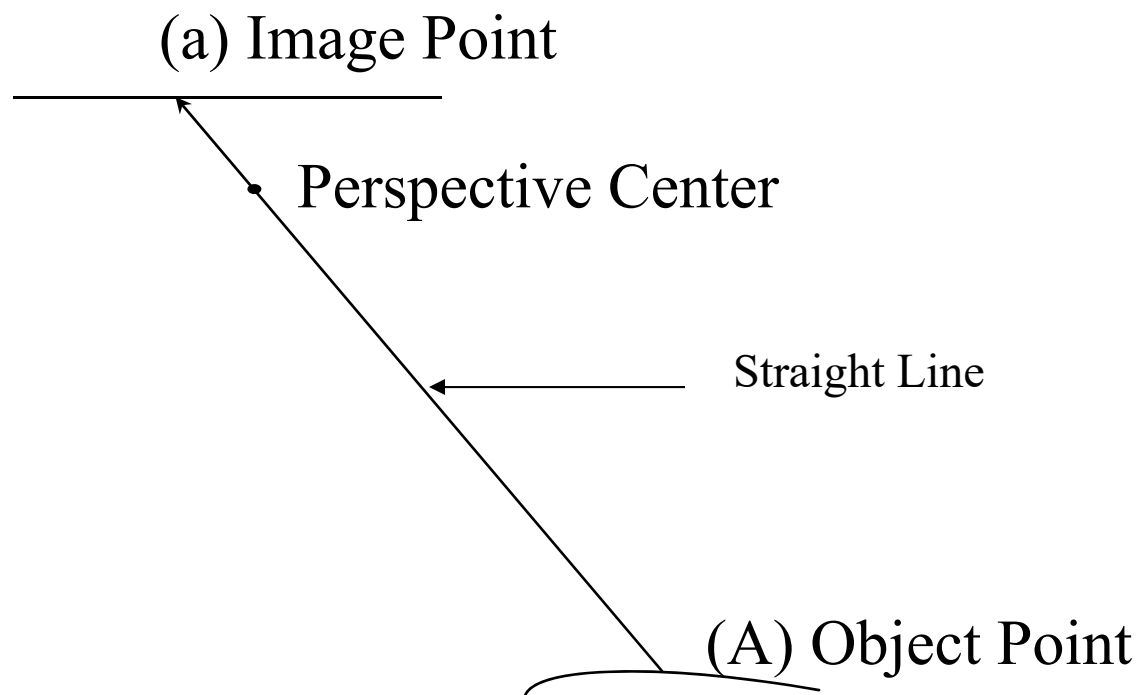
Collinearity Equations

- Objective:
 - Mathematically represent the general relationship between image and ground coordinates

- Concept:
 - Image Point, Object Point, and the Perspective Center are collinear



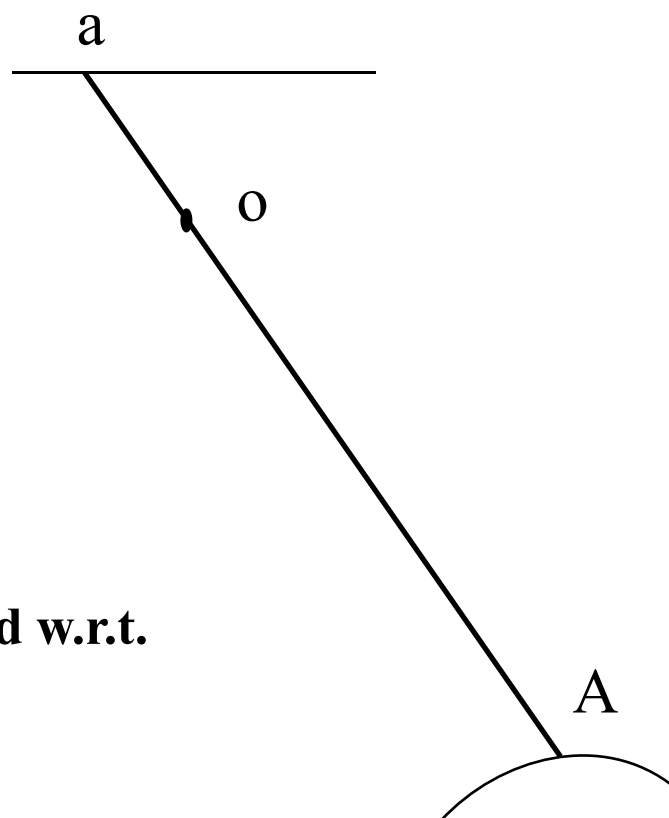
Collinearity Equations





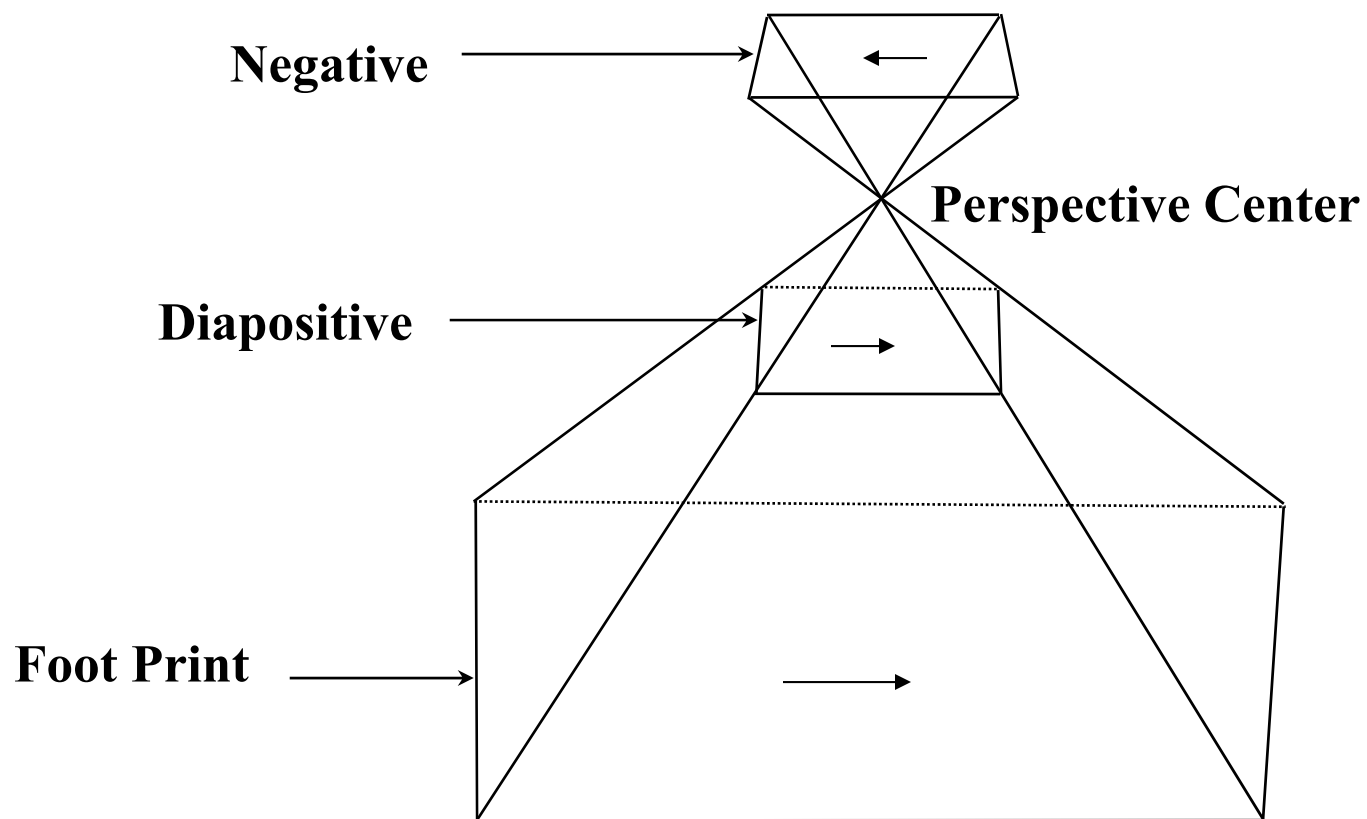
Collinearity Equations

$$\vec{oa} = \lambda \vec{oA}$$



**These vectors should be defined w.r.t.
the same coordinate system**

Frame Camera



Negative Versus Diapositive Films

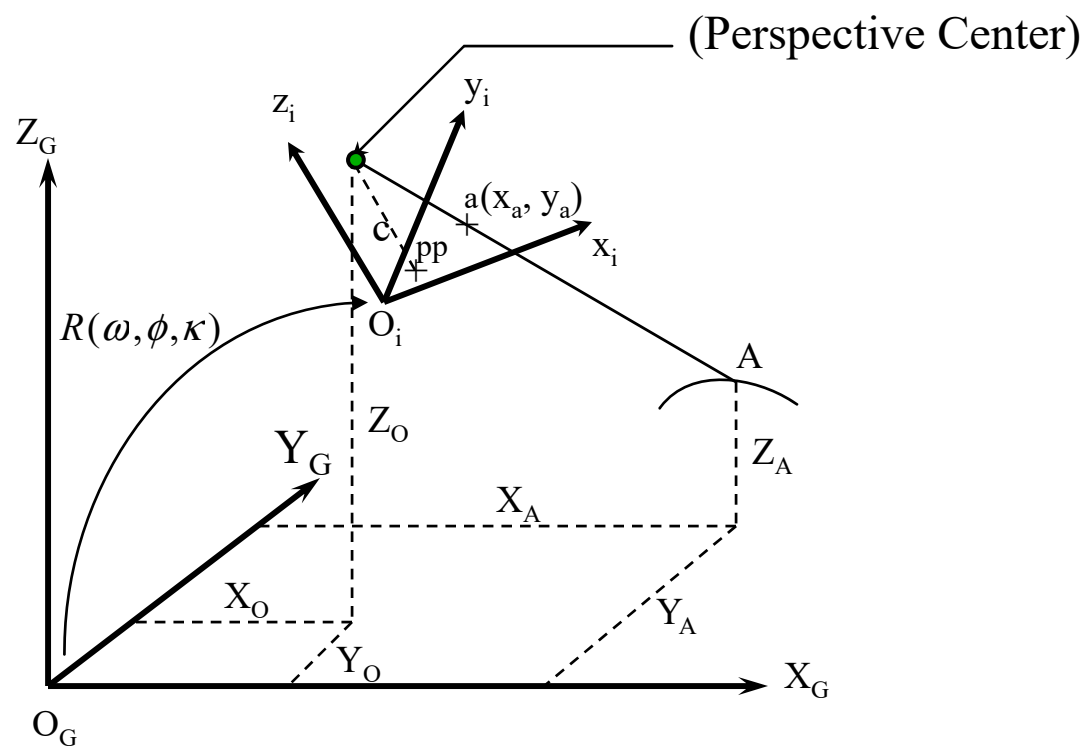


Negative Film

Diapositive

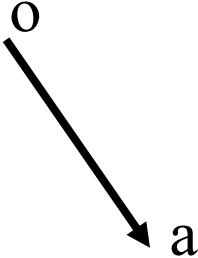


Collinearity Equations





The Vector Connecting the Perspective Center to the Image Point



A diagram showing a perspective center 'o' and an image point 'a'. A black arrow points from 'o' to 'a', representing the vector \vec{v}_i .

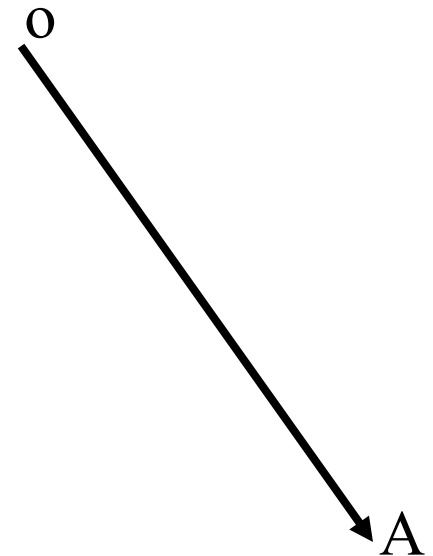
$$\vec{v}_i = r_{oa}^c = \begin{bmatrix} x_a - dist_x \\ y_a - dist_y \\ 0 \end{bmatrix} - \begin{bmatrix} x_p \\ y_p \\ c \end{bmatrix} = \begin{bmatrix} x_a - x_p - dist_x \\ y_a - y_p - dist_y \\ -c \end{bmatrix}$$

w.r.t. the image coordinate system



The Vector Connecting the Perspective Center to the Object Point

$$\vec{V}_o = r_{oA}^m = \begin{bmatrix} X_A \\ Y_A \\ Z_A \end{bmatrix} - \begin{bmatrix} X_o \\ Y_o \\ Z_o \end{bmatrix} = \begin{bmatrix} X_A - X_o \\ Y_A - Y_o \\ Z_A - Z_o \end{bmatrix}$$



w.r.t. the ground coordinate system



Collinearity Equations

$$\vec{oa} = \lambda \vec{oA}$$

$$\vec{v}_i = r_{oa}^c = \lambda M(\omega, \varphi, \kappa) \vec{V}_o = \lambda R_m^c r_{oA}^m$$

$$\begin{bmatrix} x_a - x_p - dist_x \\ y_a - y_p - dist_y \\ -c \end{bmatrix} = \lambda \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix} \begin{bmatrix} X_A - X_o \\ Y_A - Y_o \\ Z_A - Z_o \end{bmatrix}$$

Where: λ is a scale factor

- Questions:
 - Can you come up with an average estimate of λ ?
 - Is λ constant for a given image? Why?



Collinearity Equations

$$M = R_m^c$$

$$x_a = x_p - c \frac{m_{11}(X_A - X_o) + m_{12}(Y_A - Y_o) + m_{13}(Z_A - Z_o)}{m_{31}(X_A - X_o) + m_{32}(Y_A - Y_o) + m_{33}(Z_A - Z_o)} + dist_x$$

$$y_a = y_p - c \frac{m_{21}(X_A - X_o) + m_{22}(Y_A - Y_o) + m_{23}(Z_A - Z_o)}{m_{31}(X_A - X_o) + m_{32}(Y_A - Y_o) + m_{33}(Z_A - Z_o)} + dist_y$$

$$R = R_c^m$$

$$x_a = x_p - c \frac{r_{11}(X_A - X_o) + r_{21}(Y_A - Y_o) + r_{31}(Z_A - Z_o)}{r_{13}(X_A - X_o) + r_{23}(Y_A - Y_o) + r_{33}(Z_A - Z_o)} + dist_x$$

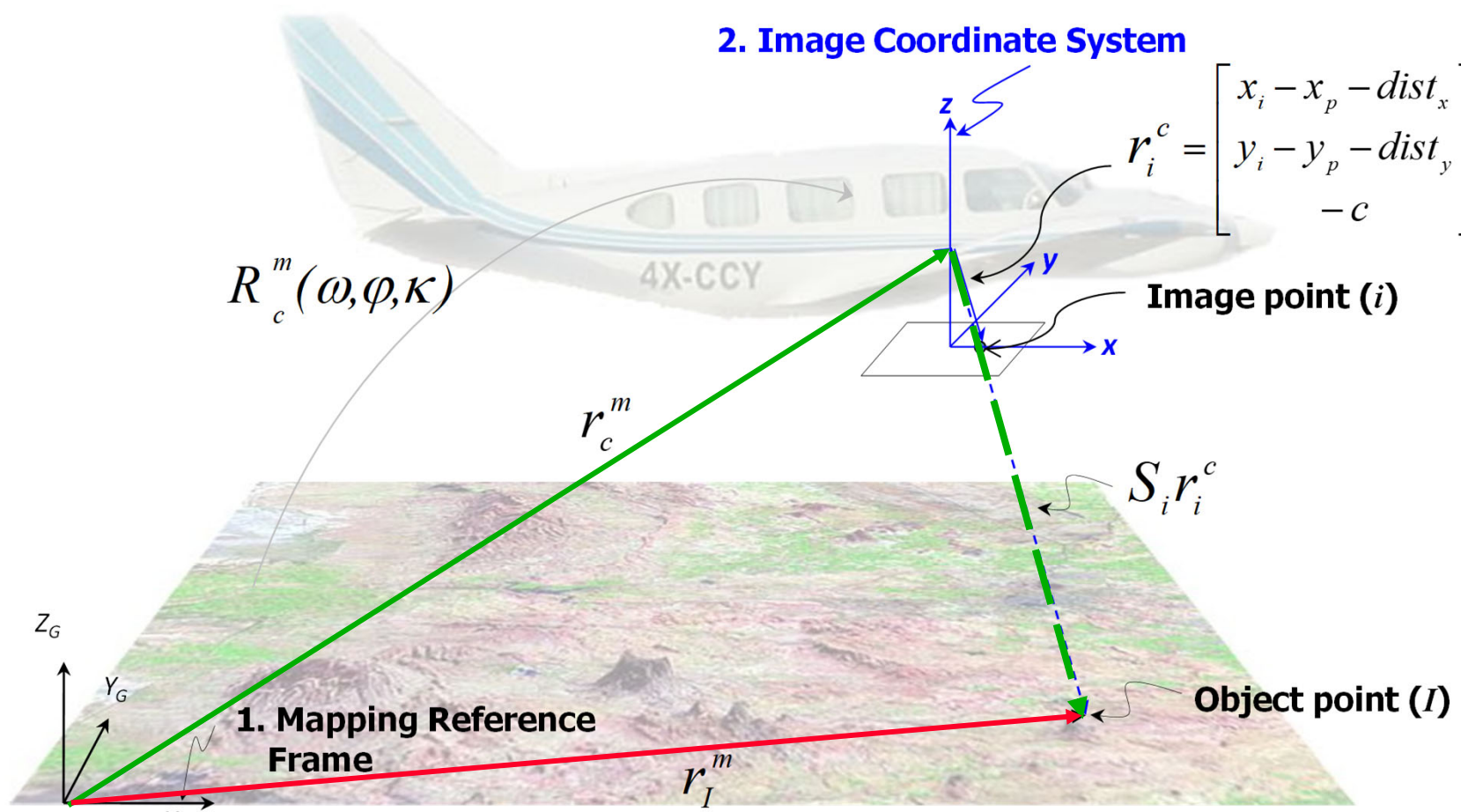
$$y_a = y_p - c \frac{r_{12}(X_A - X_o) + r_{22}(Y_A - Y_o) + r_{32}(Z_A - Z_o)}{r_{13}(X_A - X_o) + r_{23}(Y_A - Y_o) + r_{33}(Z_A - Z_o)} + dist_y$$

Collinearity Equations

$$r_I^m = r_c^m + S_i R_c^m(\omega, \phi, \kappa) r_i^c$$

2. Image Coordinate System

$$r_i^c = \begin{bmatrix} x_i - x_p - dist_x \\ y_i - y_p - dist_y \\ -c \end{bmatrix}$$



Vector Summation Procedure



Collinearity Equations

$$r_I^m = r_c^m + S_i R_c^m(\omega, \phi, \kappa) r_i^c$$

$$\begin{bmatrix} X_G \\ Y_G \\ Z_G \end{bmatrix} = \begin{bmatrix} X_o \\ Y_o \\ Z_o \end{bmatrix} + S_i R_c^m(\omega, \phi, \kappa) \begin{bmatrix} x_i - x_p - dist_{x_i} \\ y_i - y_p - dist_{y_i} \\ -c \end{bmatrix}$$

$$\begin{bmatrix} x_i - x_p - dist_{x_i} \\ y_i - y_p - dist_{y_i} \\ -c \end{bmatrix} = \frac{1}{S_i} R_c^m(\omega, \phi, \kappa) [\vec{X}_G - \vec{X}_o] = \frac{1}{S_i} \begin{bmatrix} N_x \\ N_y \\ D \end{bmatrix}$$

$$x_i = x_p - c \frac{N_x}{D} + dist_{x_i}$$

$$y_i = y_p - c \frac{N_y}{D} + dist_{y_i}$$

Vector Summation Procedure



Collinearity Equations

$$R = R_c^m$$

$$x_a = x_p - c \frac{r_{11}(X_A - X_O) + r_{21}(Y_A - Y_O) + r_{31}(Z_A - Z_O)}{r_{13}(X_A - X_O) + r_{23}(Y_A - Y_O) + r_{33}(Z_A - Z_O)}$$

$$y_a = y_p - c \frac{r_{12}(X_A - X_O) + r_{22}(Y_A - Y_O) + r_{32}(Z_A - Z_O)}{r_{13}(X_A - X_O) + r_{23}(Y_A - Y_O) + r_{33}(Z_A - Z_O)}$$

- Involved parameters:
 - Image coordinates (x_a, y_a)
 - Ground coordinates (X_A, Y_A, Z_A)
 - Exterior Orientation Parameters ($X_O, Y_O, Z_O, \omega, \phi, \kappa$)
 - Interior Orientation Parameters (x_p, y_p, c , distortion parameters)

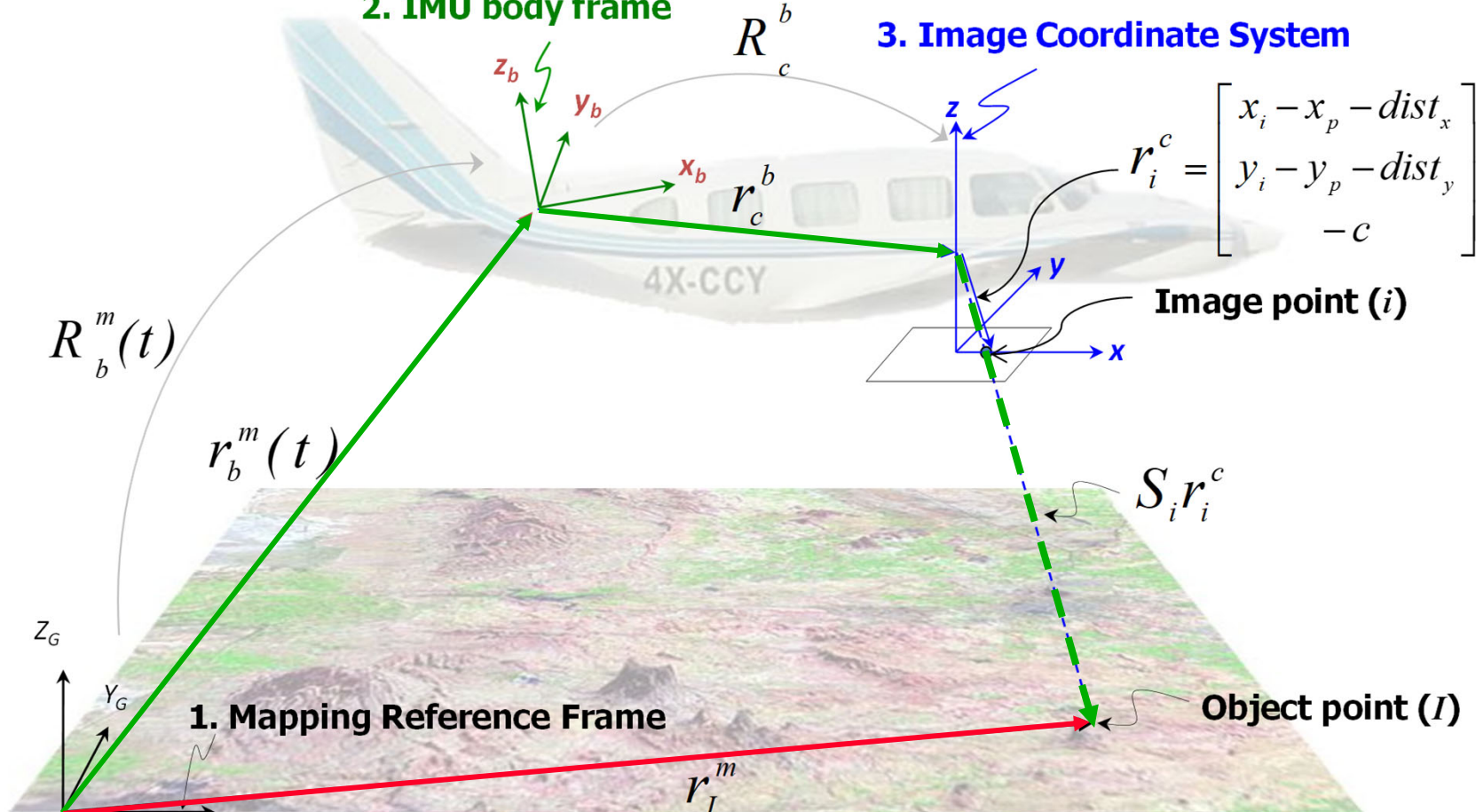
Photogrammetric Point Positioning

GNSS/INS-Assisted Photogrammetric System:

$$r_I^m = r_b^m(t) + R_b^m(t) r_c^b + S_i R_b^m(t) R_c^b r_i^c$$

2. IMU body frame

3. Image Coordinate System



$$r_i^c = \begin{bmatrix} x_i - x_p - dist_x \\ y_i - y_p - dist_y \\ -c \end{bmatrix}$$

Image point (i)

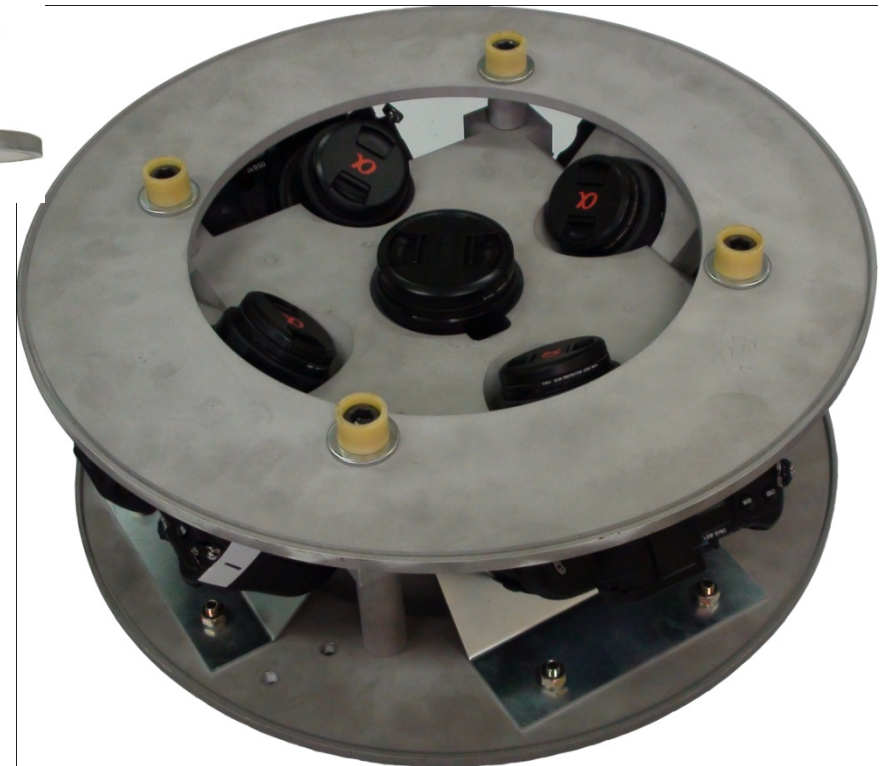
Object point (I)

Direct Georeferencing - ISO

Photogrammetric Point Positioning



Multi-Camera Photogrammetric Systems:



Multi-Camera Systems

A rigid-relationship among the cameras

Airborne Mobile Mapping System

Photogrammetric Point Positioning



Multi-Camera Photogrammetric Systems:

Multi-Camera Systems

A rigid-relationship among the cameras

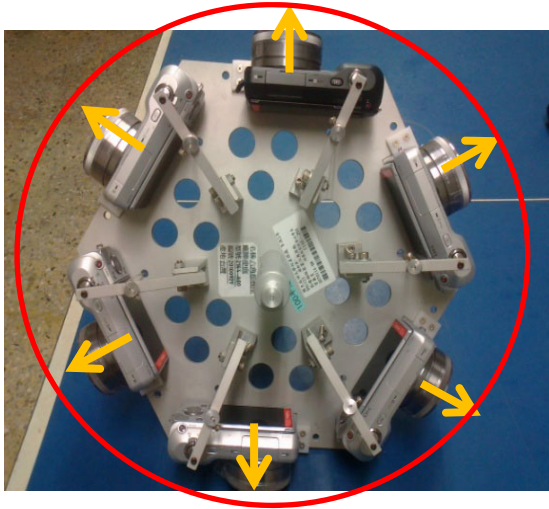
Terrestrial Mobile Mapping System



Photogrammetric Point Positioning



Multi-Camera Photogrammetric Systems:



Multi-Camera Systems

A rigid-relationship among the cameras

Portable Panoramic Image Mapping System



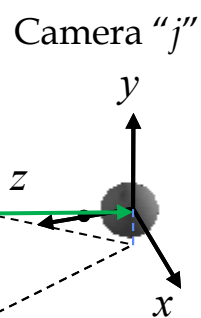
Photogrammetric Point Positioning

GNSS/INS-Assisted Multi-Camera Photogrammetric System:

$$r_I^m = r_b^m(t) + R_b^m(t) r_{c_r}^b + R_b^m(t) R_{c_r}^b r_{c_j}^{c_r} + S_i^{c_j} R_b^m(t) R_{c_r}^b R_{c_j}^{c_r} r_i^{c_j}$$

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix}_b^m (t)_{GNSS/INS} = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}_b^m (t) + \begin{bmatrix} e_X \\ e_Y \\ e_Z \end{bmatrix}_b^m (t)$$

$$\begin{bmatrix} \omega \\ \phi \\ \kappa \end{bmatrix}_b^m (t)_{GNSS/INS} = \begin{bmatrix} \omega \\ \phi \\ \kappa \end{bmatrix}_b^m (t) + \begin{bmatrix} e_\omega \\ e_\phi \\ e_\kappa \end{bmatrix}_b^m (t)$$

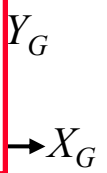


$$\begin{bmatrix} \Delta X \\ \Delta Y \\ \Delta Z \end{bmatrix}_{c_j}^{cr} (prior) = \begin{bmatrix} \Delta X \\ \Delta Y \\ \Delta Z \end{bmatrix}_{c_j}^{cr} + \begin{bmatrix} e_{\Delta X} \\ e_{\Delta Y} \\ e_{\Delta Z} \end{bmatrix}_{c_j}^{cr}$$

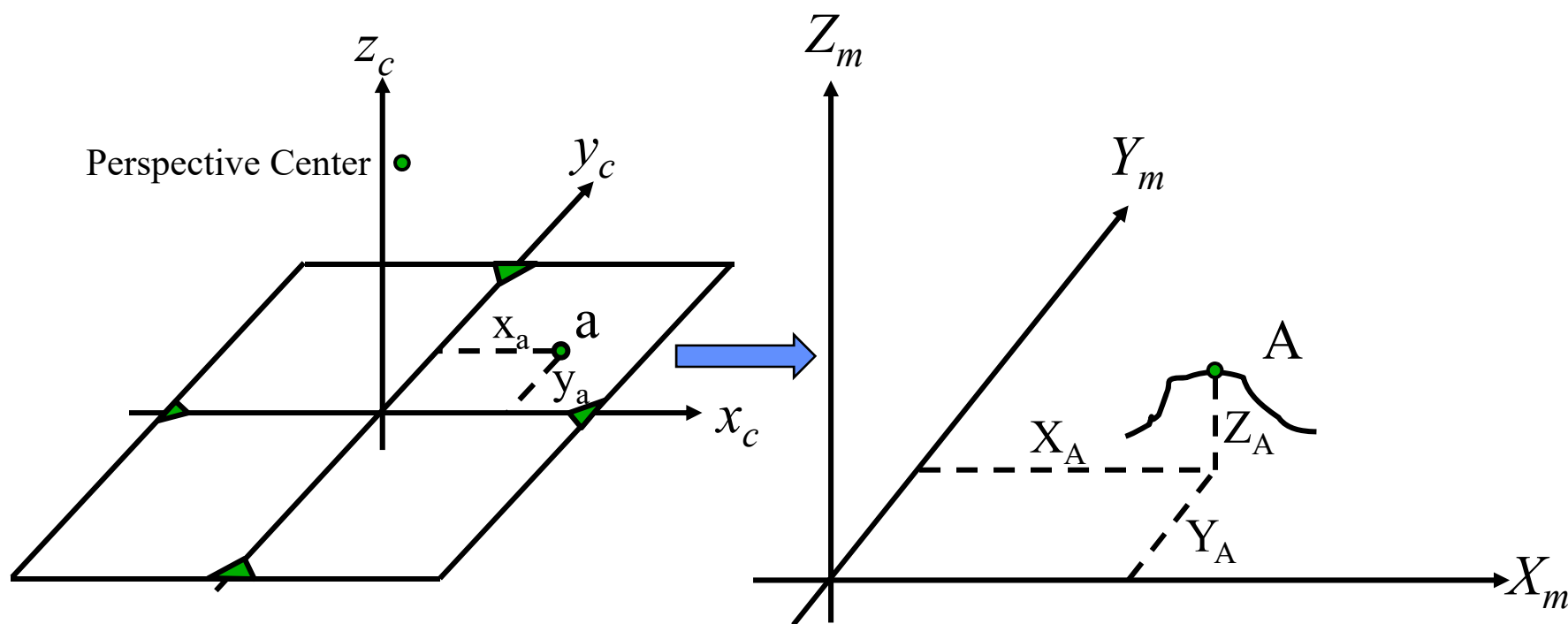
$$\begin{bmatrix} \Delta \omega \\ \Delta \phi \\ \Delta \kappa \end{bmatrix}_{c_j}^{cr} (prior) = \begin{bmatrix} \Delta \omega \\ \Delta \phi \\ \Delta \kappa \end{bmatrix}_{c_j}^{cr} + \begin{bmatrix} e_{\Delta \omega} \\ e_{\Delta \phi} \\ e_{\Delta \kappa} \end{bmatrix}_{c_j}^{cr}$$

$$\begin{bmatrix} \Delta X \\ \Delta Y \\ \Delta Z \end{bmatrix}_{c_r}^b (prior) = \begin{bmatrix} \Delta X \\ \Delta Y \\ \Delta Z \end{bmatrix}_{c_r}^b + \begin{bmatrix} e_{\Delta X} \\ e_{\Delta Y} \\ e_{\Delta Z} \end{bmatrix}_{c_r}^b$$

$$\begin{bmatrix} \Delta \omega \\ \Delta \phi \\ \Delta \kappa \end{bmatrix}_{c_r}^b (prior) = \begin{bmatrix} \Delta \omega \\ \Delta \phi \\ \Delta \kappa \end{bmatrix}_{c_r}^b + \begin{bmatrix} e_{\Delta \omega} \\ e_{\Delta \phi} \\ e_{\Delta \kappa} \end{bmatrix}_{c_r}^b$$



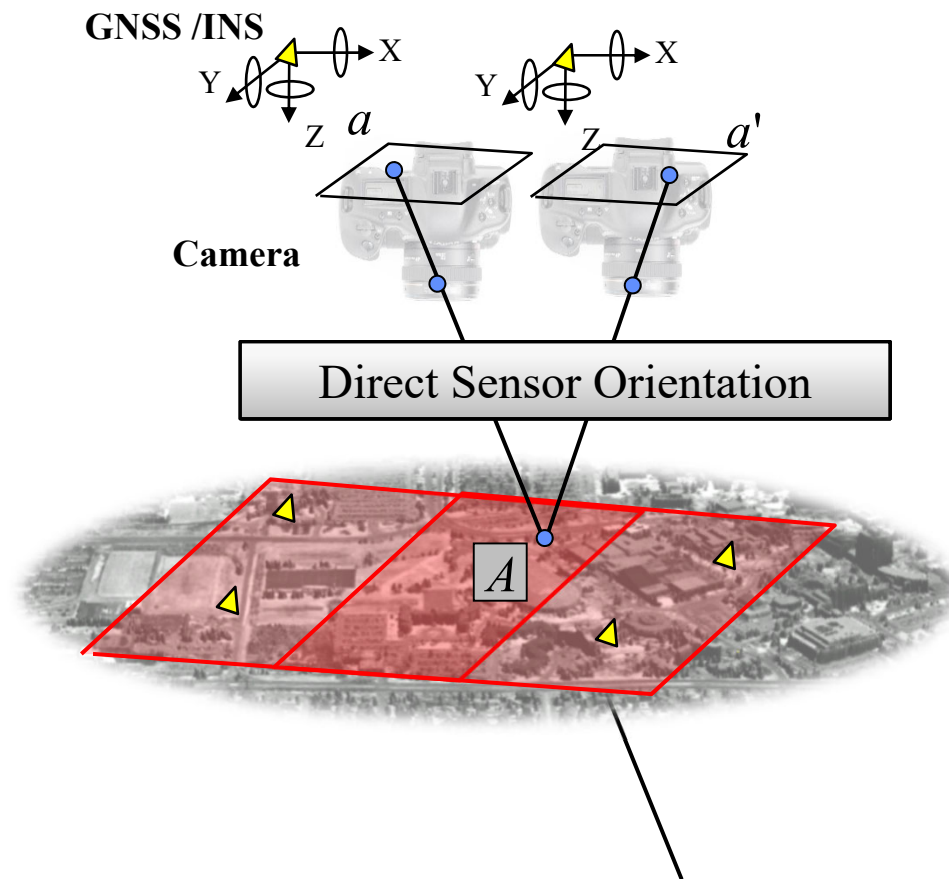
Photogrammetric Point Positioning



$$x_a = f_x (X_A, Y_A, Z_A, IOPs, EOPs)$$

$$y_a = f_y (X_A, Y_A, Z_A, IOPs, EOPs)$$

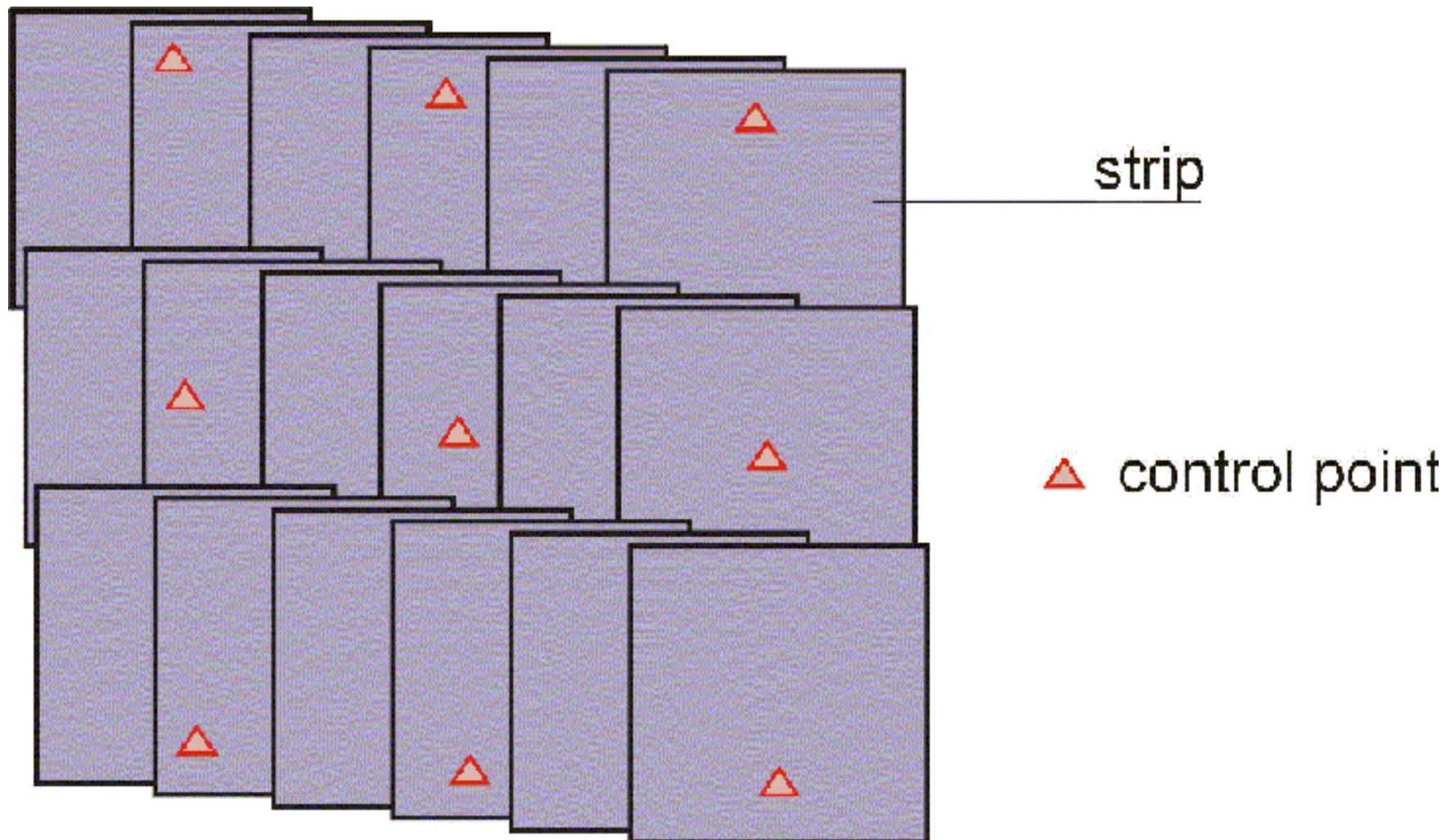
Photogrammetric Point Positioning





Bundle Block Adjustment

Bundle Block Adjustment



60% Overlap and 20% side lap



Bundle Block Adjustment

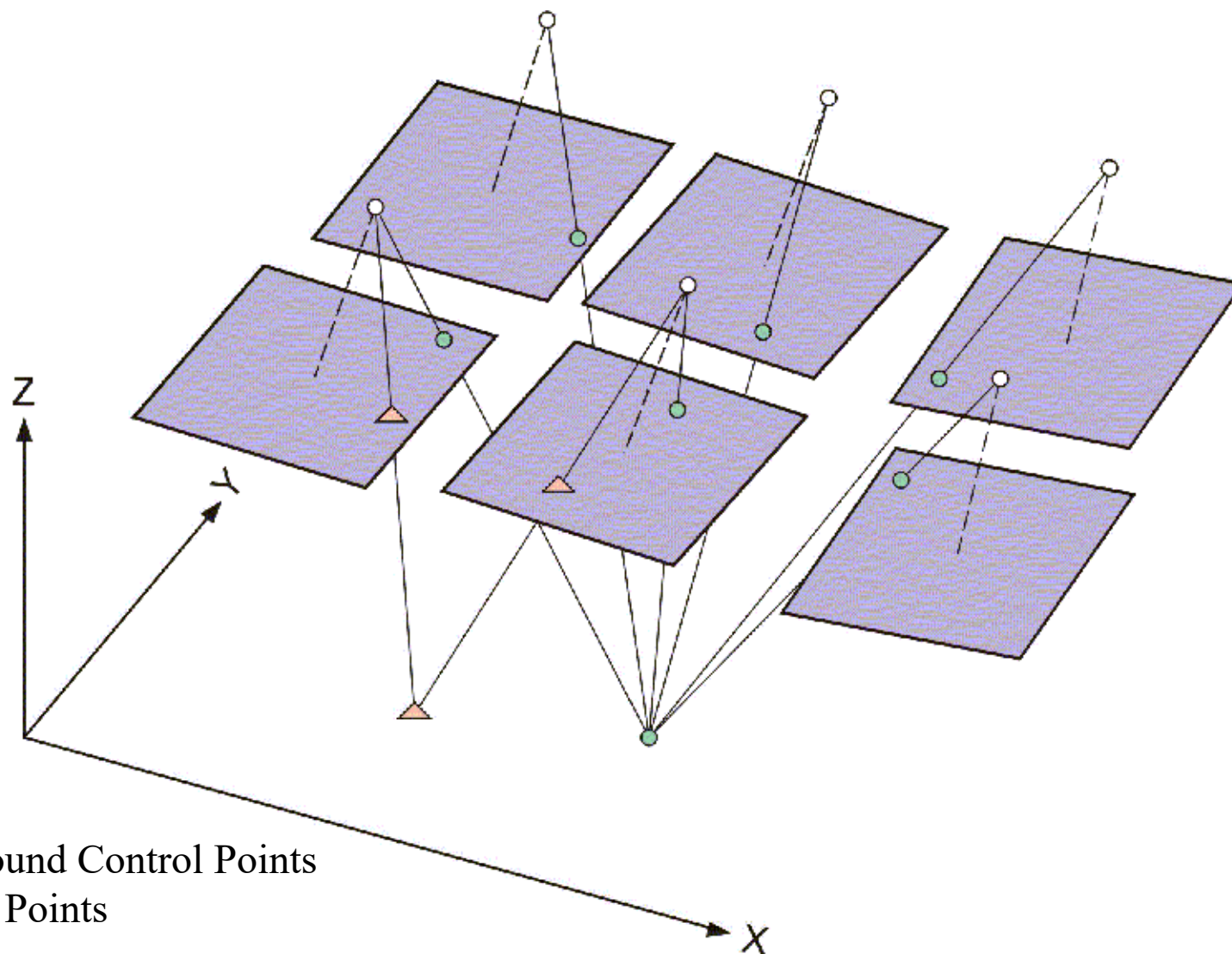
- Direct relationship between image and ground coordinates
- We measure the image coordinates in the images of the block.
- Using the collinearity equations, we can relate the image coordinates, corresponding ground coordinates, IOPs, and EOPs.
- Using a simultaneous least squares adjustment, we can solve for the:
 - Ground coordinates of tie points,
 - EOPs, and
 - IOPs (Camera Calibration Procedure).

Bundle Block Adjustment: Concept



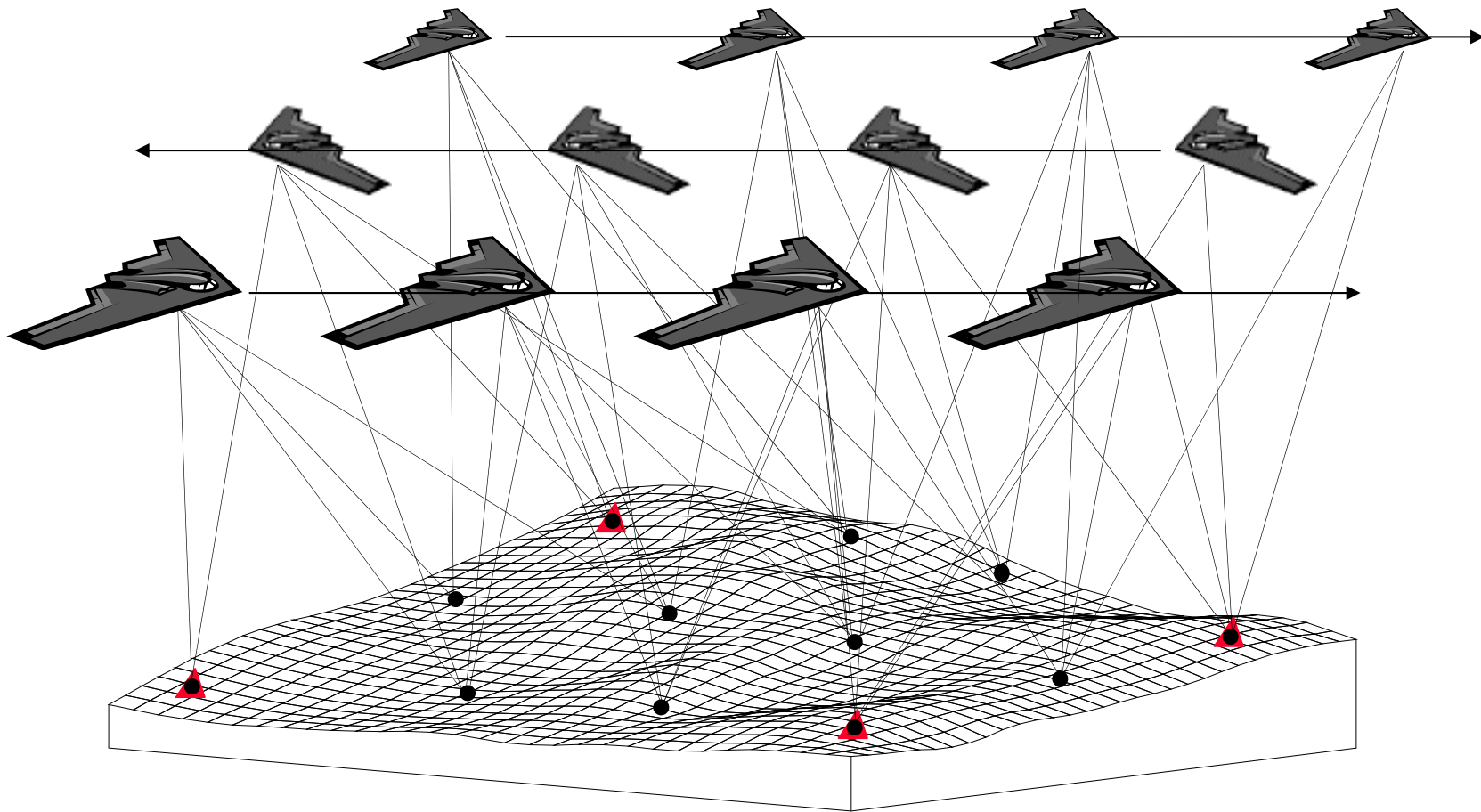
- The image coordinate measurements and IOPs define a bundle of light rays.
- The EOPs define the position and attitude of the bundles in space.
- During the adjustment: The bundles are rotated (ω , ϕ , κ) and shifted (X_o , Y_o , Z_o) until:
 - Conjugate light rays intersect as well as possible at the locations of object space tie points.
 - Light rays corresponding to ground control points pass through the object points as close as possible.

Bundle Block Adjustment: Concept



- ▲ Ground Control Points
- Tie Points

Bundle Block Adjustment: Concept



- ▲ Ground Control Points
- Tie Points



Least Squares Adjustment

- Prior to the adjustment, we need to identify:
 - The unknown parameters
 - Observable quantities
 - The mathematical relationship between the unknown parameters and the observable quantities
- Linearize the mathematical relationship (if it is not linear)
- Apply least squares adjustment formulas



Unknown Parameters

- Unknown parameters might include:
 - Ground coordinates of tie points (**points that appear in more than one image**)
 - Exterior orientation parameters of the involved imagery
 - Interior orientation parameters of the involved cameras (**for camera calibration purposes**)



Observable Quantities

- Observable quantities might include:
 - The ground coordinates of control points
 - Image coordinates of tie as well as control points
 - Interior orientation parameters of the involved cameras
 - Exterior orientation parameters of the involved imagery (**from a GNSS/INS unit onboard**)



Mathematical Model

$$x_a = x_p - c \frac{r_{11}(X_A - X_O) + r_{21}(Y_A - Y_O) + r_{31}(Z_A - Z_O)}{r_{13}(X_A - X_O) + r_{23}(Y_A - Y_O) + r_{33}(Z_A - Z_O)} + \Delta x + e_x$$

$$y_a = y_p - c \frac{r_{12}(X_A - X_O) + r_{22}(Y_A - Y_O) + r_{32}(Z_A - Z_O)}{r_{13}(X_A - X_O) + r_{23}(Y_A - Y_O) + r_{33}(Z_A - Z_O)} + \Delta y + e_y$$

$$\begin{bmatrix} e_x \\ e_y \end{bmatrix} \sim (0, \sigma_o^2 P^{-1})$$



Mathematical Model

- $\Delta x = \Delta x_{\text{Radial Lens Distortion}} + \Delta x_{\text{Decentric Lens Distortion}} + \Delta x_{\text{Atmospheric Refraction}} + \Delta x_{\text{Affine Deformation}} + \text{etc....}$

- $\Delta y = \Delta y_{\text{Radial Lens Distortion}} + \Delta y_{\text{Decentric Lens Distortion}} + \Delta y_{\text{Atmospheric Refraction}} + \Delta y_{\text{Affine Deformations}} + \text{etc....}$



Distortion Parameters

$$\Delta x_{\text{Radial Lens Distortion}} = \bar{x} (k_1 r^2 + k_2 r^4 + k_3 r^6 + \dots)$$

$$\Delta y_{\text{Radial Lens Distortion}} = \bar{y} (k_1 r^2 + k_2 r^4 + k_3 r^6 + \dots)$$

$$\Delta x_{\text{Decentric Lens Distortion}} = (1 + p_3^2 r^2) \{ p_1 (r^2 + 2\bar{x}^2) + 2p_2 \bar{x} \bar{y} \}$$

$$\Delta y_{\text{Decentric Lens Distortion}} = (1 + p_3^2 r^2) \{ 2p_1 \bar{x} \bar{y} + p_2 (r^2 + 2\bar{y}^2) \}$$

$$\text{where: } r = \{(x - x_p)^2 + (y - y_p)^2\}^{0.5}$$

$$\bar{x} = x - x_p$$

$$\bar{y} = y - y_p$$



Least Squares Adjustment

- Gauss Markov Model
Observation Equations

$$y = A x + e \quad e \sim (0, \sigma_o^2 P^{-1})$$

y $n \times 1$ *observation vector*

A $n \times m$ *design matrix*

x $m \times 1$ *vector of unknowns*

e $n \times 1$ *noise contaminating the observation vector*

$\sigma_o^2 P^{-1}$ $n \times n$ *variance covariance matrix of the noise vector*



Least Squares Adjustment

$$\hat{x} = (A^T P A)^{-1} A^T P y$$

$$D\{\hat{x}\} = \sigma_o^2 (A^T P A)^{-1}$$

$$\tilde{e} = y - A\hat{x}$$

$$\hat{\sigma}_o^2 = (\tilde{e}^T P \tilde{e}) / (n - m)$$



Non-Linear System

$$Y = a(X) + e$$

$a(X)$ is the non – linear function

We use Taylor Series Expansion

$$Y \approx a(X_o) + \left. \frac{\partial a}{\partial X} \right|_{X_o} (X - X_o) + e \quad (\text{We ignore higher order terms})$$

Where :

X_o is approximate values for the unknown parameters

$$Y - a(X_o) = \left. \frac{\partial a}{\partial X} \right|_{X_o} (X - X_o) + e$$

$$y = Ax + e$$

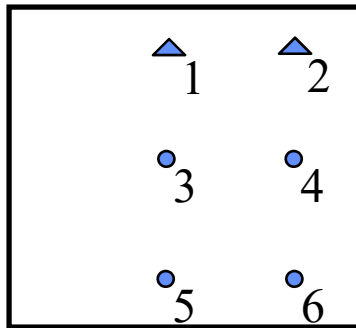
Where :

$$y = Y - a(X_o)$$

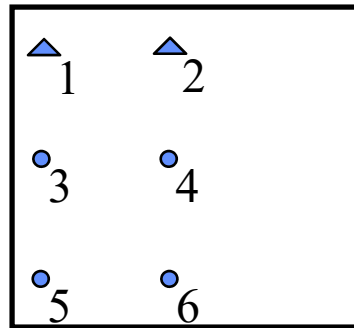
$$A = \left. \frac{\partial a}{\partial X} \right|_{X_o}$$

- Iterative solution for the unknown parameters
- When should we stop the iterations?

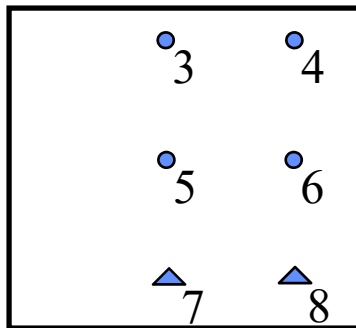
Example (4 Images in Two Strips)



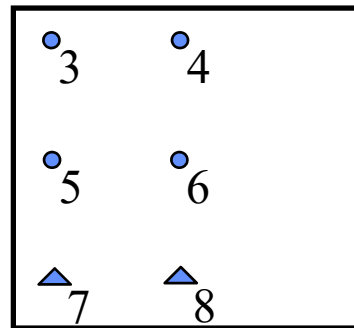
I



II



III



IV

▲ Control Point

● Tie Point



Balance Between Observations & Unknowns

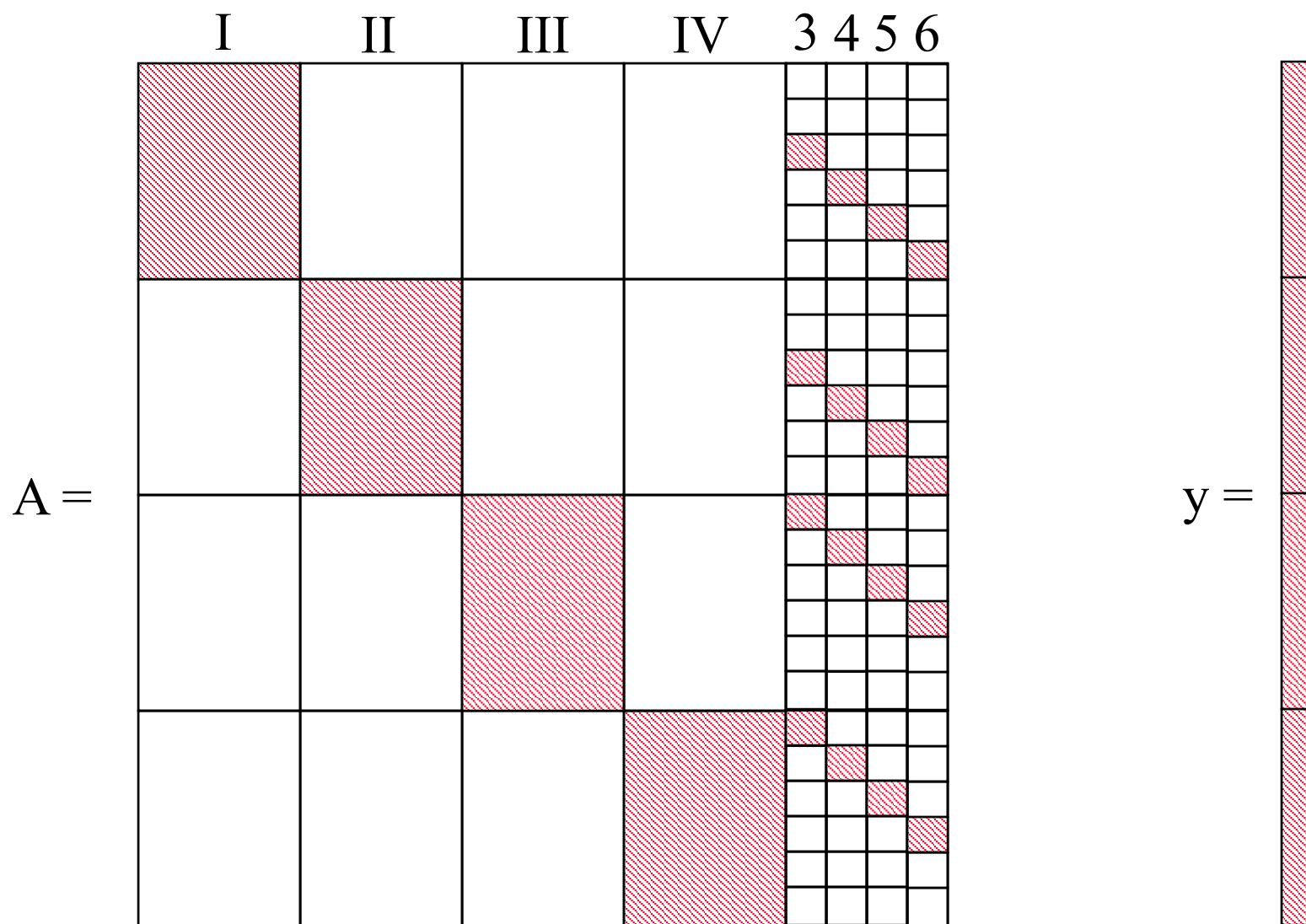
- Number of observations:
 - $4 \times 6 \times 2 = 48$ observations (collinearity equations)
- Number of unknowns:
 - $4 \times 6 + 3 \times 4 = 36$ unknowns
- Redundancy:
 - 12
- Assumptions:
 - IOPs are assumed to be known and errorless.
 - Ground coordinates of the control points are errorless.

Structure of the Design Matrix (BA)

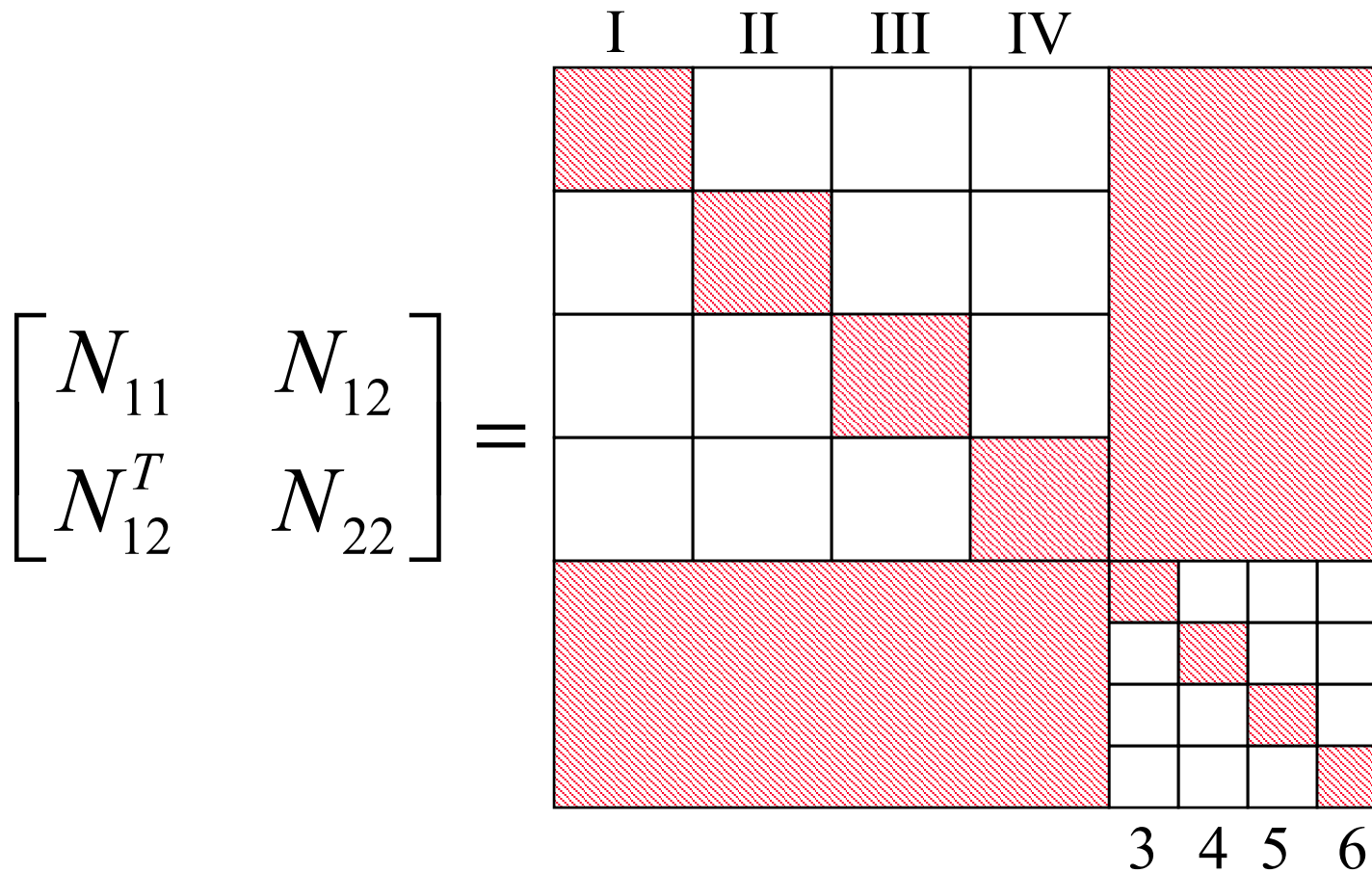


- $Y = a(X) + e \quad e \sim (0, \sigma^2 P^{-1})$
- Using approximate values for the unknown parameters (X^0) and partial derivatives, the above equations can be linearized leading to the following equations:
- $y_{48 \times 1} = A_{48 \times 36} x_{36 \times 1} + e_{48 \times 1} \quad e \sim (0, \sigma^2 P^{-1})$

Structure of the Design Matrix



Structure of the Normal Matrix



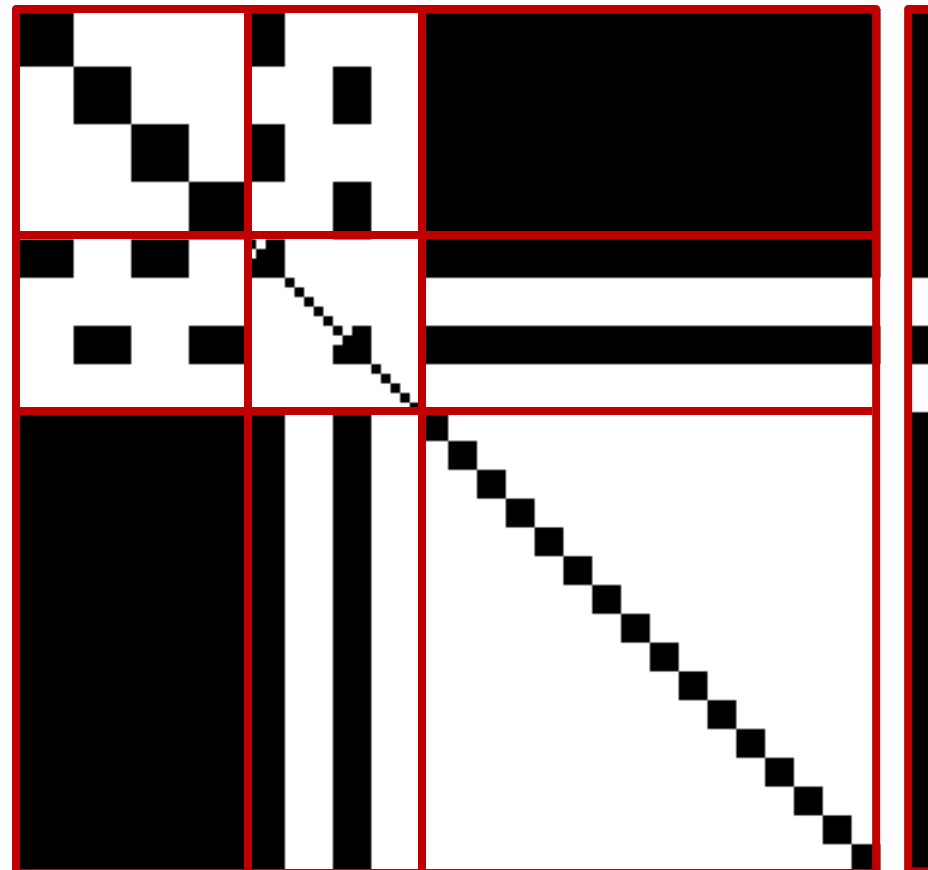
Sample Data



- 2 cameras.
- 4 images.
- 16 points.

- All the points appear in all the images
- Two images were captured by each camera

Structure of the Normal Matrix: Example



N

C



Observation Equations

$$y_{n \times 1} = A_{n \times m} x_{m \times 1} + e_{n \times 1} \quad e \sim (0, \sigma_o^2 P^{-1})$$

$$y_{n \times 1} = A_{1_{n \times 6 m_1}} x_{1_{6 m_1 \times 1}} + A_{2_{n \times 3 m_2}} x_{2_{3 m_2 \times 1}} + e_{n \times 1}$$

$$y_{n \times 1} = \begin{bmatrix} A_{1_{n \times 6 m_1}} & A_{2_{n \times 3 m_2}} \end{bmatrix} \begin{bmatrix} x_{1_{6 m_1 \times 1}} \\ x_{2_{3 m_2 \times 1}} \end{bmatrix} + e_{n \times 1}$$

- **n** \equiv Number of observations (image coordinate measurements)
- **m** \equiv Number of unknowns:
 - **m₁** \equiv Number of images $\Rightarrow 6 m_1$ (EOPs of the images)
 - **m₂** \equiv Number of tie points $\Rightarrow 3 m_2$ (ground coordinates of tie points)
 - **m** = $6 m_1 + 3 m_2$



Normal Equation Matrix

$$N_{(6m_1+3m_2) \times (6m_1+3m_2)} = \begin{bmatrix} A_1^T \\ A_2^T \end{bmatrix} P \begin{bmatrix} A_1 & A_2 \end{bmatrix}$$

$$N = \begin{bmatrix} N_{11}_{6m_1 \times 6m_1} & N_{12}_{6m_1 \times 3m_2} \\ N_{12}^T_{3m_2 \times 6m_1} & N_{22}_{3m_2 \times 3m_2} \end{bmatrix}$$

$$C_{(6m_1+3m_2) \times 1} = \begin{bmatrix} A_1^T \\ A_2^T \end{bmatrix} P y = \begin{bmatrix} A_1^T P y \\ A_2^T P y \end{bmatrix} = \begin{bmatrix} C_1_{6m_1 \times 1} \\ C_2_{3m_2 \times 1} \end{bmatrix}$$

Normal Equation Matrix

- N_{11} is a block diagonal matrix with 6x6 sub-blocks along the diagonal.
- N_{22} is a block diagonal matrix with 3x3 sub-blocks along the diagonal.

$$\begin{bmatrix} N_{11_{6m_1 \times 6m_1}} & N_{12_{6m_1 \times 3m_2}} \\ N_{12^T_{3m_2 \times 6m_1}} & N_{22_{3m_2 \times 3m_2}} \end{bmatrix} \begin{bmatrix} \hat{x}_1_{6m_1 \times 1} \\ \hat{x}_2_{3m_2 \times 1} \end{bmatrix} = \begin{bmatrix} C_1_{6m_1 \times 1} \\ C_2_{3m_2 \times 1} \end{bmatrix}$$

- Question: Under which circumstances will we deviate from this structure?

Reduction of the Normal Equation Matrix



$$N_{11_{6m_1 \times 6m_1}} \hat{x}_{1_{6m_1 \times 1}} + N_{12_{6m_1 \times 3m_2}} \hat{x}_{2_{3m_2 \times 1}} = C_{1_{6m_1 \times 1}}$$

$$N_{12_{3m_2 \times 6m_1}}^T \hat{x}_{1_{6m_1 \times 1}} + N_{22_{3m_2 \times 3m_2}} \hat{x}_{2_{3m_2 \times 1}} = C_{2_{3m_2 \times 1}}$$

- Solving for x_2 first:

$$N_{12_{3m_2 \times 6m_1}}^T \left(N_{11_{6m_1 \times 6m_1}}^{-1} C_{1_{6m_1 \times 1}} - N_{11_{6m_1 \times 6m_1}}^{-1} N_{12_{6m_1 \times 3m_2}} \hat{x}_{2_{3m_2 \times 1}} \right) + N_{22_{3m_2 \times 3m_2}} \hat{x}_{2_{3m_2 \times 1}} = C_{2_{3m_2 \times 1}}$$

$$\hat{x}_{2_{3m_2 \times 1}} = \left(N_{22_{3m_2 \times 3m_2}} - N_{12_{3m_2 \times 6m_1}}^T N_{11_{6m_1 \times 6m_1}}^{-1} N_{12_{6m_1 \times 3m_2}} \right)^{-1} \left(C_{2_{3m_2 \times 1}} - N_{12_{3m_2 \times 6m_1}}^T N_{11_{6m_1 \times 6m_1}}^{-1} C_{1_{6m_1 \times 1}} \right)$$

$$\hat{x}_{1_{6m_1 \times 1}} = \left(N_{11_{6m_1 \times 6m_1}}^{-1} C_{1_{6m_1 \times 1}} - N_{11_{6m_1 \times 6m_1}}^{-1} N_{12_{6m_1 \times 3m_2}} \hat{x}_{2_{3m_2 \times 1}} \right)$$

- $3m_2 < 6m_1$
- Remember: N_{11} is a block diagonal matrix with 6x6 sub-blocks along the diagonal.

Reduction of the Normal Equation Matrix



$$N_{11_{6m_1 \times 6m_1}} \hat{x}_{1_{6m_1 \times 1}} + N_{12_{6m_1 \times 3m_2}} \hat{x}_{2_{3m_2 \times 1}} = C_{1_{6m_1 \times 1}}$$

$$N_{12_{3m_2 \times 6m_1}}^T \hat{x}_{1_{6m_1 \times 1}} + N_{22_{3m_2 \times 3m_2}} \hat{x}_{2_{3m_2 \times 1}} = C_{2_{3m_2 \times 1}}$$

- Solving for x_1 first:

$$N_{11_{6m_1 \times 6m_1}} \hat{x}_{1_{6m_1 \times 1}} + N_{12_{3m_2 \times 3m_2}} \left(N_{22_{3m_2 \times 3m_2}}^{-1} C_{2_{3m_2 \times 1}} - N_{22_{3m_2 \times 3m_2}}^{-1} N_{12_{3m_2 \times 6m_1}}^T \hat{x}_{1_{6m_1 \times 1}} \right) = C_{1_{6m_1 \times 1}}$$

$$\hat{x}_{1_{6m_1 \times 1}} = \left(N_{11_{6m_1 \times 6m_1}} - N_{12_{6m_1 \times 3m_2}} N_{22_{3m_2 \times 3m_2}}^{-1} N_{12_{3m_2 \times 6m_1}}^T \right)^{-1} \left(C_{1_{6m_1 \times 1}} - N_{12_{6m_1 \times 3m_2}} N_{22_{3m_2 \times 3m_2}}^{-1} C_{2_{3m_2 \times 1}} \right)$$

$$\hat{x}_{2_{3m_2 \times 1}} = \left(N_{22_{3m_2 \times 3m_2}}^{-1} C_{2_{3m_2 \times 1}} - N_{22_{3m_2 \times 3m_2}}^{-1} N_{12_{3m_2 \times 6m_1}}^T \hat{x}_{1_{6m_1 \times 1}} \right)$$

- $6m_1 < 3m_2$
- Remember: N_{22} is a block diagonal matrix with 3x3 sub-blocks along the diagonal.

Reduction of the Normal Equation Matrix



- Variance covariance matrix of the estimated parameters:

$$D\{\hat{x}_{1_{6m_1 \times 1}}\} = \sigma_o^2 \left(N_{11_{6m_1 \times 6m_1}} - N_{12_{6m_1 \times 3m_2}} N_{22_{3m_2 \times 3m_2}}^{-1} N_{12_{3m_2 \times 6m_1}}^T \right)^{-1}$$

$$D\{\hat{x}_{2_{3m_2 \times 1}}\} = \sigma_o^2 \left(N_{22_{3m_2 \times 3m_2}} - N_{12_{3m_2 \times 6m_1}}^T N_{11_{6m_1 \times 6m_1}}^{-1} N_{12_{6m_1 \times 3m_2}} \right)^{-1}$$



Building the Normal Equation Matrix

- We would like to investigate the possibility of sequentially building up the normal equation matrix without fully building the design matrix.
- (x_{ij}, y_{ij}) image coordinates of the i^{th} point in the j^{th} image

$$y_{2 \times 1_{ij}} = A_{1_{2 \times 6_{ij}}} x_{1_{6 \times 1_j}} + A_{2_{2 \times 3_{ij}}} x_{2_{3 \times 1_i}} + e_{2 \times 1_{ij}}$$

$$y_{2 \times 1_{ij}} = \begin{bmatrix} A_{1_{2 \times 6_{ij}}} & A_{2_{2 \times 3_{ij}}} \end{bmatrix} \begin{bmatrix} x_{1_{6 \times 1_j}} \\ x_{2_{3 \times 1_i}} \end{bmatrix} + e_{2 \times 1_{ij}}$$

Normal Equation Matrix

$$y_{2 \times 1_{ij}} = \begin{bmatrix} A_{1_{2 \times 6_{ij}}} & A_{2_{2 \times 3_{ij}}} \end{bmatrix} \begin{bmatrix} x_{1_{6 \times 1_j}} \\ x_{2_{3 \times 1_i}} \end{bmatrix} + e_{2 \times 1_{ij}}$$

$$\begin{bmatrix} A_{1_{6 \times 2_{ij}}}^T \\ A_{2_{3 \times 2_{ij}}}^T \end{bmatrix} P_{ij} \begin{bmatrix} A_{1_{2 \times 6_{ij}}} & A_{2_{2 \times 3_{ij}}} \end{bmatrix} \begin{bmatrix} x_{1_{6 \times 1_j}} \\ x_{2_{3 \times 1_i}} \end{bmatrix} = \begin{bmatrix} A_{1_{6 \times 2_{ij}}}^T \\ A_{2_{3 \times 2_{ij}}}^T \end{bmatrix} P_{ij} y_{2 \times 1_{ij}}$$

$$\begin{bmatrix} A_{1_{6 \times 2_{ij}}}^T P_{ij} A_{1_{2 \times 6_{ij}}} & A_{1_{6 \times 2_{ij}}}^T P_{ij} A_{2_{2 \times 3_{ij}}} \\ A_{2_{3 \times 2_{ij}}}^T P_{ij} A_{1_{2 \times 6_{ij}}} & A_{2_{3 \times 2_{ij}}}^T P_{ij} A_{2_{2 \times 3_{ij}}} \end{bmatrix} \begin{bmatrix} x_{1_{6 \times 1_j}} \\ x_{2_{3 \times 1_i}} \end{bmatrix} = \begin{bmatrix} A_{1_{6 \times 2_{ij}}}^T P_{ij} y_{2 \times 1_{ij}} \\ A_{2_{3 \times 2_{ij}}}^T P_{ij} y_{2 \times 1_{ij}} \end{bmatrix}$$

Normal Equation Matrix

$$\begin{bmatrix} A_{1\ 6 \times 2\ ij}^T & P_{ij} & A_{1\ 2 \times 6\ ij} \\ A_{2\ 3 \times 2\ ij}^T & P_{ij} & A_{2\ 2 \times 3\ ij} \end{bmatrix} \begin{bmatrix} x_{1\ 6 \times 1\ j} \\ x_{2\ 3 \times 1\ i} \end{bmatrix} = \begin{bmatrix} A_{1\ 6 \times 2\ ij}^T & P_{ij} & y_{2 \times 1\ ij} \\ A_{2\ 3 \times 2\ ij}^T & P_{ij} & y_{2 \times 1\ ij} \end{bmatrix}$$

$$\begin{bmatrix} N_{11\ ij} & N_{12\ ij} \\ N_{12\ ij}^T & N_{22\ ij} \end{bmatrix}_{9 \times 9} \begin{bmatrix} x_{1\ 6 \times 1\ j} \\ x_{2\ 3 \times 1\ i} \end{bmatrix}_{9 \times 1} = \begin{bmatrix} C_{1\ ij} \\ C_{2\ ij} \end{bmatrix}_{9 \times 1}$$

- Note: We cannot solve this matrix for the:
 - The Exterior Orientation Parameters of the j^{th} image, and
 - The ground coordinates of the i^{th} point.



Normal Equation Matrix

$$\begin{bmatrix} N_{11}_{6m_1 \times 6m_1} & N_{12}_{6m_1 \times 3m_2} \\ N_{12}^T_{3m_2 \times 6m_1} & N_{22}_{3m_2 \times 3m_2} \end{bmatrix} \begin{bmatrix} \hat{x}_1_{6m_1 \times 1} \\ \hat{x}_2_{3m_2 \times 1} \end{bmatrix} = \begin{bmatrix} C_1_{6m_1 \times 1} \\ C_2_{3m_2 \times 1} \end{bmatrix}$$

- Question: How can we sequentially build the above matrices?
- Assumption: All the points are common to all the images.

N_{11} - Matrix

$$N_{11(6m_1 \times 6m_1)} = \begin{bmatrix} \sum_{i=1}^{m_2} N_{11_{i1}} & 0 & 0 & \dots & \dots & 0 \\ 0 & \sum_{i=1}^{m_2} N_{11_{i2}} & 0 & \dots & \dots & 0 \\ 0 & 0 & \sum_{i=1}^{m_2} N_{11_{i3}} & \dots & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & \dots & \sum_{i=1}^{m_2} N_{11_{im_1}} \end{bmatrix}$$

- If all the points are not common to all the images:
 - The summation should be carried over all the points that appear in the image under consideration.

N_{22} - Matrix

$$N_{22(3m_2 \times 3m_2)} = \begin{bmatrix} \sum_{j=1}^{m_1} N_{22_{1j}} & 0 & 0 & \dots & \dots & 0 \\ 0 & \sum_{j=1}^{m_1} N_{22_{2j}} & 0 & \dots & \dots & 0 \\ 0 & 0 & \sum_{j=1}^{m_1} N_{22_{3j}} & \dots & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & \dots & \sum_{j=1}^{m_1} N_{22_{m_2j}} \end{bmatrix}$$

- If all the points are not common to all the images:
 - The summation should be carried over all the images within which the point under consideration appears.

N_{12} - Matrix

$$N_{12(6m_1 \times 3m_2)} = \begin{bmatrix} N_{12_{11}} & N_{12_{21}} & N_{12_{31}} & \cdots & \cdots & N_{12_{m_2 1}} \\ N_{12_{12}} & N_{12_{22}} & N_{12_{32}} & \cdots & \cdots & N_{12_{m_2 2}} \\ N_{12_{13}} & N_{12_{23}} & N_{12_{33}} & \cdots & \cdots & N_{12_{m_2 3}} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ N_{12_{1m_1}} & N_{12_{2m_1}} & N_{12_{3m_1}} & \cdots & \cdots & N_{12_{m_2 m_1}} \end{bmatrix}$$

- If point “i” does not appear in image “j”:
 - $(N_{12})_{ij} = 0$

C - Matrix

$$C_{1_{6m_1 \times 1}} = \begin{bmatrix} \sum_{i=1}^{m_2} C_{1_{i1}} \\ \sum_{i=1}^{m_2} C_{1_{i2}} \\ \sum_{i=1}^{m_2} C_{1_{i3}} \\ \vdots \\ \sum_{i=1}^{m_2} C_{1_{im_1}} \end{bmatrix}$$

$$C_{2_{3m_2 \times 1}} = \begin{bmatrix} \sum_{j=1}^{m_1} C_{2_{1j}} \\ \sum_{j=1}^{m_1} C_{2_{2j}} \\ \sum_{j=1}^{m_1} C_{2_{3j}} \\ \vdots \\ \sum_{j=1}^{m_1} C_{2_{m_2j}} \end{bmatrix}$$

- Once again, we assumed that all the points are common to all the images.

Precision of Bundle Block Adjustment



- The precision of the estimated EOPs as well as the ground coordinates of tie points can be obtained by the product of:
 - The estimated variance component, and
 - The inverse of the normal equation matrix (cofactor matrix).
- The precision depends on the following factors:
 - Geometric configuration of the image block
 - Base-Height ratio
 - Image scale
 - Image coordinate measurement precision

Precision of Bundle Block Adjustment



- Precision of a single model: If we have
 - Bundle block adjustment with additional parameters that compensate for various distortions
 - Regular blocks with 60% overlap and 20% side lap
 - Signalized targets

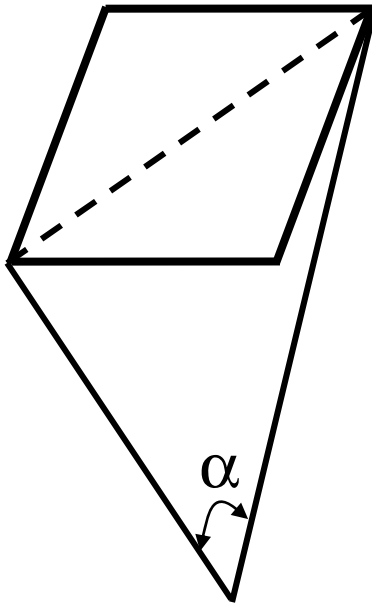
$$\sigma_{XY} = \pm 3\mu m$$

$\sigma_Z = \pm 0.003\%$ of the camera principal distance (NA and WA cameras)

$\sigma_Z = \pm 0.004\%$ of the camera principal distance (SWA cameras)

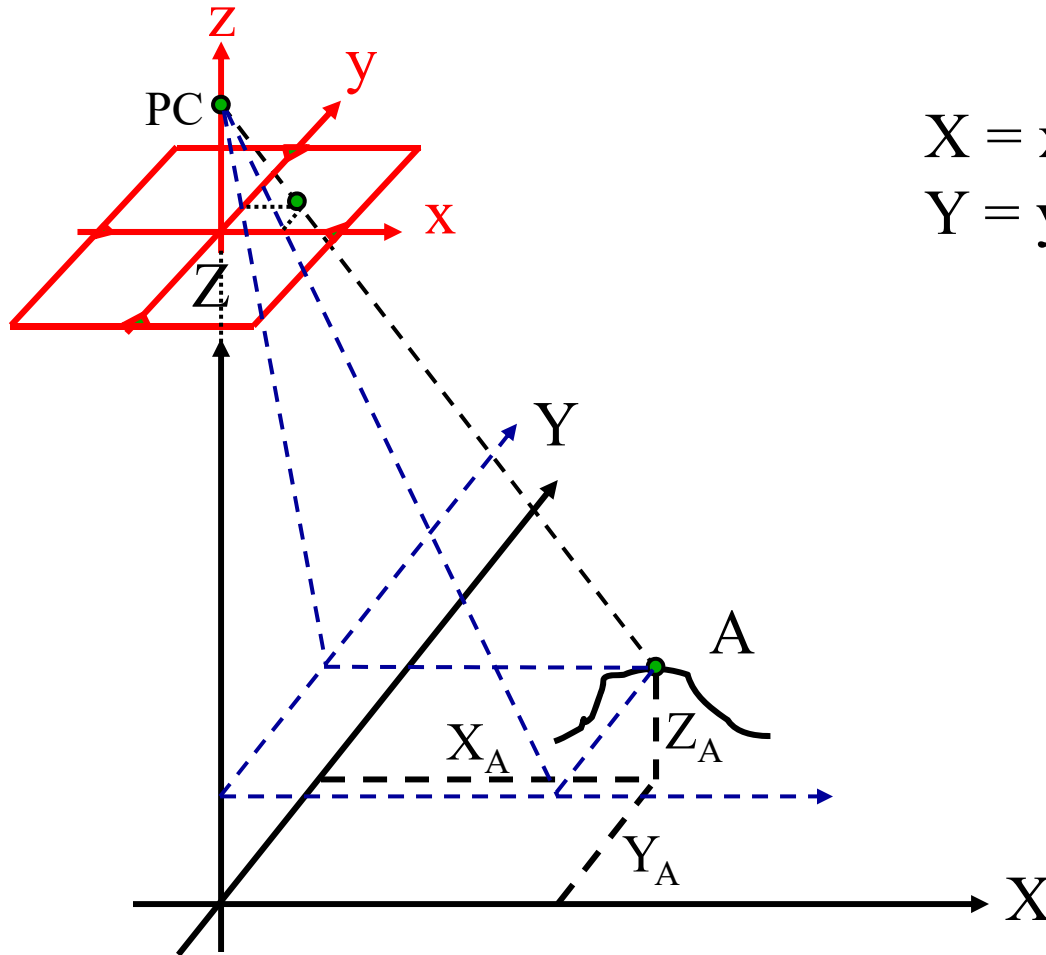
These precision values are given in the image space

Camera Classification



- $\alpha < 75^\circ$ Normal angle camera (NA)
- $100^\circ > \alpha > 75^\circ$ Wide angle camera (WA)
- $\alpha > 100^\circ$ Super wide angle camera (SWA)

Precision of Bundle Block Adjustment



$$X = x * Z / c$$

$$Y = y * Z / c$$

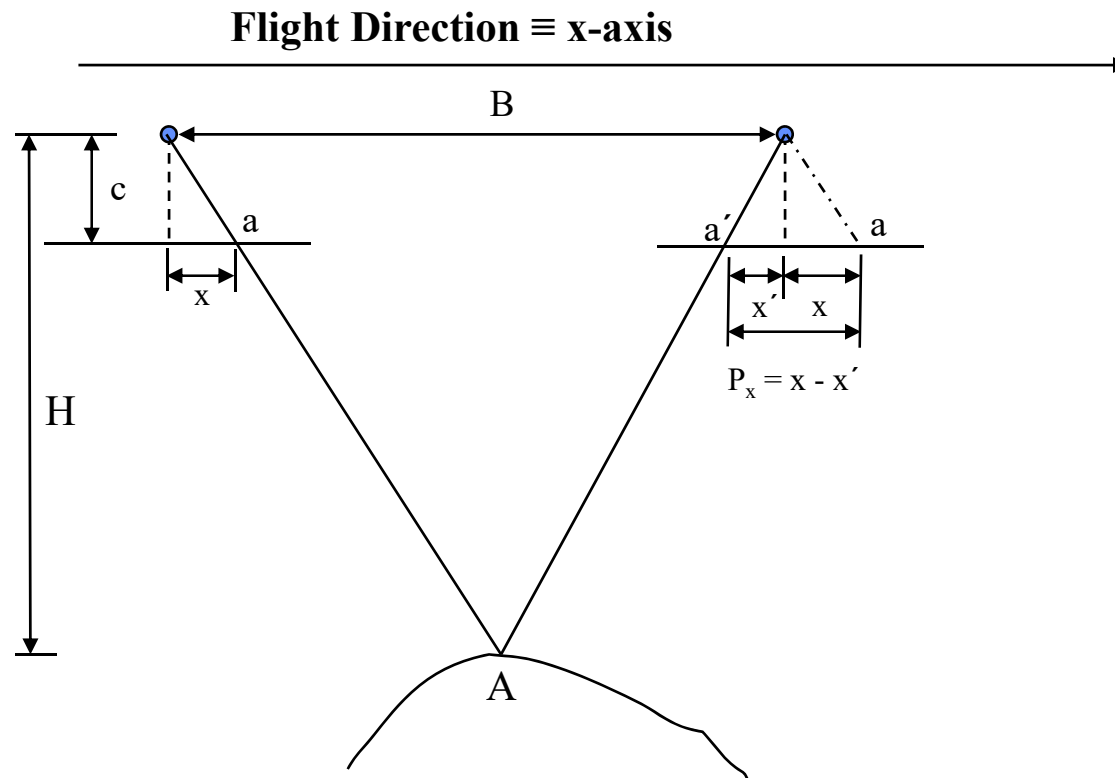
$$\sigma_X = \frac{Z}{c} \sigma_x$$

$$\sigma_Y = \frac{Z}{c} \sigma_y$$

Precision of Bundle Block Adjustment



Vertical Precision



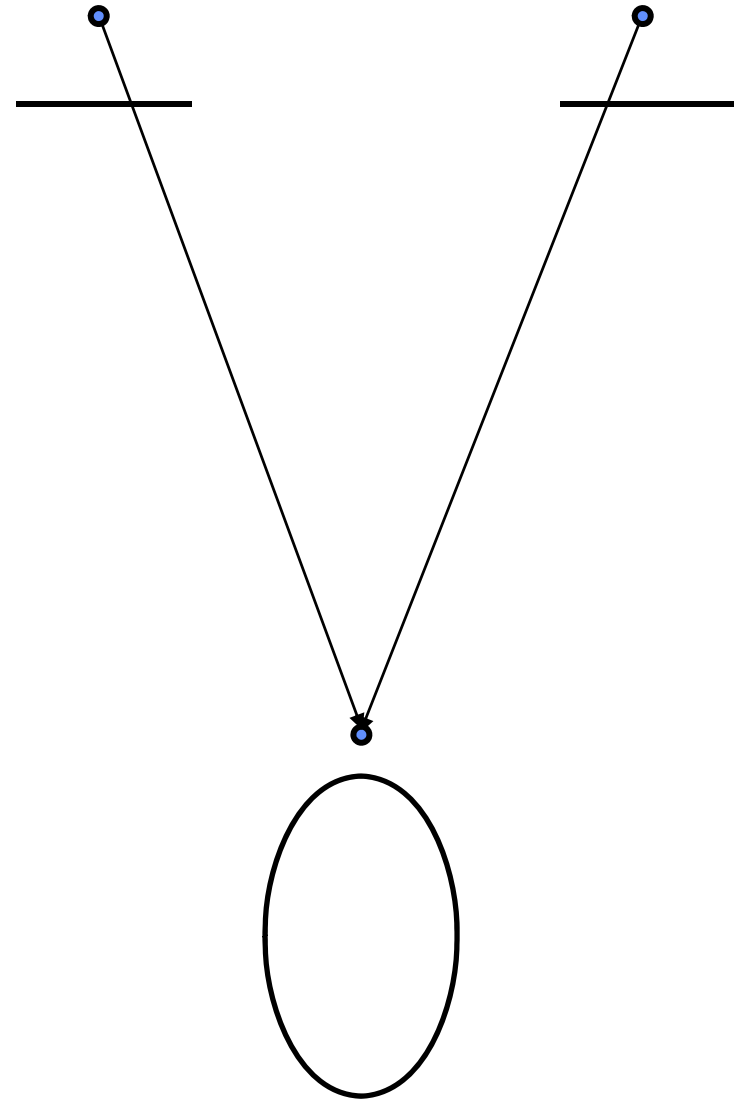
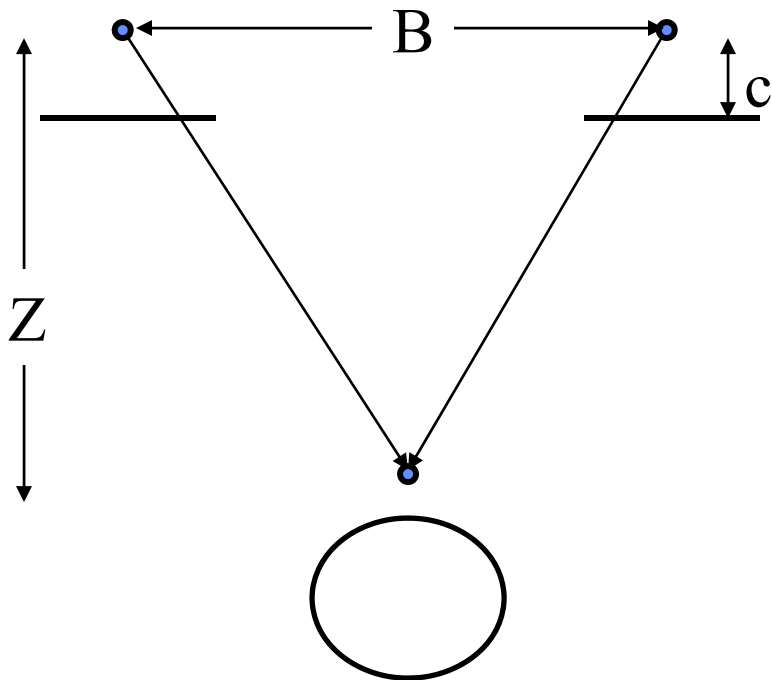
$$P_x / B = c / H$$
$$H = B c / P_x$$

Precision of Bundle Block Adjustment



Vertical Precision

$$\sigma_Z = \frac{Z}{c} \frac{Z}{B} \sigma_{p_x}$$

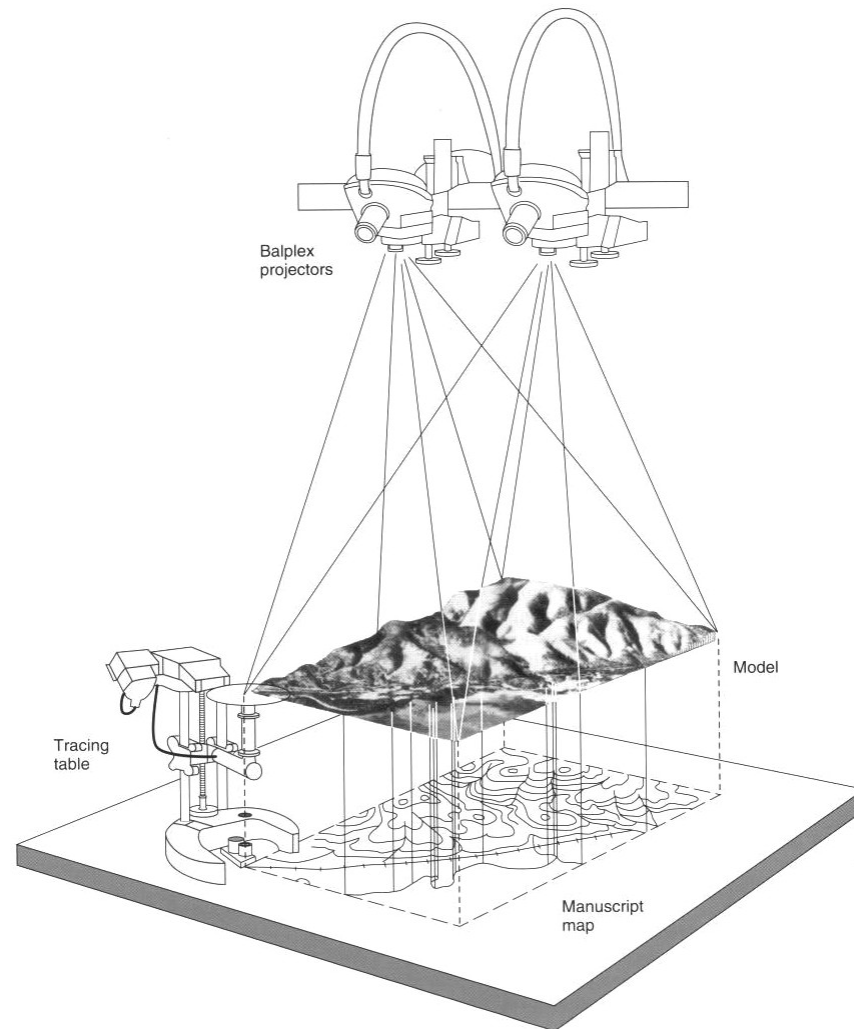


Advantages of Bundle Block Adjustment



- Most accurate triangulation technique since we have direct transformation between image and ground coordinates.
- Straight forward to include parameters that compensate for various deviations from the collinearity model.
- Straight forward to include additional observations:
 - GNSS/INS observations at the exposure stations
 - Object space distances
- Can be used for normal, convergent, aerial, and close range imagery
- After the adjustment, the EOPs can be set on analogue and analytical plotters for compilation purposes.

Photogrammetric Compilation





Disadvantages of Bundle Block Adjustment

- Model is non linear: approximations as well as partial derivatives are needed.
- Requires computer intensive computations.
- Analogue instruments cannot be used (they cannot measure image coordinate measurements).
- The adjustment cannot be separated into planimetric and vertical adjustment.

Bundle Adjustment: Final Remarks



- Elementary Unit: Images
- Measurements: Image coordinates
- Mathematical model: Collinearity equations
- Instruments: Comparators, analytical plotters, and Digital Photogrammetric Workstations (DPW)
- Required computer power: Very large
- Expected accuracy: High



Special Cases

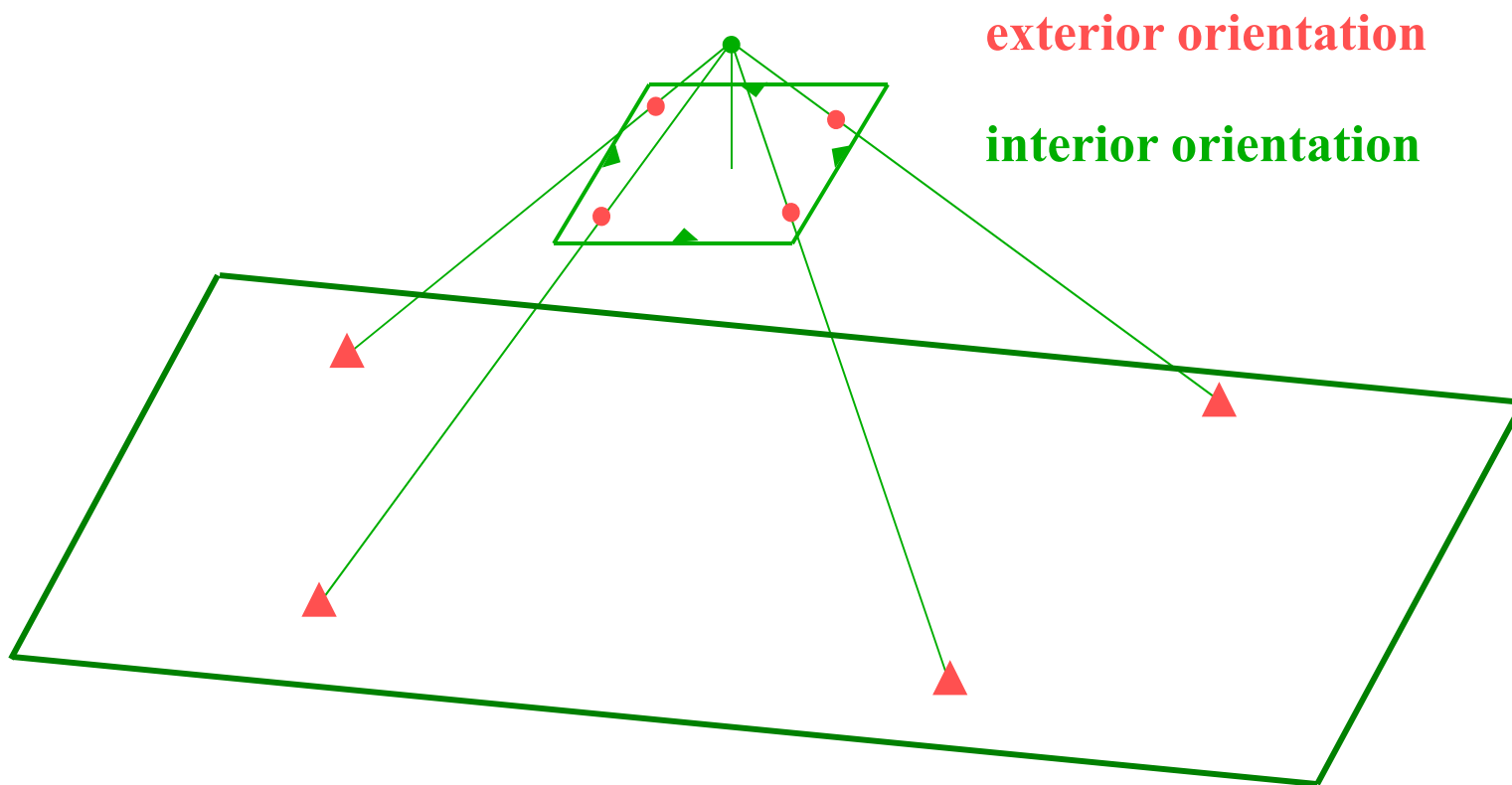
- Resection
- Intersection
- Stereo-pair orientation
- Relative orientation



Resection

- We are dealing with one image.
- We would like to determine the EOPs of this image using GCPs.
- Q: What is the minimum GCPs requirements?
 - At least 3 non-collinear GCPs are required to estimate the 6 EOPs.
 - At least 5 non-collinear (well distributed in 3-D) GCPs are required to estimate the 6 EOPs and the 3 IOPs (x_p , y_p , c).
- Critical surface:
 - The GCPs and the perspective center lie on a common cylinder.

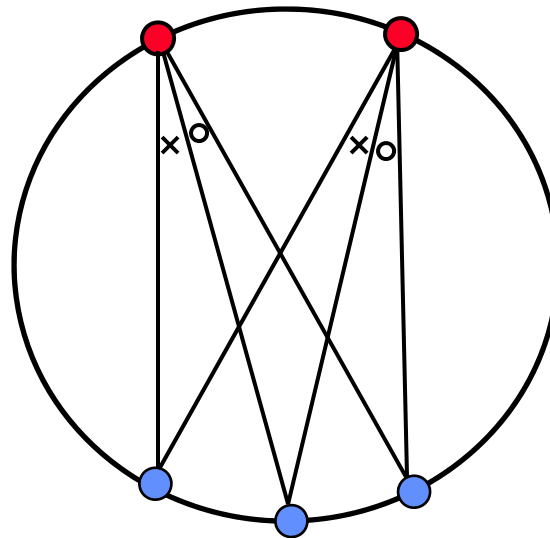
Resection



exterior orientation

interior orientation

Resection - Critical Surface



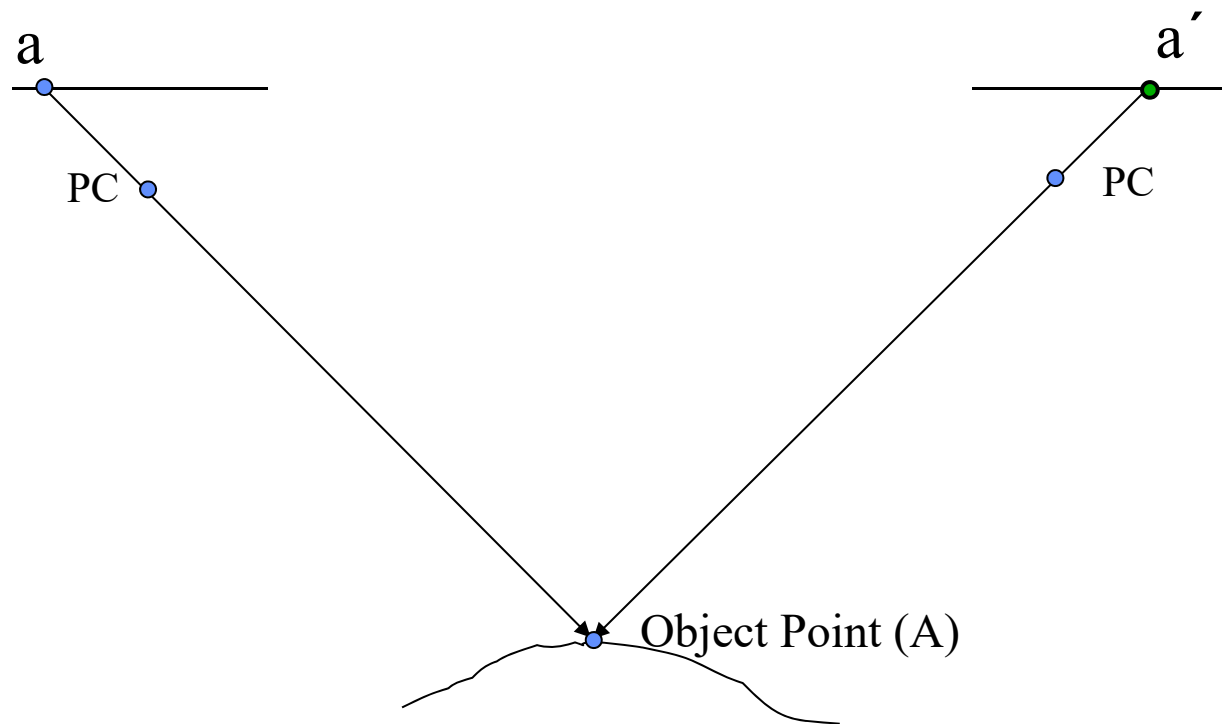
- Question: Which one of the EOPs cannot be determined?



Intersection

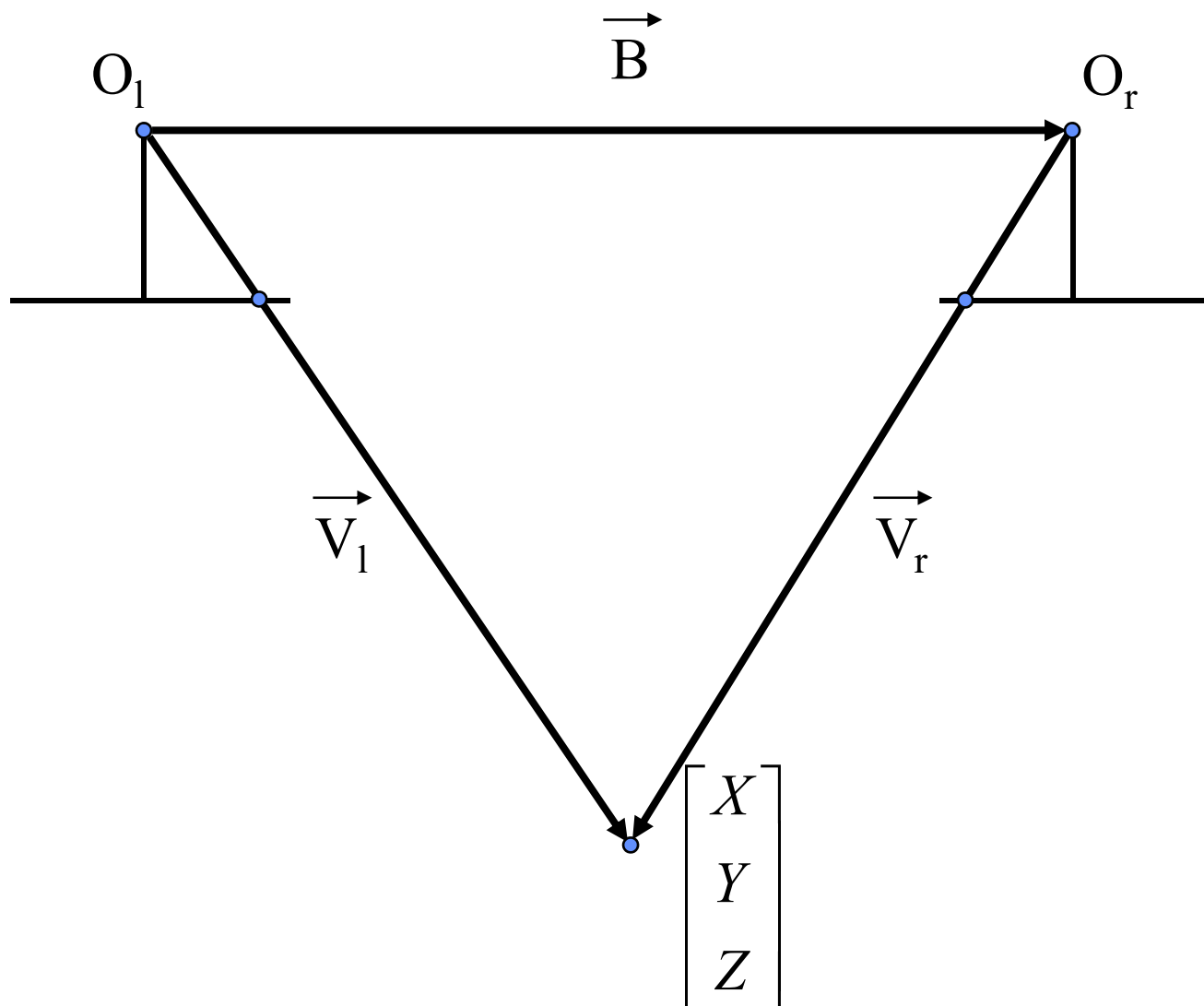
- We are dealing with two images.
- The EOPs of these images are available.
- The IOPs of the involved camera(s) are also available.
- We want to estimate the ground coordinates of points in the overlap area.
- For each tie point, we have:
 - 4 Observation equations
 - 3 Unknowns
 - Redundancy = 1
- Non-linear model: approximations are needed

Intersection





Intersection: Linear Model





Intersection: Linear Model

$$\vec{B} = \begin{bmatrix} X_{O_r} - X_{O_l} \\ Y_{O_r} - Y_{O_l} \\ Z_{O_r} - Z_{O_l} \end{bmatrix}$$

- These vectors are given w.r.t. the ground coordinate system.

$$\vec{V}_l = \lambda R_{(\omega_l, \phi_l, \kappa_l)} \begin{bmatrix} x_l - x_p \\ y_l - y_p \\ -c \end{bmatrix}$$

$$\vec{V}_r = \mu R_{(\omega_r, \phi_r, \kappa_r)} \begin{bmatrix} x_r - x_p \\ y_r - y_p \\ -c \end{bmatrix}$$



Intersection: Linear Model

$$\vec{V}_l = \vec{B} + \vec{V}_r$$

$$\begin{bmatrix} X_{o_r} - X_{o_l} \\ Y_{o_r} - Y_{o_l} \\ Z_{o_r} - Z_{o_l} \end{bmatrix} = \lambda R_{(\omega_l, \phi_l, \kappa_l)} \begin{bmatrix} x_l - x_p \\ y_l - y_p \\ -c \end{bmatrix} - \mu R_{(\omega_r, \phi_r, \kappa_r)} \begin{bmatrix} x_r - x_p \\ y_r - y_p \\ -c \end{bmatrix}$$

- Three equations in two unknowns (λ , μ).
- They are linear equations.



Intersection: Linear Model

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} X_{O_l} \\ Y_{O_l} \\ Z_{O_l} \end{bmatrix} + \lambda R_{(\omega_l, \phi_l, \kappa_l)} \begin{bmatrix} x_l - x_p \\ y_l - y_p \\ -c \end{bmatrix}$$

Or:

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} X_{O_r} \\ Y_{O_r} \\ Z_{O_r} \end{bmatrix} + \mu R_{(\omega_r, \phi_r, \kappa_r)} \begin{bmatrix} x_r - x_p \\ y_r - y_p \\ -c \end{bmatrix}$$



Stereo-pair Orientation

- Given:
 - Stereo-pair: two images with at least 50% overlap
 - Image coordinates of some tie points
 - Image and ground coordinates of control points
- Required:
 - The ground coordinates of the tie points
 - The EOPs of the involved images
- **Mini-Bundle Adjustment Procedure**



Stereo-pair Orientation

- Example:
 - Given:
 - 1 Stereo-pair
 - 20 tie points
 - No ground control points
 - Question:
 - Can we estimate the ground coordinates of the points as well as the exterior orientation parameters of that stereo-pair?
 - Answer:
 - NO



Summary

- Photogrammetry: Definition and applications
- Photogrammetric tools:
 - Rotation matrices
 - Photogrammetric orientation: interior and exterior orientation
 - Collinearity equations/conditions
- Photogrammetric bundle adjustment
 - Structure of the design and normal matrices
- Special cases:
 - Resection, intersection, and stereo-pair orientation