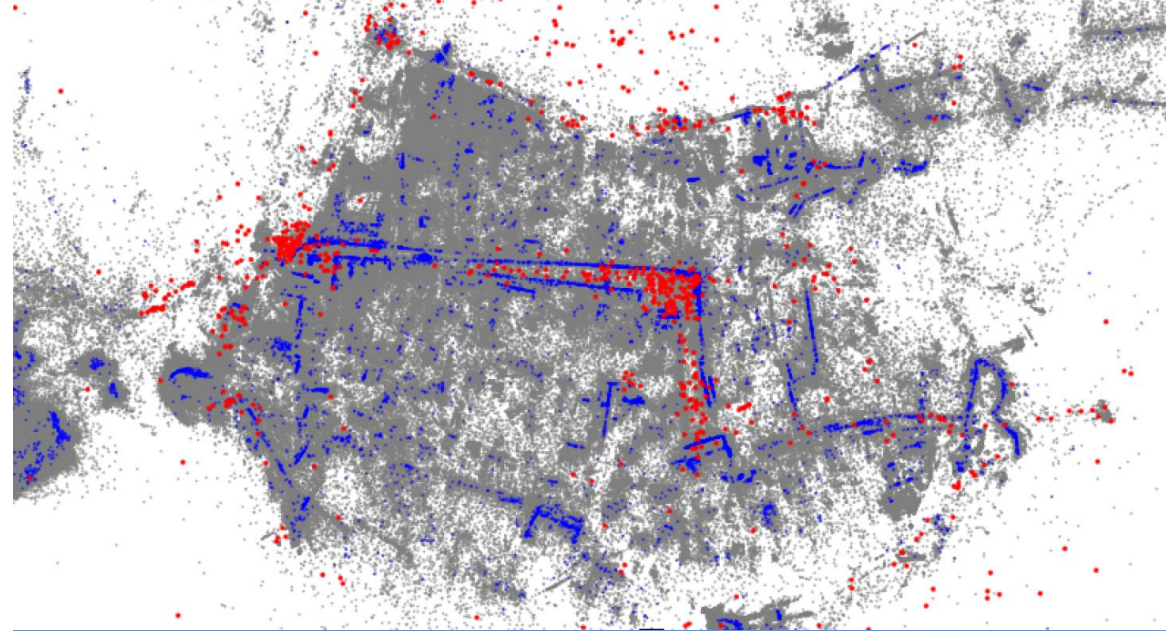


Structure from Motion: A Factorization Approach



Announcement

- HW #4 and 5 are graded.
- HW #6 is out
 - Two deadlines: May 3 (Problem 2,3,4,5) and May 10 (Problem 6 with full reconstruction)
 - 5980 students: < 5 image reconstruction
 - 8980 students: <10 image reconstruction
- Course Evaluation Today

Longuet-Higgins: Epipolar Geometry



A computer algorithm for reconstructing a scene from two projections

H. C. Longuet-Higgins

Laboratory of Experimental Psychology, University of Sussex,
Brighton BN1 9QG, UK

A simple algorithm for computing the three-dimensional structure of a scene from a correlated pair of perspective projections is described here, when the spatial relationship between the two projections is unknown. This problem is relevant not only to photographic surveying¹ but also to binocular vision², where the non-visual information available to the observer about the scene is limited by the geometry of the eyes and the brain.



Carlos Tomasi



Takeo Kanade

Shape and Motion from Image Streams under Orthography: a Factorization Method

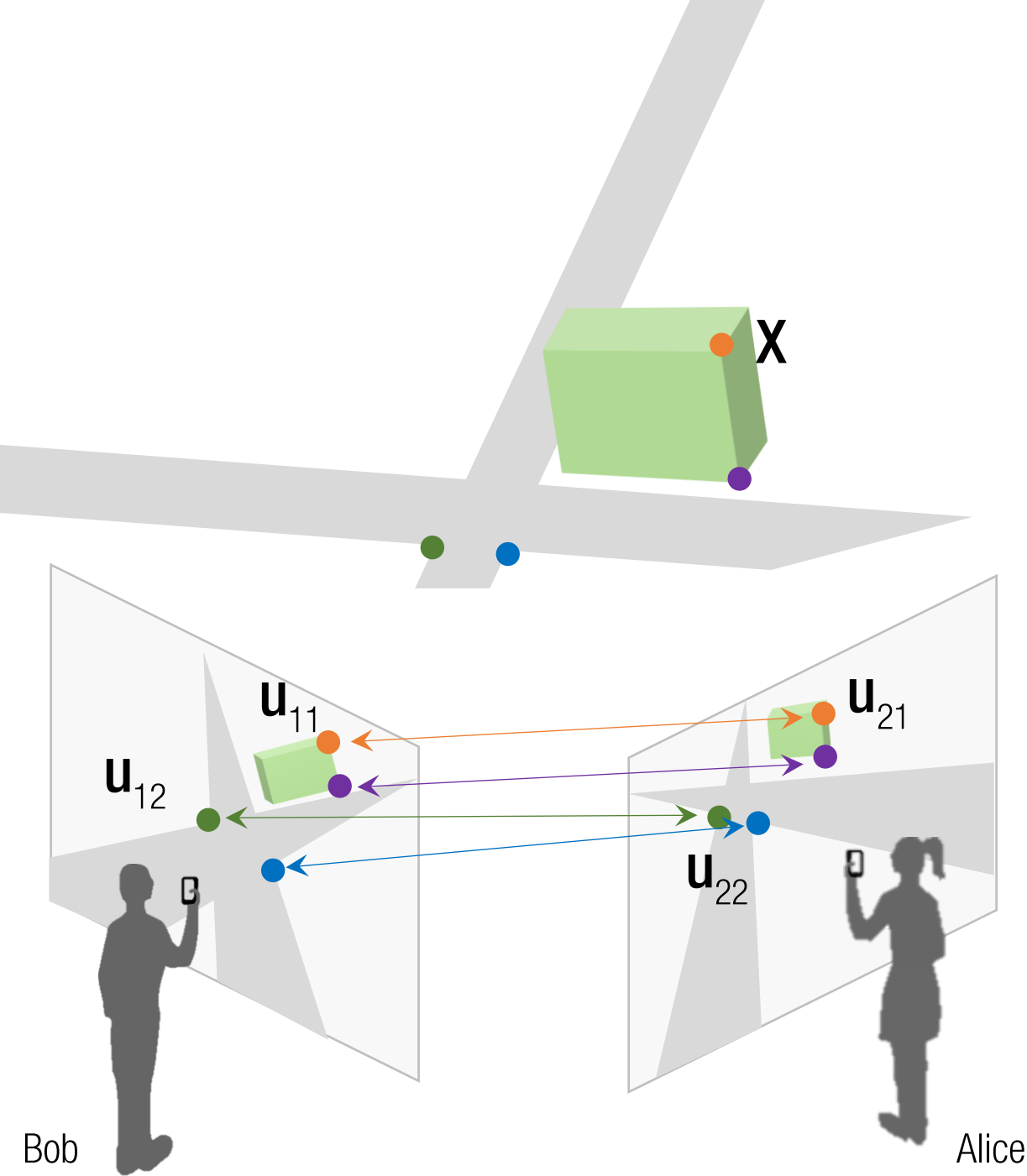
CARLO TOMASI

Department of Computer Science, Cornell University, Ithaca, NY 14850

TAKEO KANADE

School of Computer Science, Carnegie Mellon University, Pittsburgh, PA 15213

Tomasi-Kanade Factorization



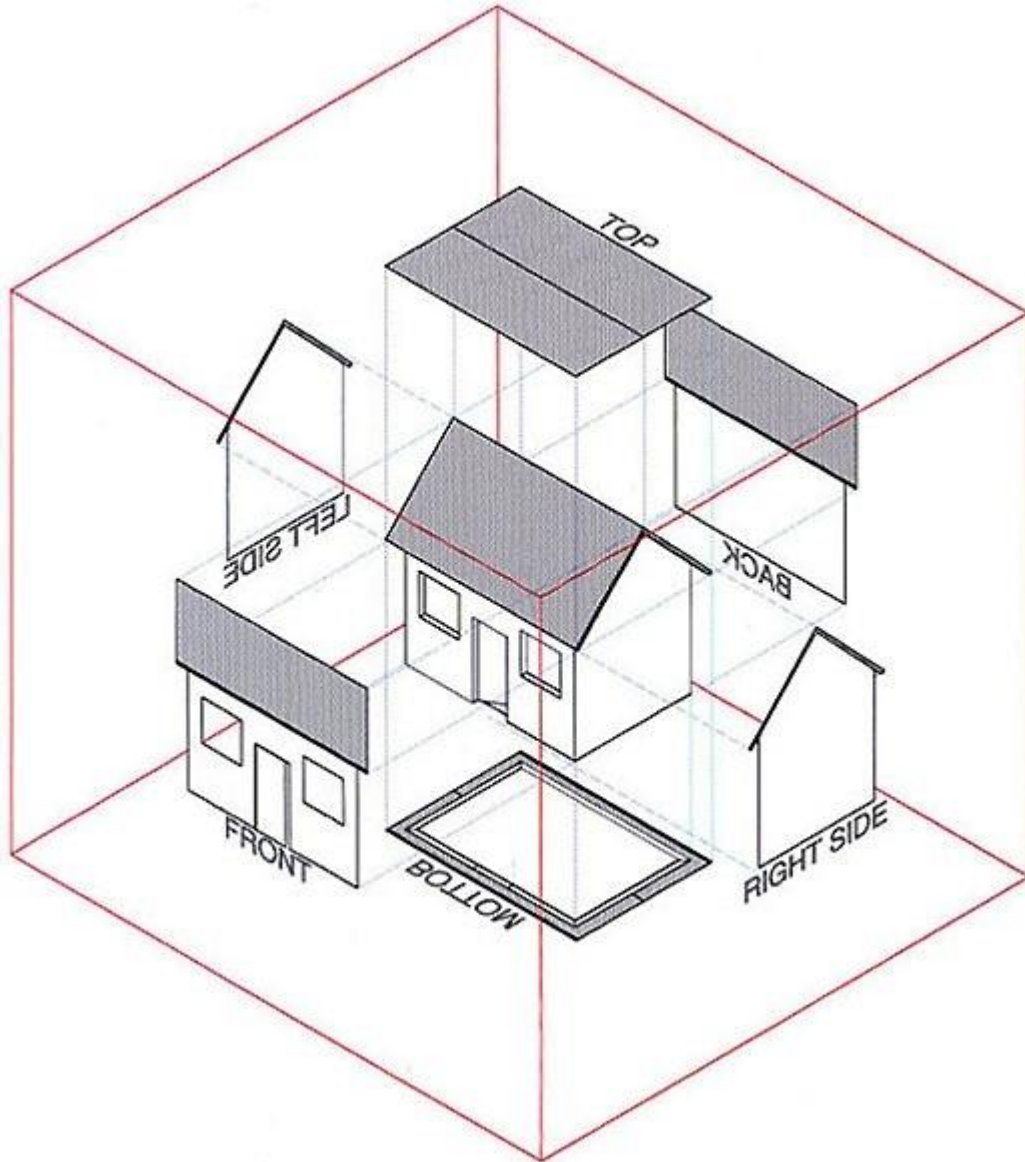
$$\lambda_{11} \mathbf{u}_{11} = \mathbf{R}_1 \mathbf{X}_1 + \mathbf{t}_1$$

$$\lambda_{12} \mathbf{u}_{12} = \mathbf{R}_1 \mathbf{X}_2 + \mathbf{t}_1$$

$$\lambda_{21} \mathbf{u}_{11} = \mathbf{R}_2 \mathbf{X}_2 + \mathbf{t}_2$$

$$\lambda_{22} \mathbf{u}_{22} = \mathbf{R}_2 \mathbf{X}_2 + \mathbf{t}_2$$

Orthographic Camera



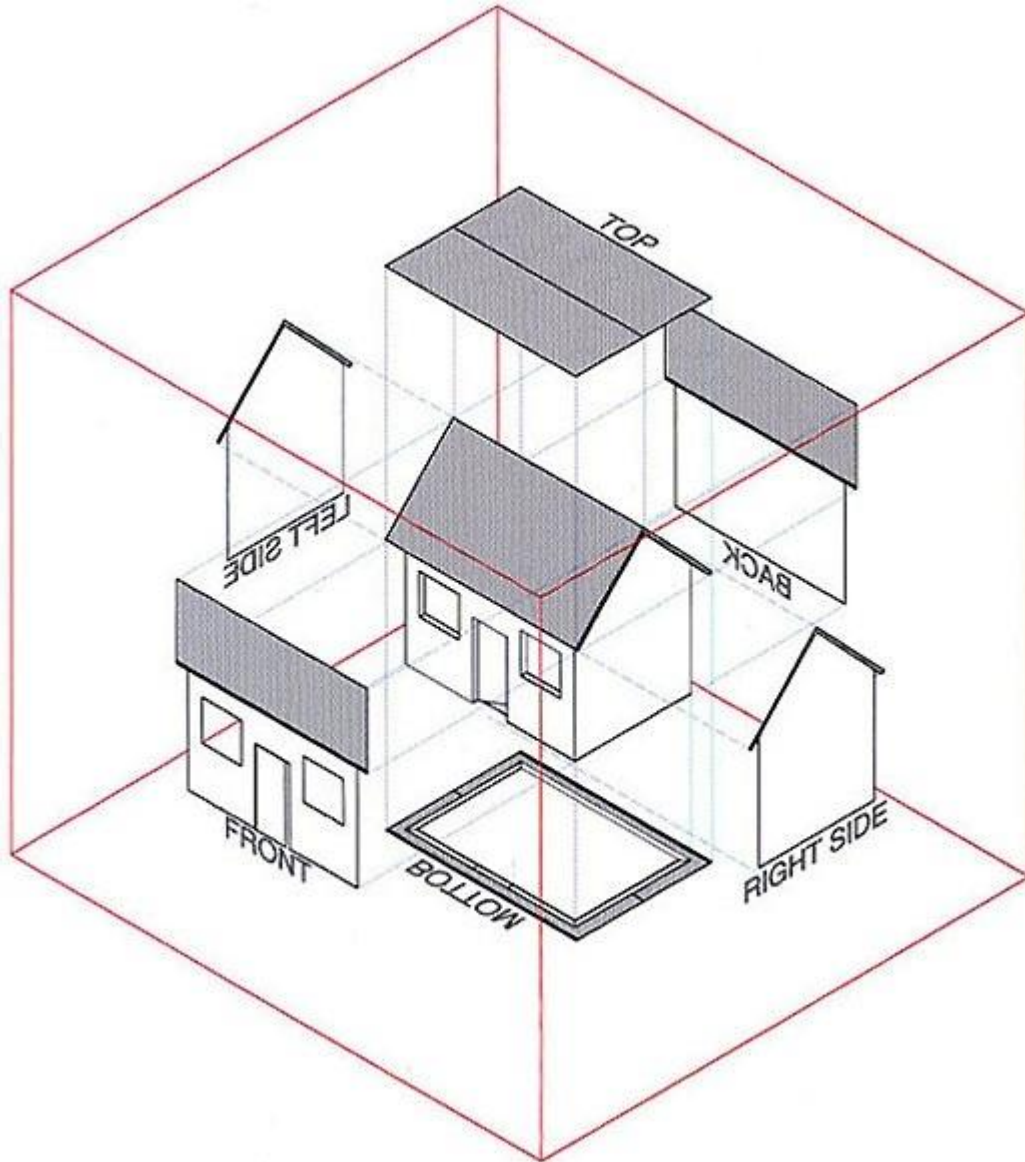
Affine camera:

$$\mathbf{P}_A = \begin{bmatrix} f & p_x \\ & f & p_y \\ & & 1 \end{bmatrix} \begin{bmatrix} r_{x1} & r_{x2} & r_{x3} & 0 \\ r_{y1} & r_{y2} & r_{y3} & 0 \\ 0 & 0 & 0 & d \end{bmatrix} = d \begin{bmatrix} f/d & p_x \\ & f/d & p_y \\ & & 1 \end{bmatrix} \begin{bmatrix} r_{x1} & r_{x2} & r_{x3} & 0 \\ r_{y1} & r_{y2} & r_{y3} & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Orthographic camera:

$$f = 1 \quad p_x = p_y = 0$$

Orthographic Camera



Affine camera:

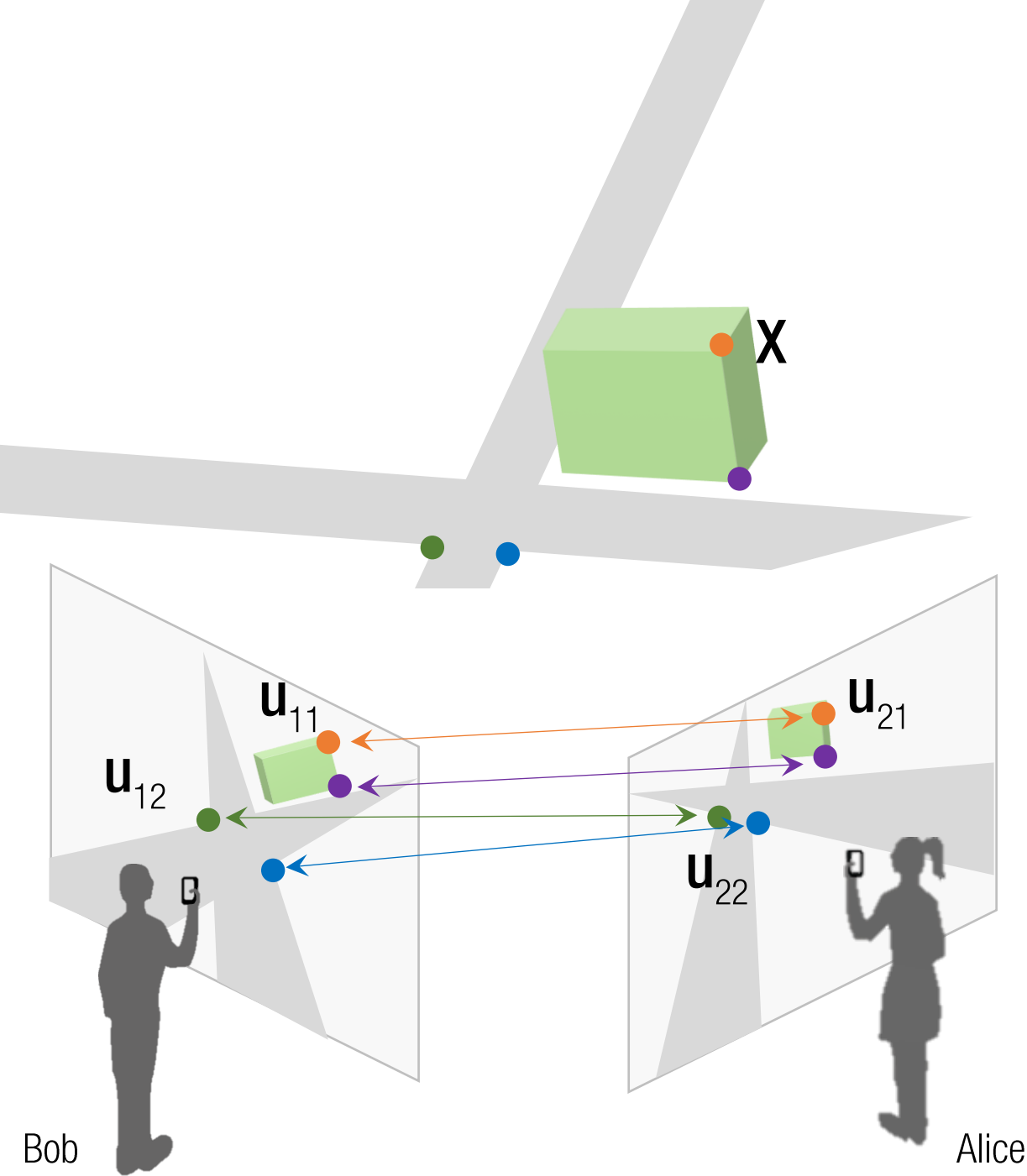
$$\mathbf{P}_A = \begin{bmatrix} f & p_x \\ & f & p_y \\ & & 1 \end{bmatrix} \begin{bmatrix} r_{x1} & r_{x2} & r_{x3} & 0 \\ r_{y1} & r_{y2} & r_{y3} & 0 \\ 0 & 0 & 0 & d \end{bmatrix} = d \begin{bmatrix} f/d & p_x \\ & f/d & p_y \\ & & 1 \end{bmatrix} \begin{bmatrix} r_{x1} & r_{x2} & r_{x3} & 0 \\ r_{y1} & r_{y2} & r_{y3} & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

Orthographic camera:

$$f = 1 \quad p_x = p_y = 0$$

$$\mathbf{P}_0 = \begin{bmatrix} r_{x1} & r_{x2} & r_{x3} & 0 \\ r_{y1} & r_{y2} & r_{y3} & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{u} = \begin{bmatrix} r_{x1} & r_{x2} & r_{x3} & t_x \\ r_{y1} & r_{y2} & r_{y3} & t_y \end{bmatrix} \begin{bmatrix} \mathbf{X} \\ 1 \end{bmatrix}$$

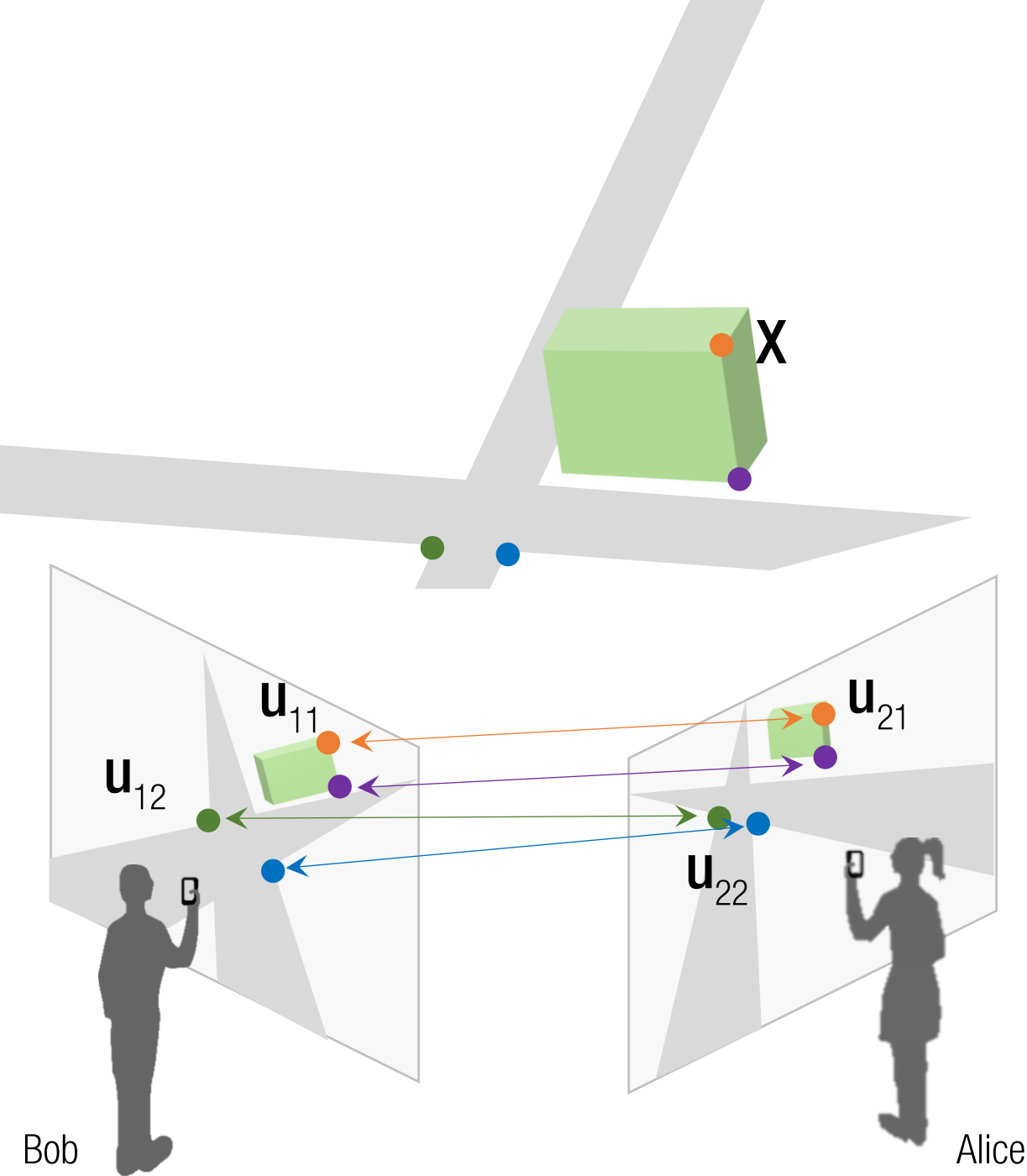


$$\lambda_{11} \mathbf{u}_{11} = \mathbf{R}_1 \mathbf{X}_1 + \mathbf{t}_1$$

$$\lambda_{12} \mathbf{u}_{12} = \mathbf{R}_1 \mathbf{X}_2 + \mathbf{t}_1$$

$$\lambda_{21} \mathbf{u}_{11} = \mathbf{R}_2 \mathbf{X}_2 + \mathbf{t}_2$$

$$\lambda_{22} \mathbf{u}_{22} = \mathbf{R}_2 \mathbf{X}_2 + \mathbf{t}_2$$



Orthographic projection:

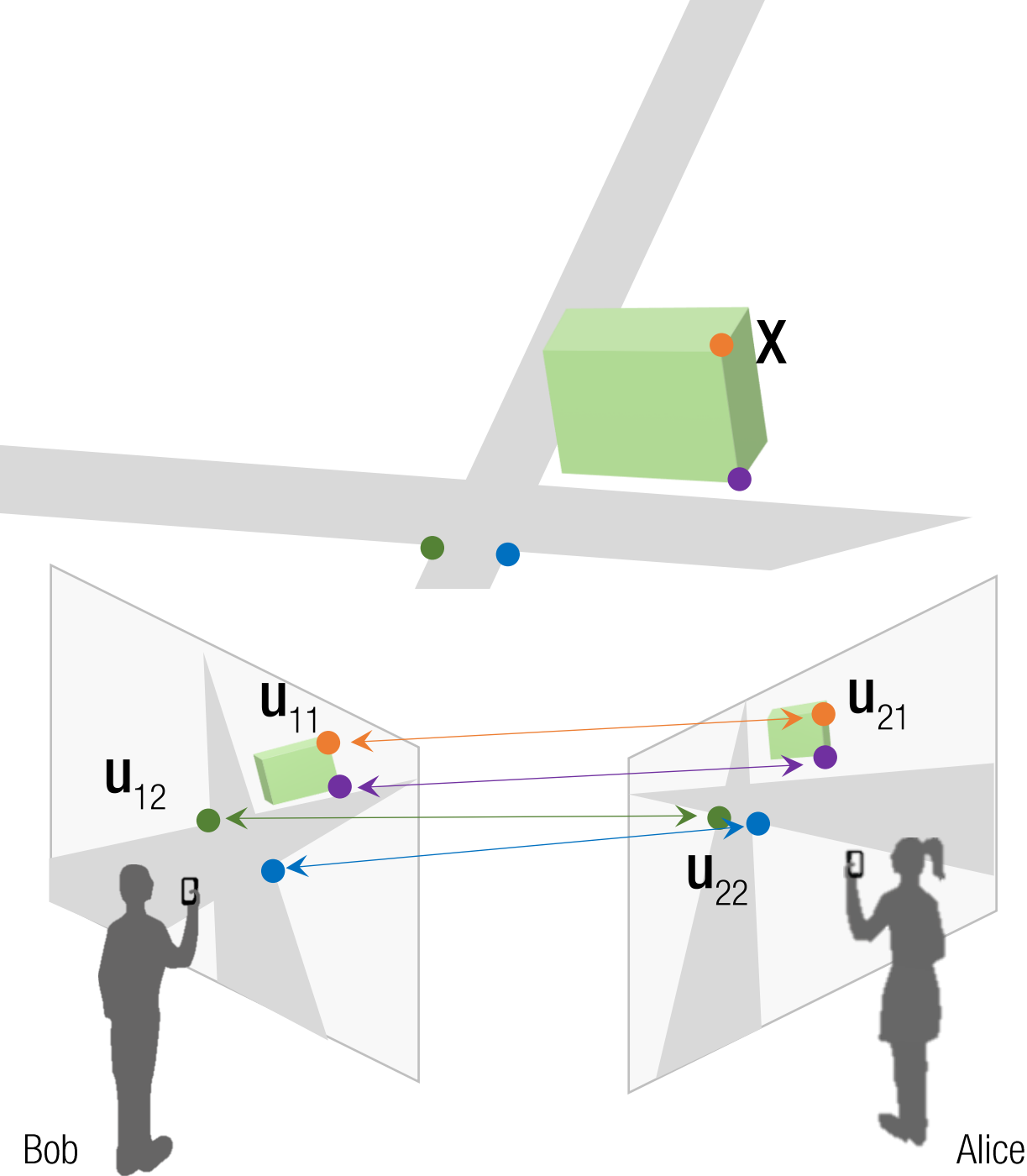
$$\mathbf{u}_{11} = \hat{\mathbf{R}}_1 \mathbf{X}_1 + \hat{\mathbf{t}}_1$$

$$\mathbf{u}_{12} = \hat{\mathbf{R}}_1 \mathbf{X}_2 + \hat{\mathbf{t}}_1$$

$$\mathbf{u}_{11} = \hat{\mathbf{R}}_2 \mathbf{X}_1 + \hat{\mathbf{t}}_2$$

$$\mathbf{u}_{22} = \hat{\mathbf{R}}_2 \mathbf{X}_2 + \hat{\mathbf{t}}_2$$

where $\hat{\mathbf{R}} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \end{bmatrix}$ $\hat{\mathbf{t}} = \begin{bmatrix} t_x \\ t_y \end{bmatrix}$



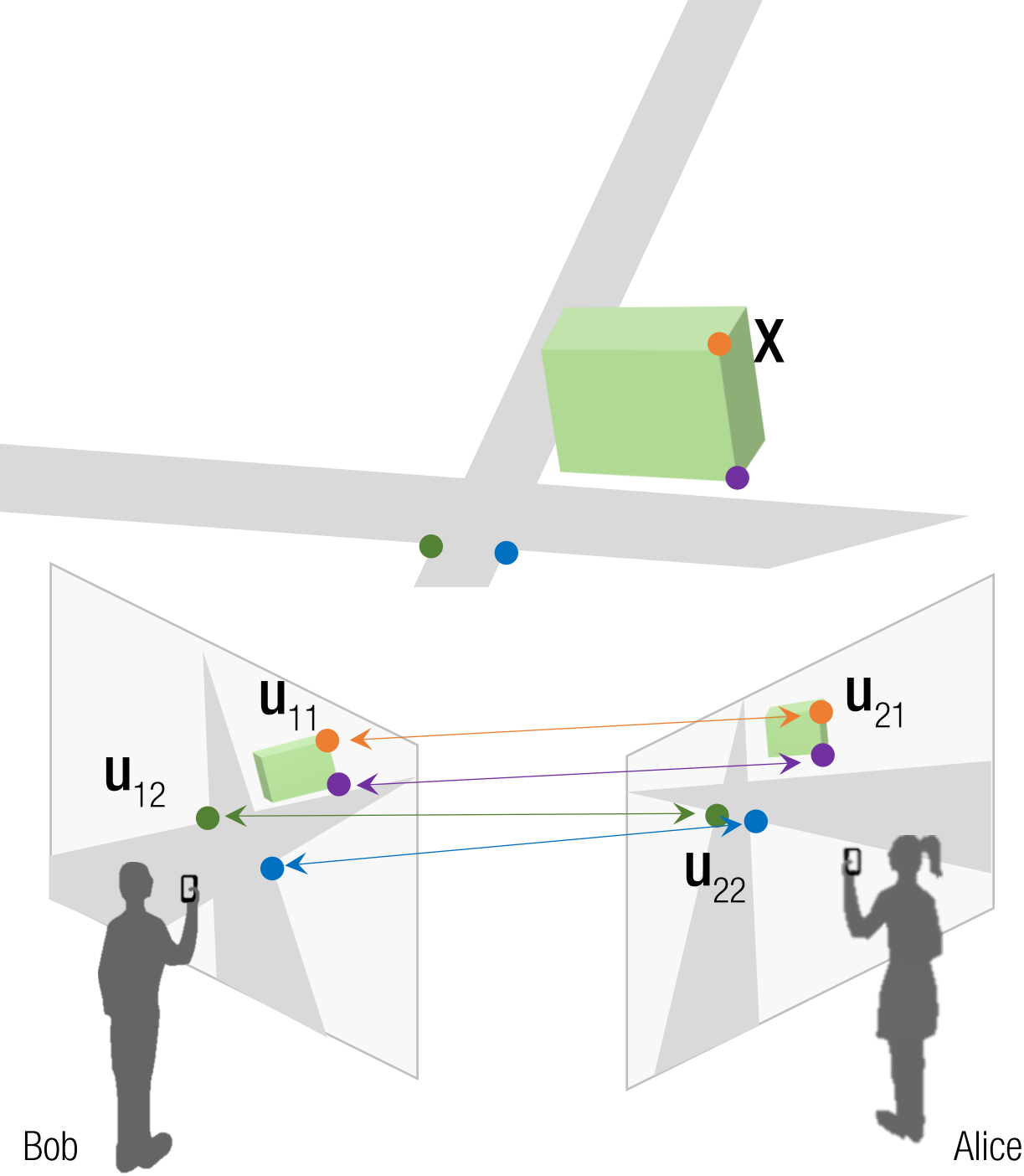
Orthographic projection:

$$\begin{aligned}\tilde{\mathbf{u}}_{11} &= \hat{\mathbf{R}}_1 \tilde{\mathbf{X}}_1 \\ \tilde{\mathbf{u}}_{12} &= \hat{\mathbf{R}}_1 \tilde{\mathbf{X}}_2\end{aligned}$$

$$\begin{aligned}\tilde{\mathbf{u}}_{11} &= \hat{\mathbf{R}}_2 \tilde{\mathbf{X}}_2 \\ \tilde{\mathbf{u}}_{22} &= \hat{\mathbf{R}}_2 \tilde{\mathbf{X}}_2\end{aligned}$$

where $\hat{\mathbf{R}} = \begin{bmatrix} r_{x1} & r_{x2} & r_{x3} \\ r_{y1} & r_{y2} & r_{y3} \end{bmatrix}$ $\hat{\mathbf{t}} = \begin{bmatrix} t_x \\ t_y \end{bmatrix}$

$$\tilde{\mathbf{X}}_j = \mathbf{X}_j - \frac{1}{P} \sum_j^P \mathbf{X}_j \longrightarrow \tilde{\mathbf{u}}_{ij} = \mathbf{u}_{ij} - \frac{1}{P} \sum_j^P \mathbf{u}_{ij}$$

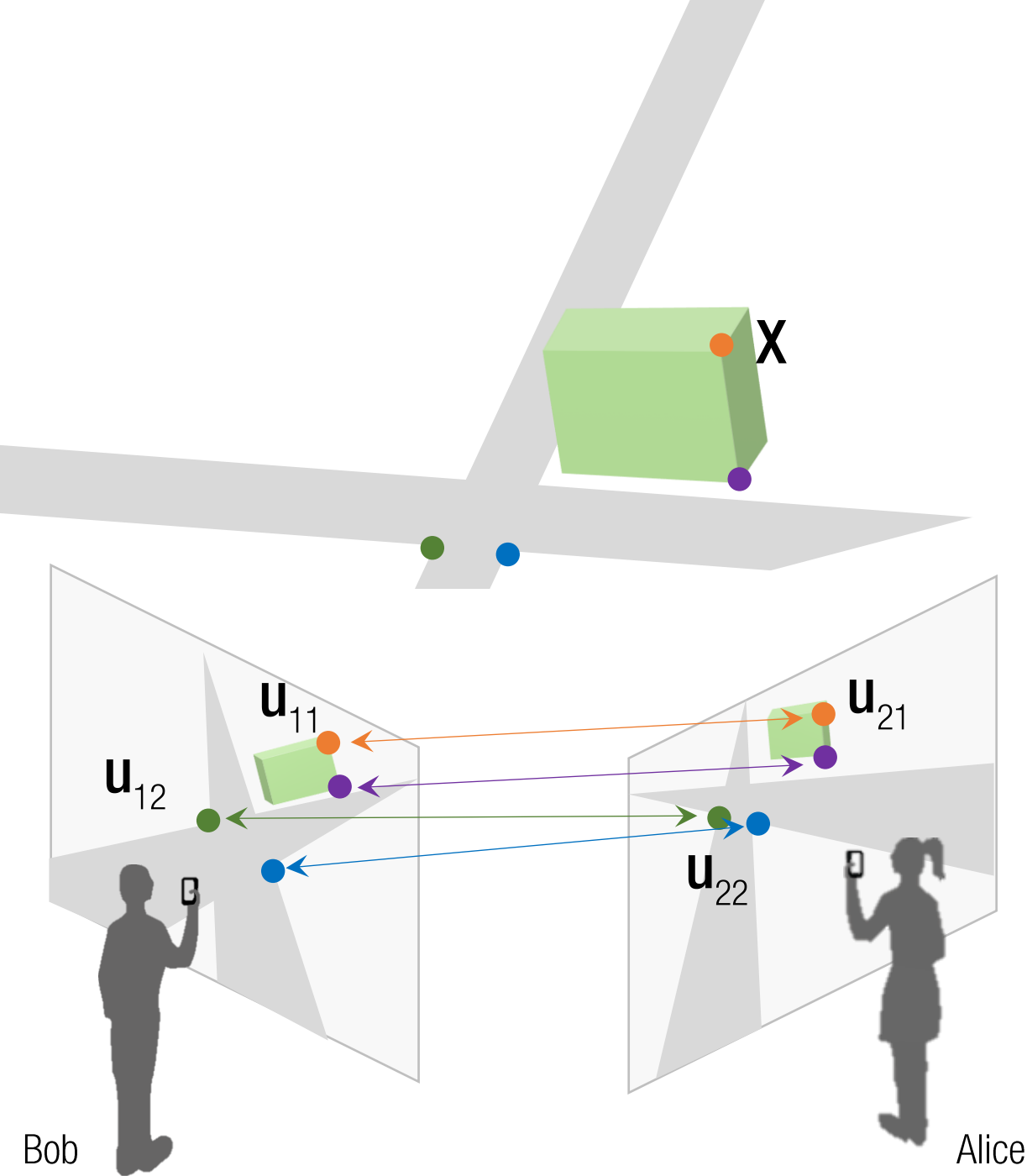


Orthographic projection:

$$\begin{aligned}\tilde{u}_{11} &= \hat{R}_1 \tilde{X}_1 \\ \tilde{u}_{12} &= \hat{R}_1 \tilde{X}_2\end{aligned}$$

$$\begin{aligned}\tilde{u}_{11} &= \hat{R}_2 \tilde{X}_2 \\ \tilde{u}_{22} &= \hat{R}_2 \tilde{X}_2\end{aligned}$$

$$\begin{bmatrix} \tilde{u}_{11} & \cdots & \tilde{u}_{1P} \\ \vdots & \ddots & \vdots \\ \tilde{u}_{F1} & \cdots & \tilde{u}_{FP} \end{bmatrix} = \begin{bmatrix} \hat{R}_1 \\ \vdots \\ \hat{R}_F \end{bmatrix} \begin{bmatrix} \tilde{X}_1 & \cdots & \tilde{X}_P \end{bmatrix}$$



Orthographic projection:

$$\begin{aligned}\tilde{u}_{11} &= \hat{R}_1 X_1 \\ \tilde{u}_{12} &= \hat{R}_1 X_2\end{aligned}$$

$$\begin{aligned}\tilde{u}_{11} &= \hat{R}_2 X_1 \\ \tilde{u}_{22} &= \hat{R}_2 X_2\end{aligned}$$

$$\begin{bmatrix} \tilde{u}_{11} & \cdots & \tilde{u}_{1P} \\ \vdots & \ddots & \vdots \\ \tilde{u}_{F1} & \cdots & \tilde{u}_{FP} \end{bmatrix} = \begin{bmatrix} \hat{R}_1 \\ \vdots \\ \hat{R}_F \end{bmatrix} \begin{bmatrix} \tilde{X}_1 & \cdots & \tilde{X}_P \end{bmatrix}$$

$2F \times P$ Knowns $2F \times 3$ Unknowns $3 \times P$

$$\begin{bmatrix} \tilde{u}_{11} & \cdots & \tilde{u}_{1P} \\ \vdots & \mathbf{W} & \vdots \\ \tilde{u}_{F1} & \cdots & \tilde{u}_{FP} \end{bmatrix} = \begin{bmatrix} \hat{R}_1 \\ \mathbf{M} \\ \hat{R}_F \end{bmatrix} \begin{bmatrix} \tilde{\mathbf{x}}_1 & \mathbf{S} & \tilde{\mathbf{x}}_P \end{bmatrix}$$

$2F \times P$ $2F \times 3$ $3 \times P$

Measurement matrix

Motion mtx

Shape mtx

$$\begin{bmatrix} \tilde{u}_{11} & \cdots & \tilde{u}_{1P} \\ \vdots & \mathbf{W} & \vdots \\ \tilde{u}_{F1} & \cdots & \tilde{u}_{FP} \end{bmatrix} = \begin{bmatrix} \hat{R}_1 \\ \mathbf{M} \\ \hat{R}_F \end{bmatrix} \begin{bmatrix} \tilde{x}_1 & \mathbf{S} & \tilde{x}_P \end{bmatrix}$$

$2F \times P$ $2F \times 3$ $3 \times P$

Measurement matrix

Motion mtx

Shape mtx

rank() = 3

$$\begin{bmatrix} \tilde{u}_{11} & \cdots & \tilde{u}_{1P} \\ \vdots & \mathbf{W} & \vdots \\ \tilde{u}_{F1} & \cdots & \tilde{u}_{FP} \end{bmatrix} = \begin{bmatrix} \hat{\mathbf{R}}_1 \\ \mathbf{M} \\ \hat{\mathbf{R}}_F \end{bmatrix} \begin{bmatrix} \tilde{\mathbf{x}}_1 & \mathbf{S} & \tilde{\mathbf{x}}_P \end{bmatrix}$$

$2F \times P$ $2F \times 3$ $3 \times P$

Measurement matrix

Motion mtx

Shape mtx

$\text{rank}(\text{yellow box}) = 3$

$$\text{yellow box } \mathbf{W} = \text{blue box } \mathbf{U} \text{ green box } \mathbf{D} \mathbf{V}^T \xrightarrow{?} \begin{matrix} \text{blue box } \mathbf{M} = \mathbf{U} \mathbf{D}^{1/2} \\ \text{green box } \mathbf{S} = \mathbf{D}^{1/2} \mathbf{V}^T \end{matrix}$$

$$\begin{bmatrix} \tilde{u}_{11} & \dots & \tilde{u}_{1P} \\ \vdots & \mathbf{W} & \vdots \\ \tilde{u}_{F1} & \dots & \tilde{u}_{FP} \end{bmatrix} = \begin{bmatrix} \hat{R}_1 \\ \mathbf{M} \\ \hat{R}_F \end{bmatrix} \begin{bmatrix} \tilde{x}_1 & \mathbf{S} & \tilde{x}_P \end{bmatrix}$$

$2F \times P$ $2F \times 3$ $3 \times P$

Measurement matrix

Motion mtx

Shape mtx

$\text{rank}(\text{yellow box}) = 3$

$$\text{yellow box} = \text{blue box} \text{UDV}^T \xrightarrow{?} \begin{matrix} \text{blue box} & \mathbf{M} = \mathbf{U}\mathbf{D}^{1/2} \\ \text{green box} & \mathbf{S} = \mathbf{D}^{1/2}\mathbf{V}^T \end{matrix}$$

No. There exists infinite \mathbf{M} and \mathbf{S} combinations.

$$\begin{matrix} \mathbf{M} = \mathbf{U}\mathbf{D}^{1/2}\mathbf{Q} \\ \mathbf{S} = \mathbf{Q}^{-1}\mathbf{D}^{1/2}\mathbf{V}^T \end{matrix} \quad \left| \right.$$

Any nonsingular \mathbf{Q} can produce solutions.

Is there any constraint?

$$\begin{bmatrix} \tilde{u}_{11} & \dots & \tilde{u}_{1P} \\ \vdots & \mathbf{W} & \vdots \\ \tilde{u}_{F1} & \dots & \tilde{u}_{FP} \end{bmatrix} = \begin{bmatrix} \hat{R}_1 \\ \mathbf{M} \\ \hat{R}_F \end{bmatrix} \begin{bmatrix} \tilde{x}_1 & \mathbf{S} & \tilde{x}_P \end{bmatrix}$$

$2F \times P$ $2F \times 3$ $3 \times P$

Measurement matrix

Motion mtx

Shape mtx

$\text{rank}(\text{yellow box}) = 3$

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

$$\mathbf{M} = \mathbf{U} \mathbf{D}^{1/2} \mathbf{Q}$$

$$\mathbf{S} = \mathbf{Q}^{-1} \mathbf{D}^{1/2} \mathbf{V}^T$$

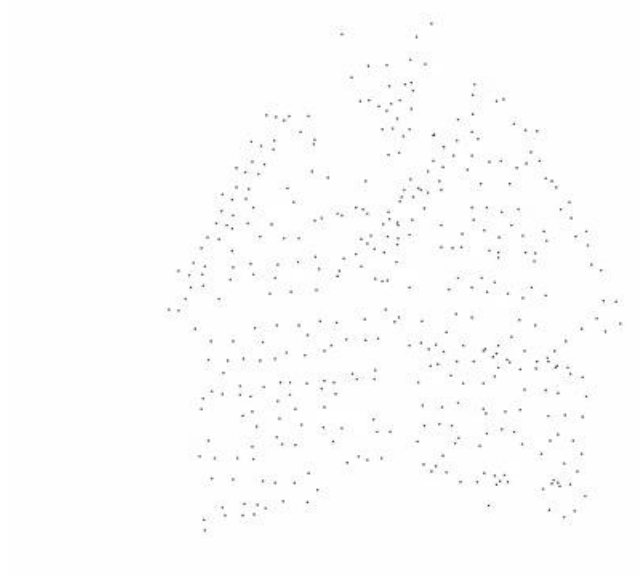
Orthogonal constraint:

$$\mathbf{M} = \mathbf{U} \mathbf{D}^{1/2} \mathbf{Q} = \begin{bmatrix} \mathbf{r}_{1x}^T \\ \mathbf{r}_{1y}^T \\ \vdots \\ \mathbf{r}_{Fx}^T \\ \mathbf{r}_{Fy}^T \end{bmatrix}$$

$$\rightarrow \mathbf{r}_{1x}^T \mathbf{r}_{1x} = 1, \quad \mathbf{r}_{1y}^T \mathbf{r}_{1y} = 1, \quad \mathbf{r}_{1x}^T \mathbf{r}_{1y} = 0$$

of unknowns: 9 (\mathbf{Q} : 3×3)

of equations: $3F$

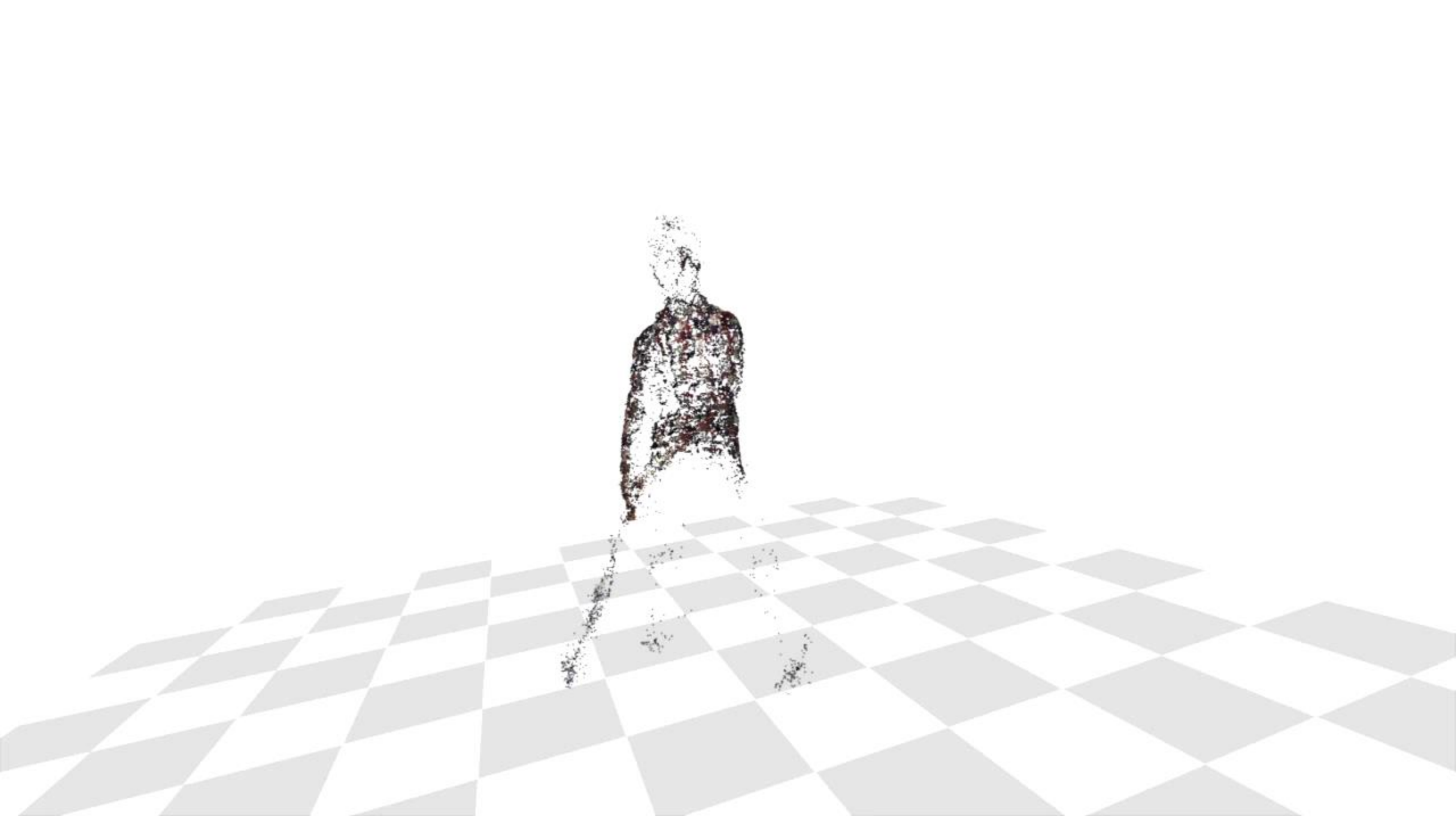


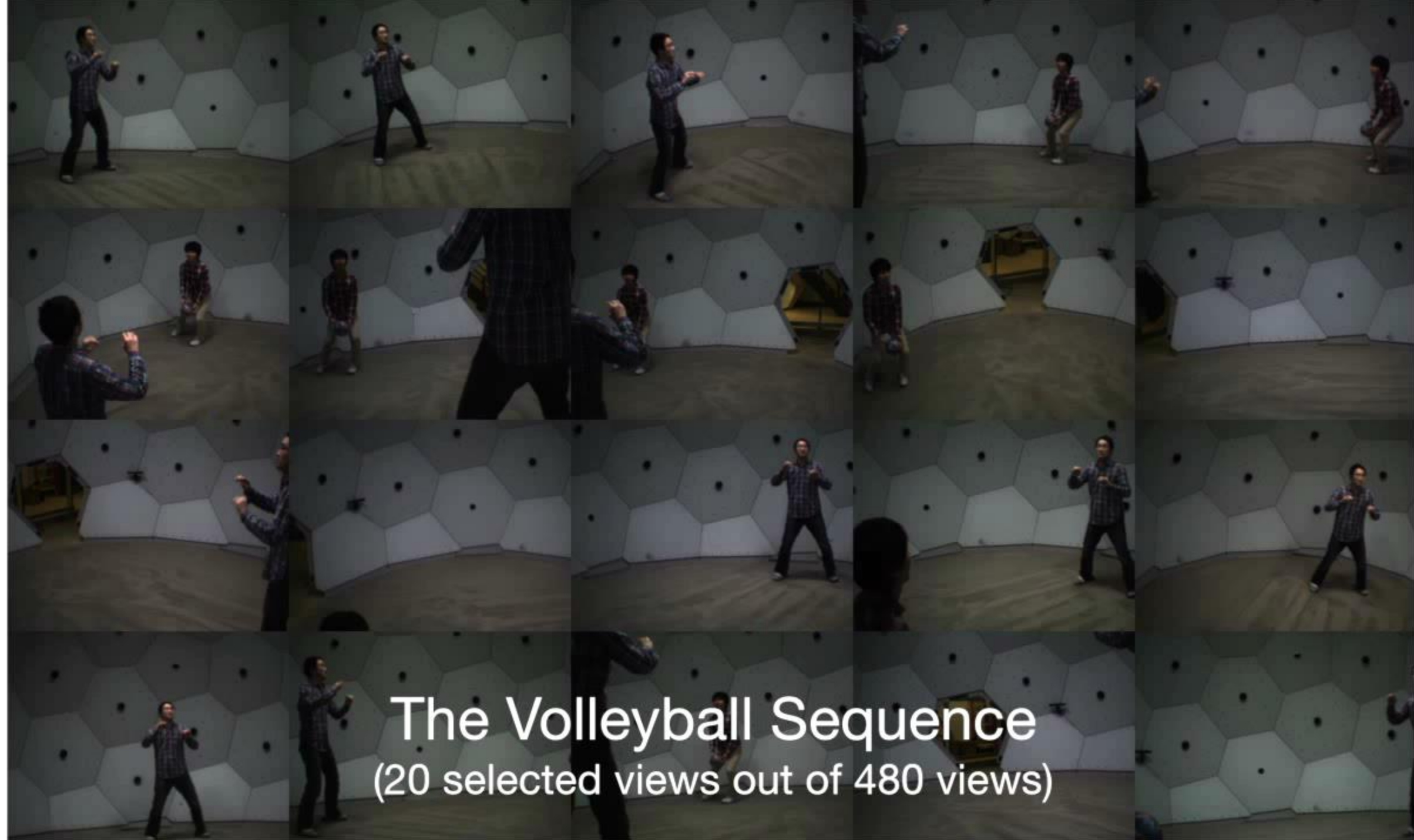
$$\begin{bmatrix} \tilde{u}_{11} & \dots & \tilde{u}_{1P} \\ \vdots & \mathbf{W} & \vdots \\ \tilde{u}_{F1} & \dots & \tilde{u}_{FP} \end{bmatrix} = \begin{bmatrix} \hat{R}_1 \\ \vdots \\ \hat{R}_F \end{bmatrix} \begin{bmatrix} \tilde{x}_1 & \mathbf{S} & \tilde{x}_P \end{bmatrix}$$

Large scale reconstruction









The Volleyball Sequence
(20 selected views out of 480 views)



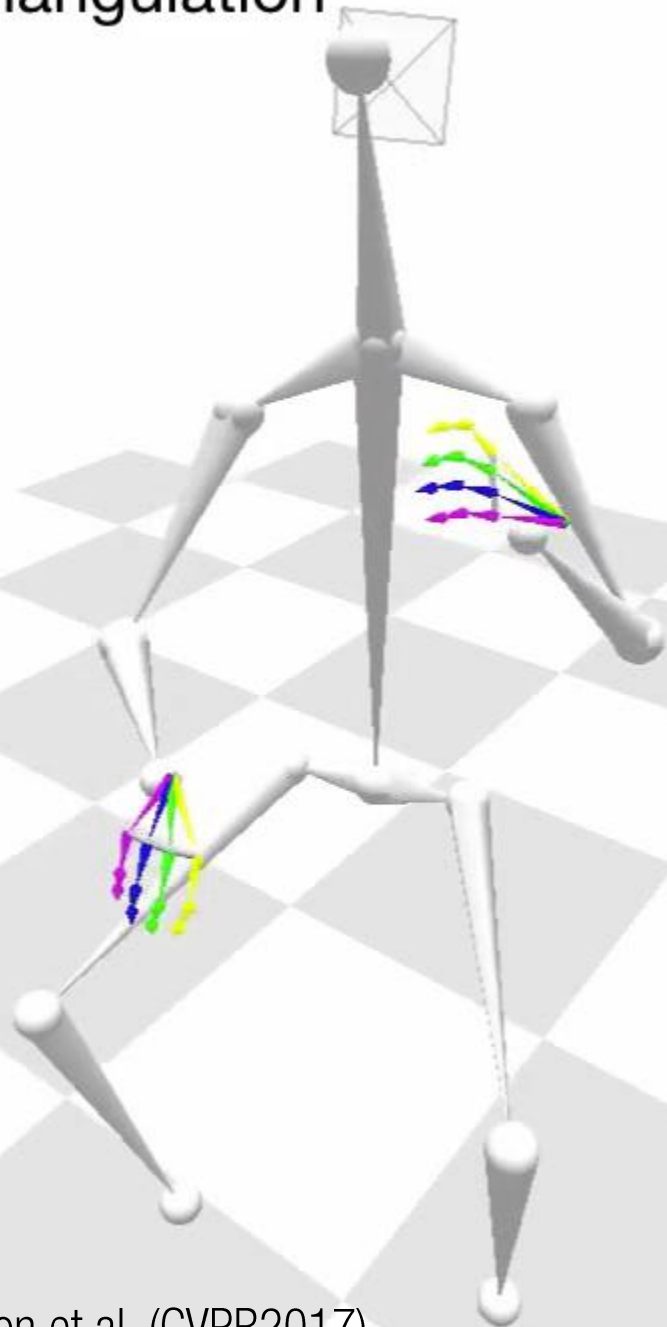
The Confetti Sequence

(20 selected views out of 480 views)





3D Triangulation



Simon et al. (CVPR2017)

Reprojections



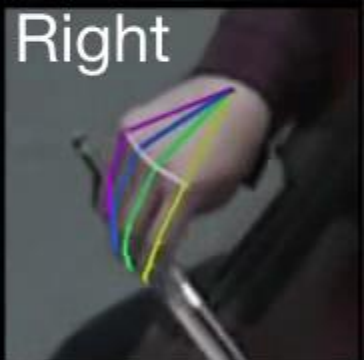
Right



Left



Right



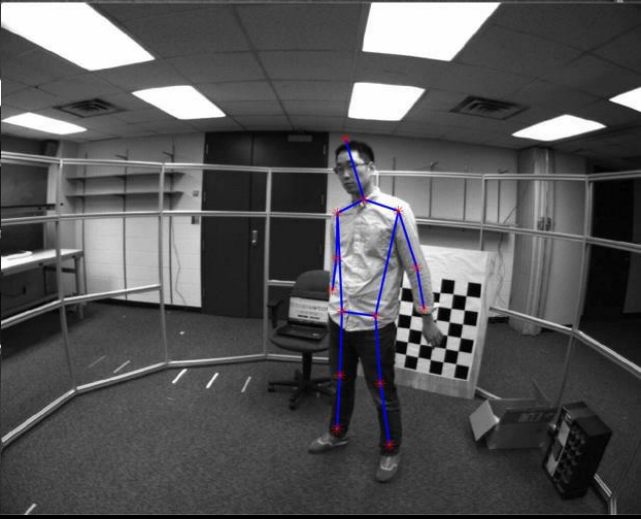
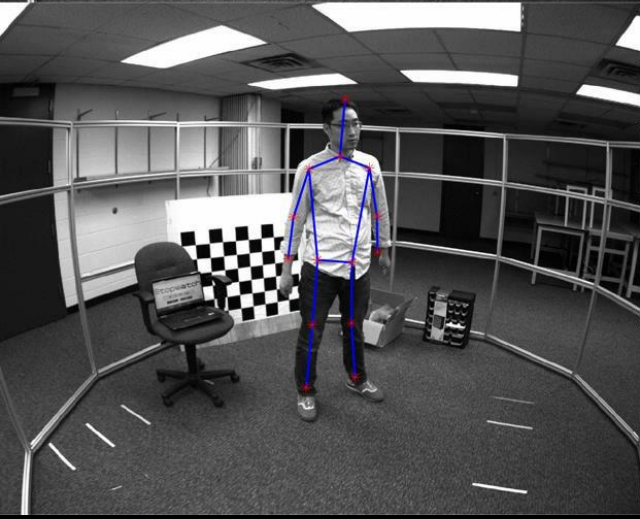
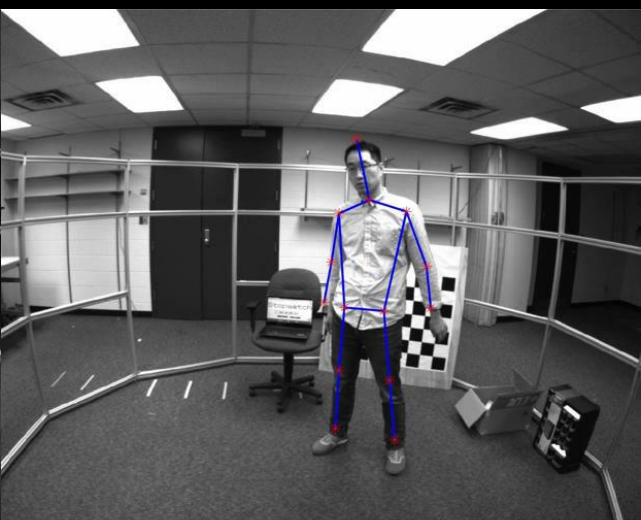
Left



@Shepherd Laboratory (UMN)



System design:
 $20 \text{ Pix/mm}^2 @ 100 \text{ fps}$



Field trip to Shepherd Laboratory