

# Computer Vision: Algorithms and Applications

*Motion and Photography*

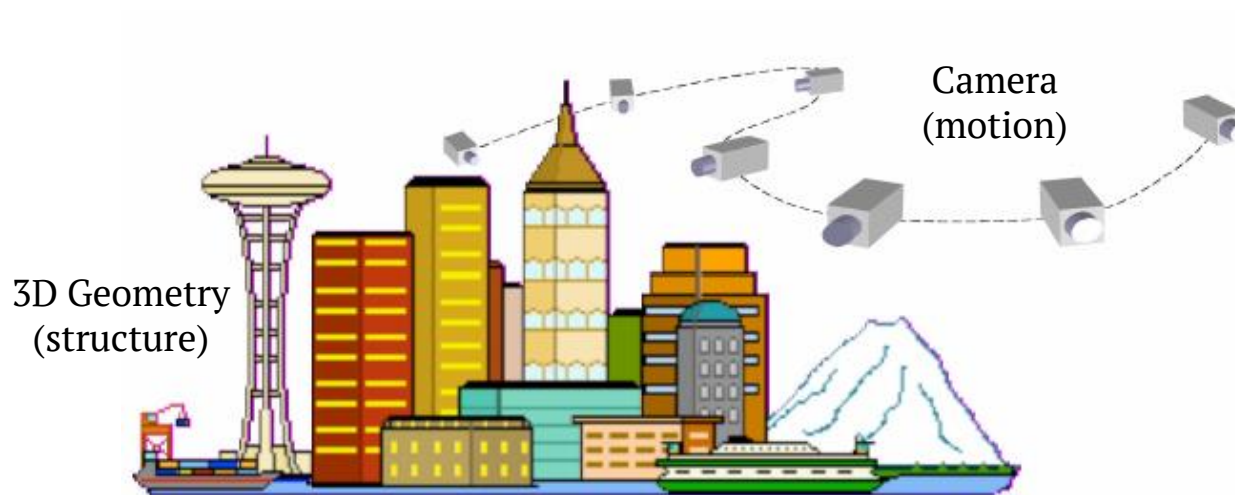
Jing Luo | Megvii Tech Talk | Mar 2018

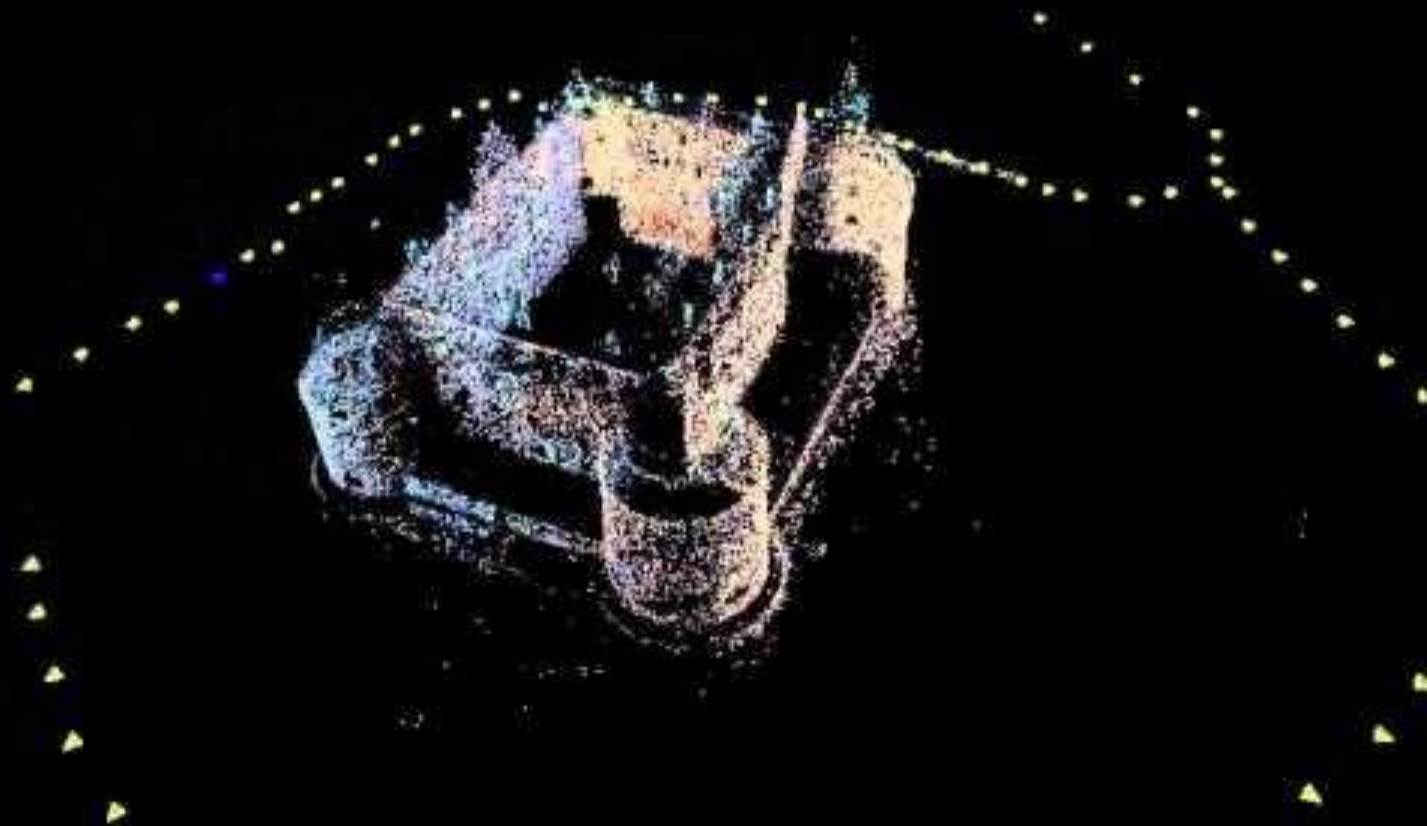
Reference: R. Szeliski. *Computer Vision: Algorithms and Applications*. 2010. 1.

1.  
Structure from motion

## Structure from motion

- Given a set of flow fields or displacement vectors from a moving camera time, determine:
  - the sequence of camera poses
  - the 3D structure of the scene

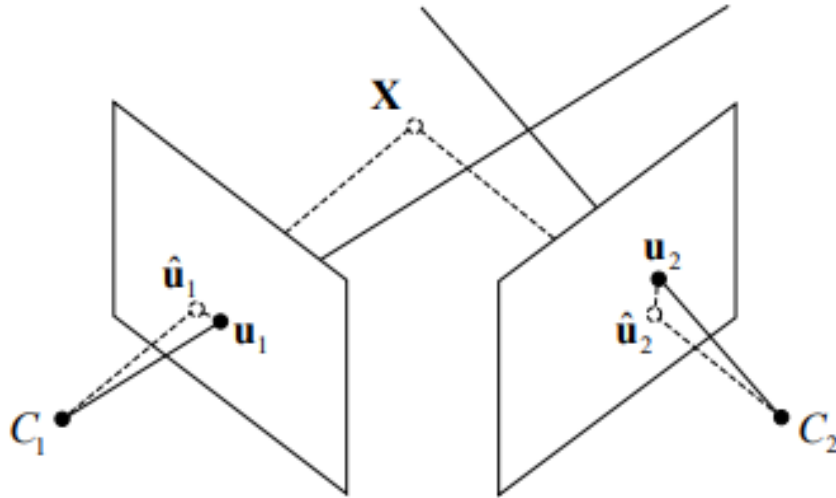






## Triangulation

- Determining a point's 3D position from a set of corresponding image locations and known camera positions.

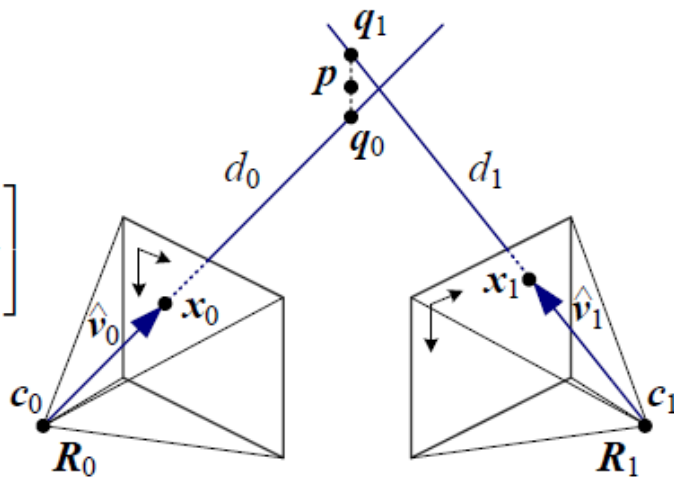


## Triangulation

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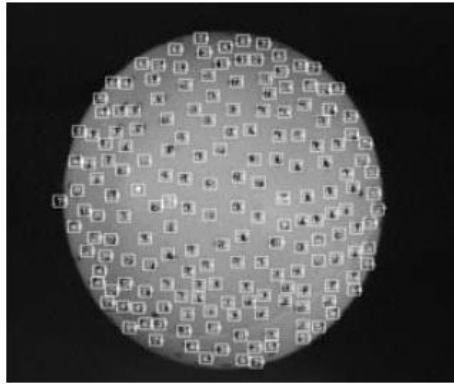
$$\|c_j + d_j \hat{v}_j - p\|^2$$

$$p = \left[ \sum_j (I - \hat{v}_j \hat{v}_j^T) \right]^{-1} \left[ \sum_j (I - \hat{v}_j \hat{v}_j^T) c_j \right]$$



## Factorization

- When processing video sequences, we often get extended feature tracks which it is possible to recover the structure and motion using a process called factorization.



(a)



(b)



(c)

**Figure 7.5** 3D reconstruction of a rotating ping pong ball using factorization (Tomasi and Kanade 1992) © 1992 Springer: (a) sample image with tracked features overlaid; (b) subsampled feature motion stream; (c) two views of the reconstructed 3D model.



## Factorization

### ▣ Assumption

- Pixel correspondence
  - via tracking
- Projection model
  - classic methods are orthographic
  - newer methods use perspective
- The positions of “P” points in “F” frames ( $F \geq 3$ ), which are not all coplanar, and have been tracked.



(a)



(b)



(c)



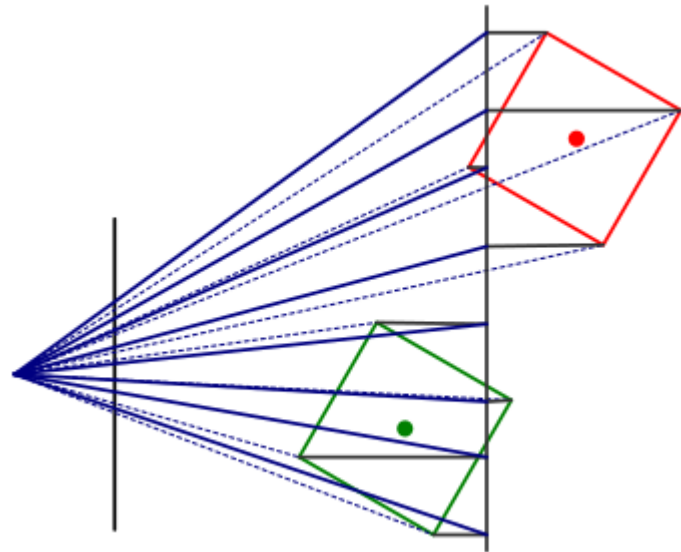
(d)

## 3D to 2D projections

### ▣ Orthography

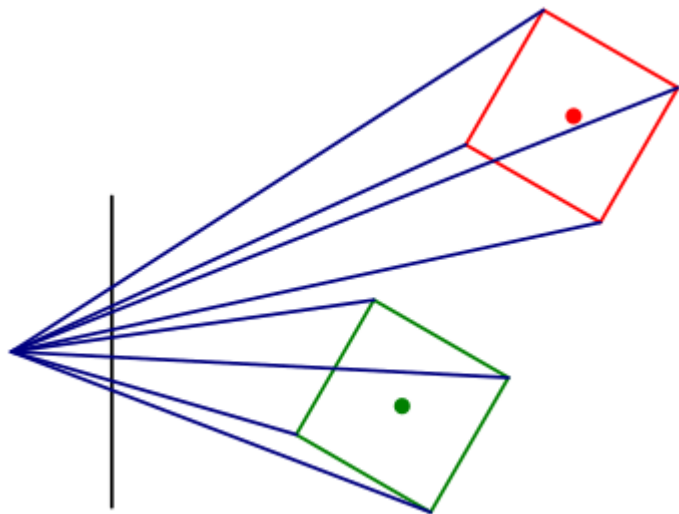
$$\mathbf{x} = [\mathbf{I}_{2 \times 2} | \mathbf{0}] \mathbf{p}$$

$$\tilde{\mathbf{x}} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \tilde{\mathbf{p}}$$



## 3D to 2D projections

### ▣ 3D perspective



$$\bar{\mathbf{x}} = \mathcal{P}_z(\mathbf{p}) = \begin{bmatrix} x/z \\ y/z \\ 1 \end{bmatrix}$$

$$\tilde{\mathbf{x}} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \tilde{\mathbf{p}}$$

## Factorization

### ▣ Feature points

$$\{(u_{fp}, v_{fp}) \mid f = 1, \dots, F, p = 1, \dots, P\}$$

$$W = \begin{bmatrix} u_{11} \dots u_{1P} \\ \vdots \\ u_{F1} \dots u_{FP} \\ v_{11} \dots v_{1P} \\ \vdots \\ v_{F1} \dots v_{FP} \end{bmatrix}$$

$$W = \begin{bmatrix} U \\ - \\ V \end{bmatrix}$$

## Factorization

- ▣ Mean normalize feature points

$$a_f = \frac{1}{P} \sum_{p=1}^P u_p \quad b_f = \frac{1}{P} \sum_{p=1}^P v_p$$

$$\tilde{u}_{fp} = u_{fp} - a_{fp}$$

$$\tilde{v}_{fp} = v_{fp} - b_{fp}$$

## Factorization

### ▣ Orthographic projection

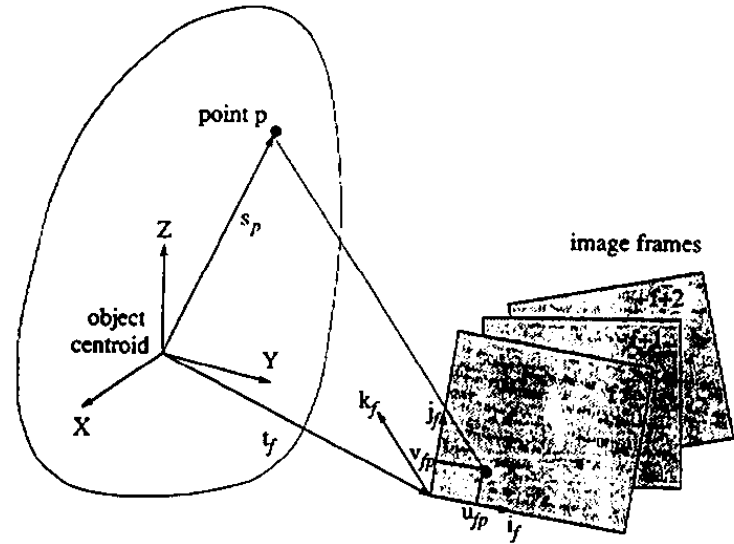
$$s_p = (X_p, Y_p, Z_p)$$

$$k_f = i_f \times j_f$$

$i, j, k$  are unit vectors along  $X, Y, Z$

$$u_{fp} = i_f^T (s_p - t_f)$$

$$v_{fp} = j_f^T (s_p - t_f)$$



## Factorization

### ▣ Subtract mean

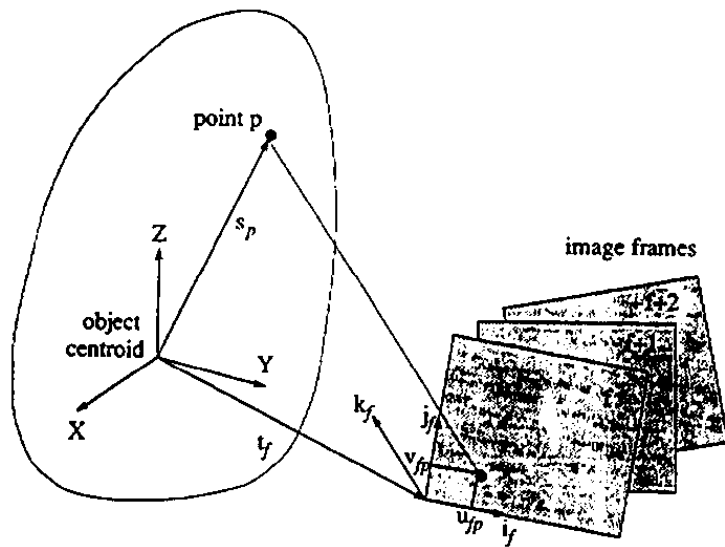
$$\tilde{u}_{fp} = u_{fp} - a_f$$

$$u_{fp} = i_f^T (s_p - t_f) \quad a_f = \frac{1}{P} \sum_{p=1}^P u_p$$

$$= i_f^T (s_p - t_f) - \frac{1}{P} \sum_{q=1}^P i_f^T (s_q - t_f)$$

$$= i_f^T \left[ s_p - \frac{1}{P} \sum_{q=1}^P s_q \right]$$

$$= i_f^T s_p$$



If Origin of world is at the centroid of object points,  
second term is zero.

## Factorization

### ▣ Measurement matrix

$$\tilde{u}_{fP} = i_f^T s_P$$

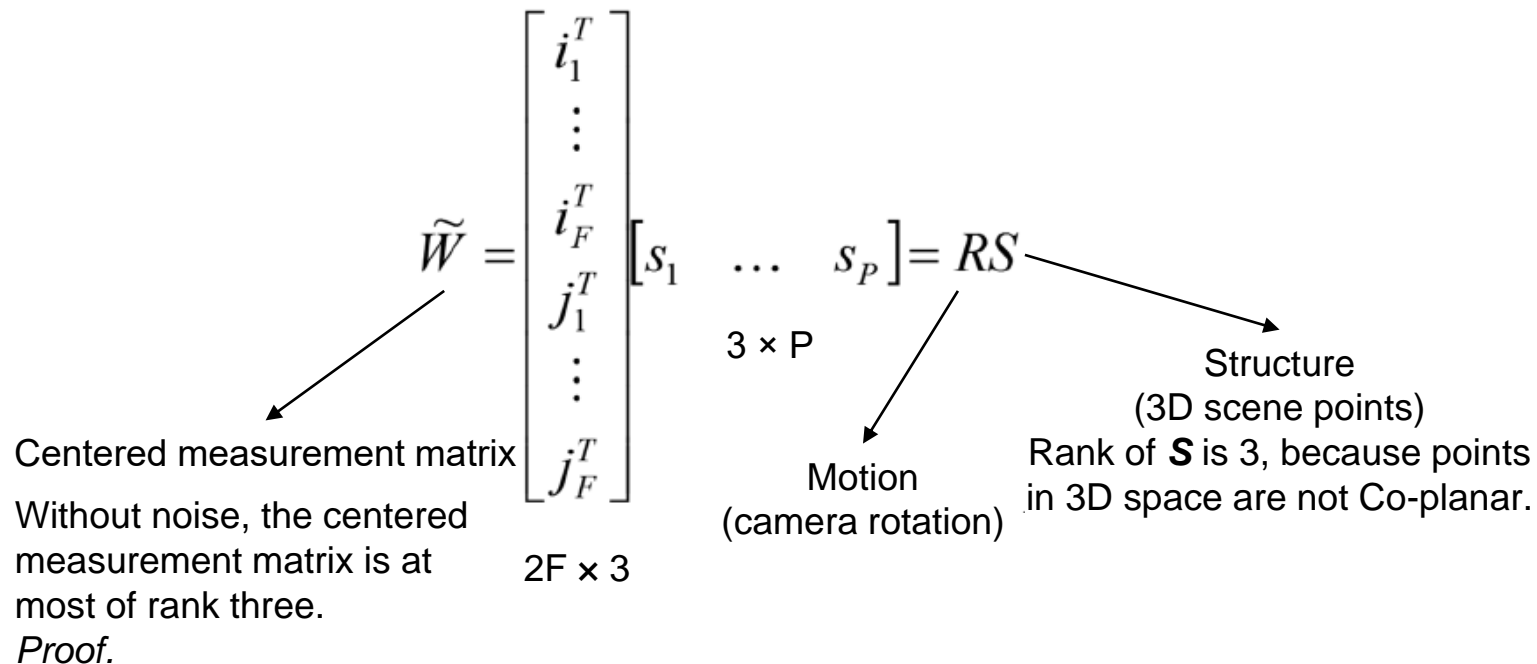
$$\tilde{v}_{fP} = j_f^T s_P$$

$$\tilde{W} = \begin{bmatrix} \tilde{U} \\ - \\ \tilde{V} \end{bmatrix} = \begin{bmatrix} \tilde{u}_{11} \dots \tilde{u}_{1P} \\ \vdots \\ \tilde{u}_{F1} \dots \tilde{u}_{FP} \\ \tilde{v}_{11} \dots \tilde{v}_{1P} \\ \vdots \\ \tilde{v}_{F1} \dots \tilde{v}_{FP} \end{bmatrix} \overset{2F \times 3}{=} \begin{bmatrix} i_1^T \\ \vdots \\ i_F^T \\ j_1^T \\ \vdots \\ j_F^T \end{bmatrix} \overset{3 \times P}{[s_1 \quad \dots \quad s_P]} = RS$$



## Factorization

### ▣ Rank theorem



## Factorization

- Importance of rank theorem
  - Shows that video data is highly redundant
  - Precisely quantifies the redundancy
  - Suggests an algorithm for solving SFM
- SVD

$$\tilde{W} = O_1 \Sigma O_2$$

- Theorem:
  - Any  $m$  by  $n$  matrix  $A$ , for which  $m \geq n$ , can be written as

$$A = O_1 \Sigma O_2$$

$m \times n \quad m \times n \quad n \times n \quad n \times n$

$\Sigma$  is diagonal

$O_1, O_2$  are orthogonal

$$O_1^T O_1 = O_2^T O_2 = I$$

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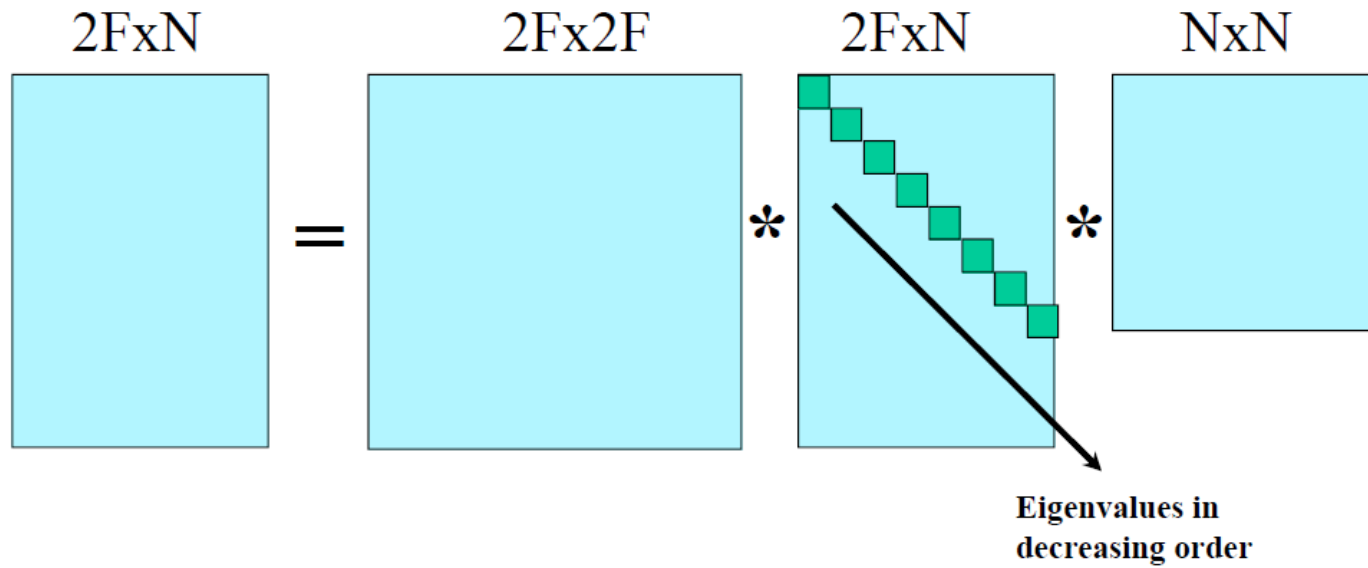
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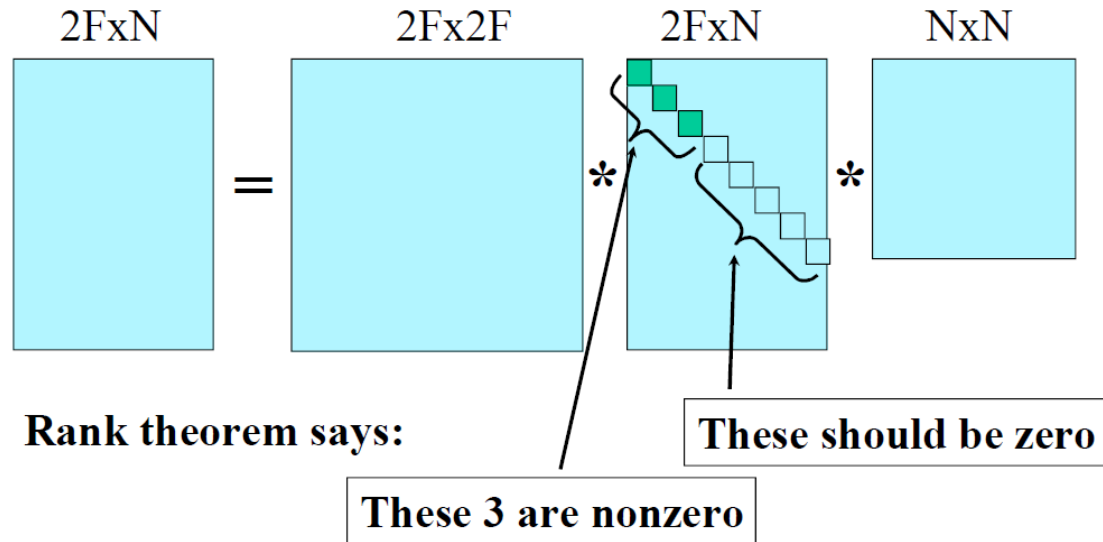
## Factorization

### ▣ SVD



## Factorization

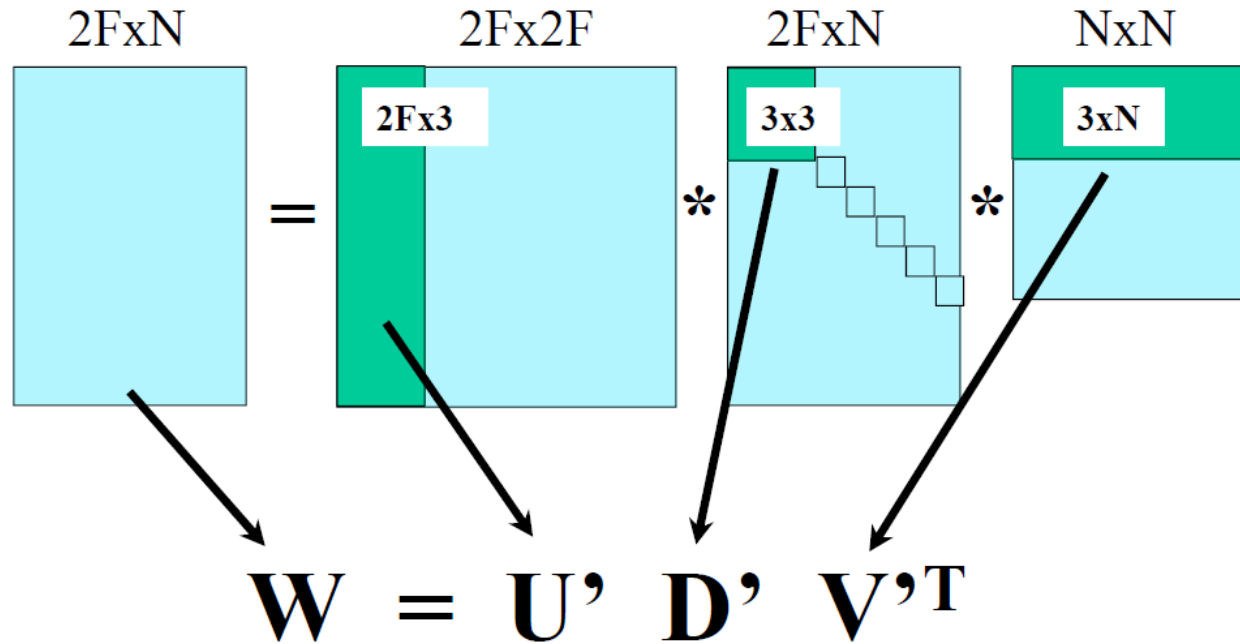
### ▣ SVD



In practice, due to noise, there may be more than 3 nonzero eigenvalues, but rank theorem tells us to ignore all but the largest three.

## Factorization

### ▣ SVD



## Factorization

### ▣ Factorization approach

Observed image points

$$\mathbf{W} \stackrel{\text{SVD}}{=} \mathbf{U}' \mathbf{D}' \mathbf{V}'^T$$

$$\mathbf{W} = \underbrace{\mathbf{U}' \mathbf{D}'^{1/2}}_{2F \times 3} \underbrace{\mathbf{D}'^{1/2} \mathbf{V}'^T}_{3 \times N}$$

$2F \times N$        $2F \times 3$        $3 \times N$

$$\mathbf{W} = \mathbf{M} \mathbf{S}$$

Camera motion      Scene structure

## Factorization

### ▣ Summary

#### **Assumptions**

- orthographic camera
- **N non-coplanar points tracking in  $F \geq 3$  frames**

**Form the centered measurement matrix  $W = [\tilde{X} ; \tilde{Y}]$**

- where  $\tilde{x}_{ij} = x_{ij} - mx_j$
- where  $\tilde{y}_{ij} = y_{ij} - my_j$
- $mx_j$  and  $my_j$  are mean of points in frame  $i$
- $j$  ranges over set of points

**Rank theorem: The centered measurement matrix has a rank of at most 3**



## Factorization

- 1) **Form the centered measurement matrix  $W$  from  $N$  points tracked over  $F$  frames.**
- 2) **Compute SVD of  $W = U D V^T$** 
  - $U$  is  $2F \times 2F$
  - $D$  is  $2F \times N$
  - $V^T$  is  $N \times N$
- 3) **Take largest 3 eigenvalues, and form**
  - $D' = 3 \times 3$  diagonal matrix of largest eigenvalues
  - $U' = 2F \times 3$  matrix of corresponding column vectors from  $U$
  - $V'^T = 3 \times N$  matrix of corresponding row vectors from  $V^T$
- 4) **Define**  
 $M = U' D'^{1/2}$  and  $S = D'^{1/2} V'^T$
- 5) **Solve for  $Q$  that makes appropriate rows of  $M$  orthogonal**
- 6) **Final solution is**  
 $M^* = M Q$  and  $S^* = Q^{-1} S$