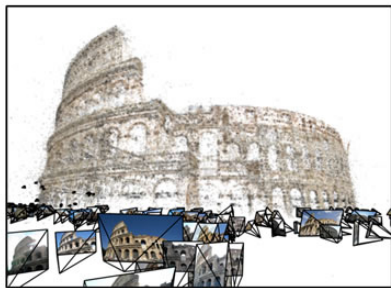


# CS325 Artificial Intelligence

## Computer Vision III – Structure from Motion (Ch. 24)

Dr. Cengiz Günay, Emory Univ.



What??

# Structure from Motion

What??

**Structure:** 3D information

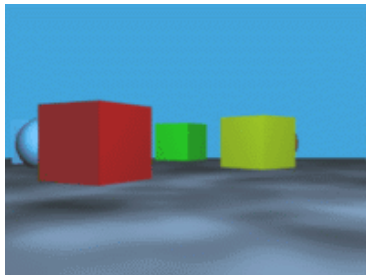
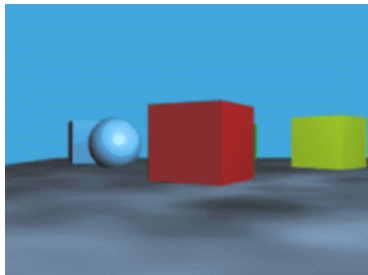
**Motion:** Camera motion

# Structure from Motion

What??

Structure: 3D information

Motion: Camera motion

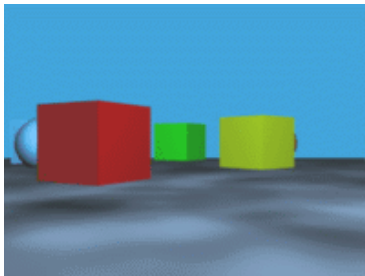
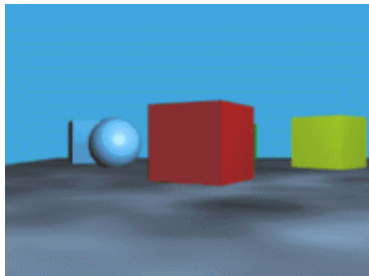


# Structure from Motion

What??

Structure: 3D information

Motion: Camera motion



Looks familiar?

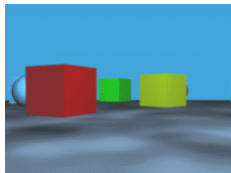
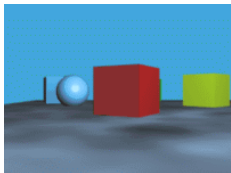
## Exit survey: Computer Vision II – 3D Vision

- Why don't we need to know the original object's size when we have stereo vision?
- What's the operating principle of the XBOX Kinect (R) motion tracker system?

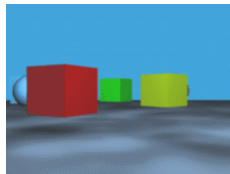
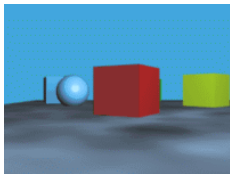
## Entry survey: Computer Vision III – Structure from Motion (0.25 pts)

- Can you think of a way to apply the 3D vision alignment algorithms from last class for extracting structure from motion (SfM)?
- What would be a good application area for SfM?

# Triangulate from Camera Positions



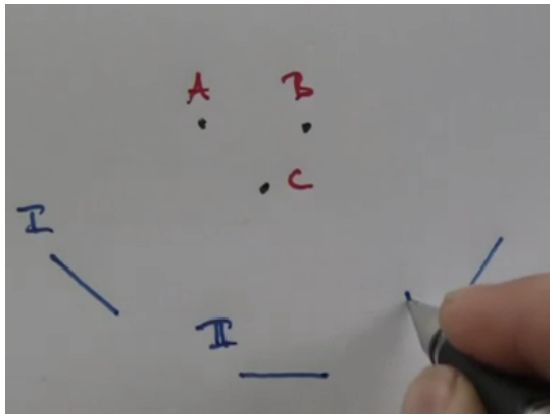
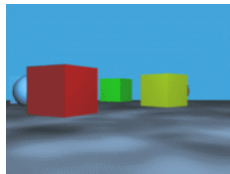
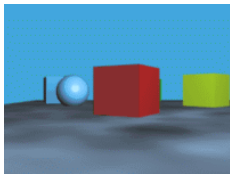
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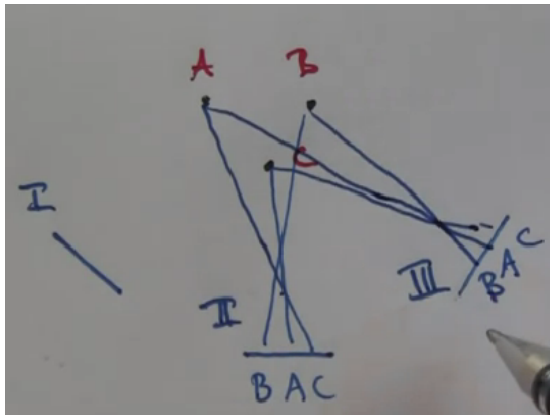
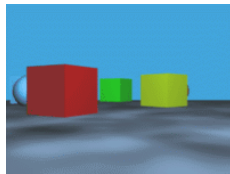
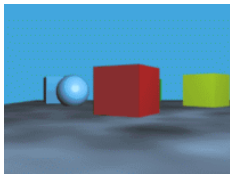
A    B  
•    •  
      • C



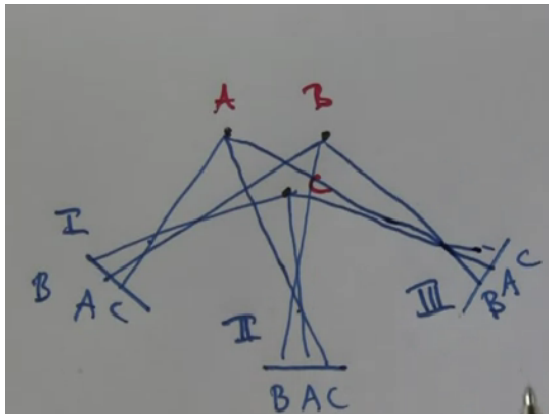
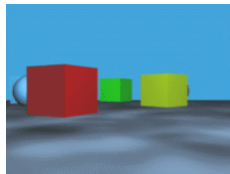
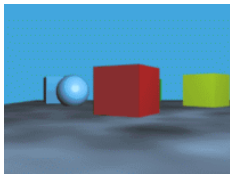
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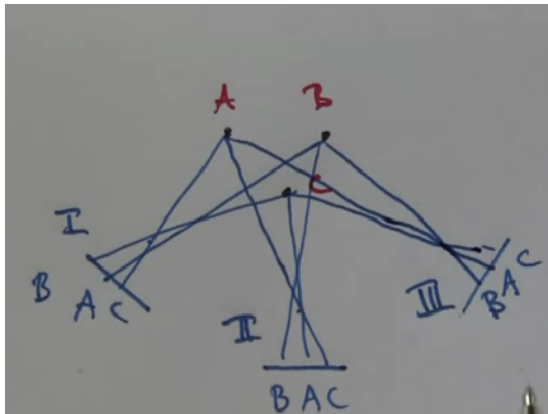
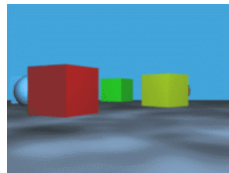
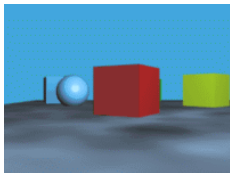
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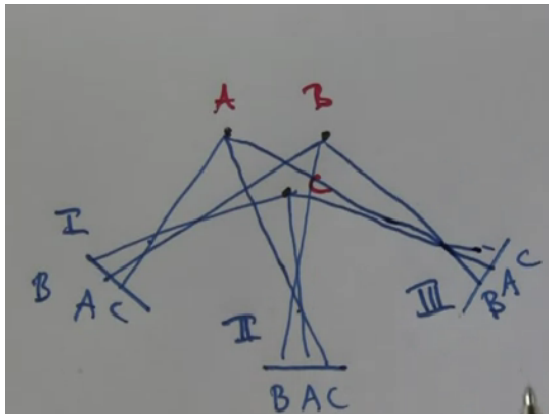
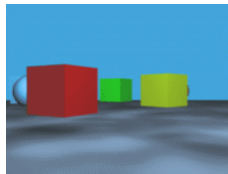
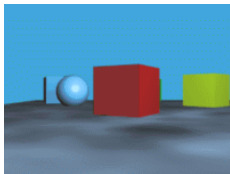
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Can we find locations of  $A, B, C$ ?

- ① Always
- ② Sometimes
- ③ Never

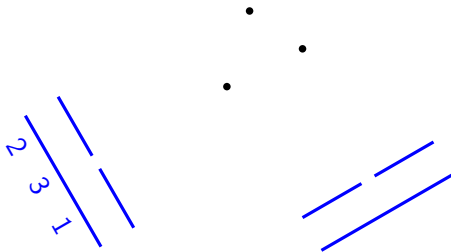
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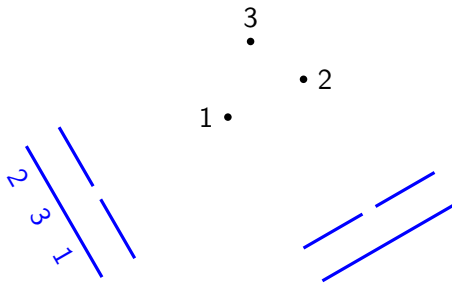
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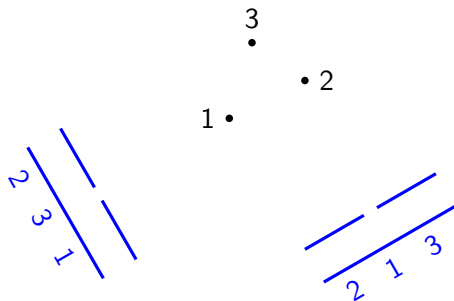
# Example with Two Cameras



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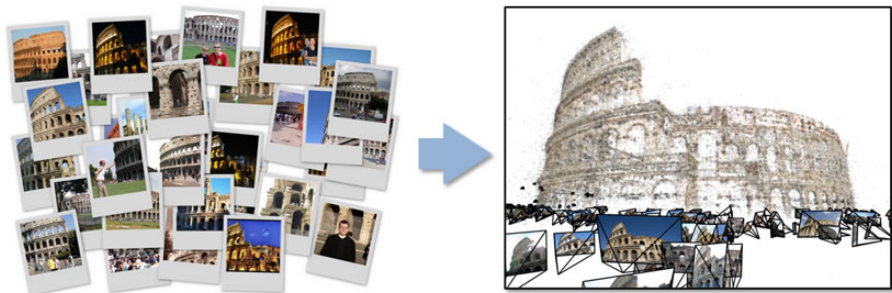


Demos:



[Tomasi & Kanade 92]

# SfM Examples: 3D Reconstruction From Snapshots



Lots of examples on the [Wikipedia page](#):

- A Fountain
- Duomo of Pisa
- An alley
- Dots and texture

# SfM is Also Called “Camera Tracking”

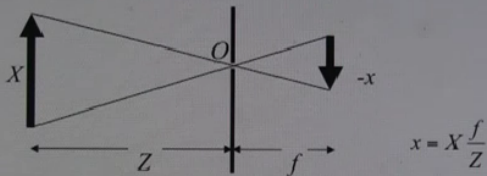
Nowadays, it is even available in open-source programs:

- **Blender** 3D modeling software:  
see [video of its camera tracking plugin](#)
- More on the [Wikipedia page](#)

# So How Does SfM Work?

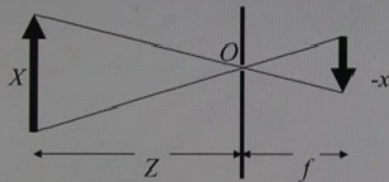
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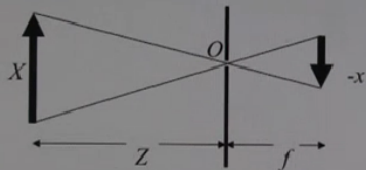


$$x = X \frac{f}{Z}$$

$$\begin{pmatrix} P_{x,j} \\ P_{y,j} \end{pmatrix} = f \frac{\begin{pmatrix} \cos \phi_i & \sin \phi_i & 0 \\ -\sin \phi_i & \cos \phi_i & 0 \end{pmatrix} \begin{pmatrix} \cos \varphi_i & 0 & \sin \varphi_i \\ 0 & 1 & 0 \\ -\sin \varphi_i & 0 & \cos \varphi_i \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \psi_i & \sin \psi_i \\ 0 & -\sin \psi_i & \cos \psi_i \end{pmatrix} \begin{pmatrix} P_{x,j} \\ P_{y,j} \\ P_{z,j} \end{pmatrix} + \begin{pmatrix} b_{x,j} \\ b_{y,j} \end{pmatrix}}{\begin{pmatrix} 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \cos \varphi_i & 0 & \sin \varphi_i \\ 0 & 1 & 0 \\ -\sin \varphi_i & 0 & \cos \varphi_i \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \psi_i & \sin \psi_i \\ 0 & -\sin \psi_i & \cos \psi_i \end{pmatrix} \begin{pmatrix} P_{x,j} \\ P_{y,j} \\ P_{z,j} \end{pmatrix} + b_{z,j}}$$

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$$\sum_{i,j} \left( \begin{array}{c} \left( \begin{array}{ccc} \cos \phi_i & \sin \phi_i & 0 \\ -\sin \phi_i & \cos \phi_i & 0 \end{array} \right) \left( \begin{array}{ccc} \cos \varphi_i & 0 & \sin \varphi_i \\ 0 & 1 & 0 \\ -\sin \varphi_i & 0 & \cos \varphi_i \end{array} \right) \left( \begin{array}{ccc} 1 & 0 & 0 \\ 0 & \cos \psi_i & \sin \psi_i \\ 0 & -\sin \psi_i & \cos \psi_i \end{array} \right) \begin{pmatrix} P_{x,j} \\ P_{y,j} \\ P_{z,j} \end{pmatrix} + \begin{pmatrix} b_{x,j} \\ b_{y,j} \end{pmatrix} \\ (0 \ 0 \ 1) \left( \begin{array}{ccc} \cos \varphi_i & 0 & \sin \varphi_i \\ 0 & 1 & 0 \\ -\sin \varphi_i & 0 & \cos \varphi_i \end{array} \right) \left( \begin{array}{ccc} 1 & 0 & 0 \\ 0 & \cos \psi_i & \sin \psi_i \\ 0 & -\sin \psi_i & \cos \psi_i \end{array} \right) \begin{pmatrix} P_{x,j} \\ P_{y,j} \\ P_{z,j} \end{pmatrix} + b_{z,j} \end{array} \right)^2 \rightarrow \min$$

Non-linear least-squares optimization problem:

- Gradient descent
- Conjugate gradient
- Gauss Newton methods (e.g., Levenberg-Marquardt)
- Singular Value Decomposition (e.g., PCA)



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Let's assume we have

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$n$  points to recover

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Which parameters can't we recover at all?

- Absolute frame of reference  $(x, y, z)$
- Absolute orientation angle  $(\alpha, \beta, \phi)$
- Scale

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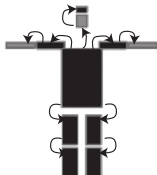
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Remember?



Need to have models of objects:



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SfM converts

- From camera images:
- To object locations:

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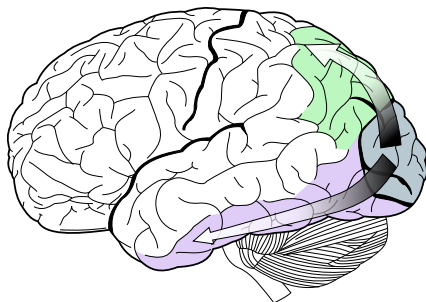
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- To object locations: **Allocentric** or object-centered representation

The brain has **two separate visual pathways** for these:



**Ventral** is allocentric and **dorsal** is egocentric. Read more [here](#).