

# CS231A Midterm Review

2/11/2021

# Agenda

- Exam logistics
- Preparation tips
- Core topics
- Review & practice question walkthrough

# Midterm logistics

- Format:
  - Online exam, on Canvas
  - 15 True/False questions, 10 multiple choice questions, and 4 short answer questions
- 2 hour time limit (+ 5 mins to upload)
- Practice exam on Canvas
- Available for 24 hour window @ 2/14
- *Open notes & open book*

# Preparing for the midterm

Resources:

- Lectures 1 - 10
- Lecture notes
- Homeworks

Again: open book & open notes!

- Focus on foundations & high-level understanding; you will have time to look up details.

# Core topics (1/2)

- General background
  - Homogeneous coordinates
  - Transformations
  - Formulating & solving least squares problems (when do we use an SVD?)
- Camera models
  - Perspective & non-perspective
  - Degrees of freedom
  - Distortion
  - Calibration
- Single view metrology
  - Vanishing points, lines

# Core topics (2/2)

- Multiview geometry
  - Epipolar geometry; essential and fundamental matrices; 8-point algorithm
  - Structure from motion
  - Stereo
  - Perspective, affine, similarity ambiguities
- Active and volumetric stereo
  - Structured lighting
  - Voxel coloring
- Fitting and matching
  - Least squares
  - RANSAC
  - Hough transforms
- Representation learning

(disclaimer: this guide is not meant to be comprehensive)

# Homogeneous Coordinates

Augmented space for writing coordinates:

$$\text{2D:} \quad \begin{bmatrix} x \\ y \end{bmatrix} \iff \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} \iff \begin{bmatrix} wx \\ wy \\ w \end{bmatrix}$$

$$\text{3D:} \quad \begin{bmatrix} x \\ y \\ z \end{bmatrix} \iff \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \iff \begin{bmatrix} wx \\ wy \\ wz \\ w \end{bmatrix}$$

# 2D Lines

Homogeneous coordinates give us a neat way of representing 2D lines as vectors/orthogonality constraints:

$$ax + by + c = 0$$
$$\begin{bmatrix} a & b & c \end{bmatrix} \begin{bmatrix} x & y & 1 \end{bmatrix}^T = 0$$

=> symmetry between lines and points

=> cross products suddenly becomes very useful!



# 2D Lines

How can we get the line connecting two points?

**Given:**

$$\begin{bmatrix} x_1 & y_1 & 1 \end{bmatrix}$$
$$\begin{bmatrix} x_2 & y_2 & 1 \end{bmatrix}$$

**Unknown:**

$$\begin{bmatrix} a & b & c \end{bmatrix}$$

**Subject to:**

$$\begin{bmatrix} a & b & c \end{bmatrix} \begin{bmatrix} x_1 & y_1 & 1 \end{bmatrix}^T = 0$$
$$\begin{bmatrix} a & b & c \end{bmatrix} \begin{bmatrix} x_2 & y_2 & 1 \end{bmatrix}^T = 0$$

**Solution:**

$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} x_1 \\ y_1 \\ 1 \end{bmatrix} \times \begin{bmatrix} x_2 \\ y_2 \\ 1 \end{bmatrix}$$

# 2D Lines

How can we get the intersection of two lines?

**Given:**

$$\begin{bmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \end{bmatrix}$$

**Unknown:**

$$\begin{bmatrix} x & y & 1 \end{bmatrix}$$

**Subject to:**

$$\begin{bmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \end{bmatrix} \begin{bmatrix} x & y & 1 \end{bmatrix}^T = 0$$
$$\begin{bmatrix} a_2 & b_2 & c_2 \end{bmatrix} \begin{bmatrix} x & y & 1 \end{bmatrix}^T = 0$$

**Solution:**

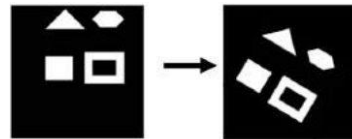
$$\begin{bmatrix} wx \\ wy \\ w \end{bmatrix} = \begin{bmatrix} a_1 \\ b_1 \\ c_1 \end{bmatrix} \times \begin{bmatrix} a_2 \\ b_2 \\ c_2 \end{bmatrix}$$

# Transformations

## Isometric transformations:

Distances preserved

$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} R & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$



## Similarity transformations:

Shapes preserved

$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} SR & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}, \quad S = \begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix}$$



## Affine transformations:

Parallelism preserved

$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} A & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$



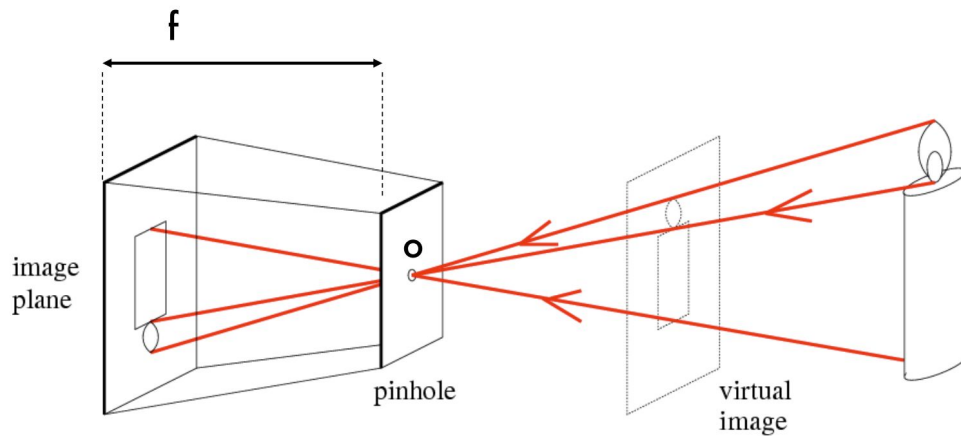
## Projective transformations:

Lines preserved

$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} A & t \\ v & b \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$



# Pinhole Cameras

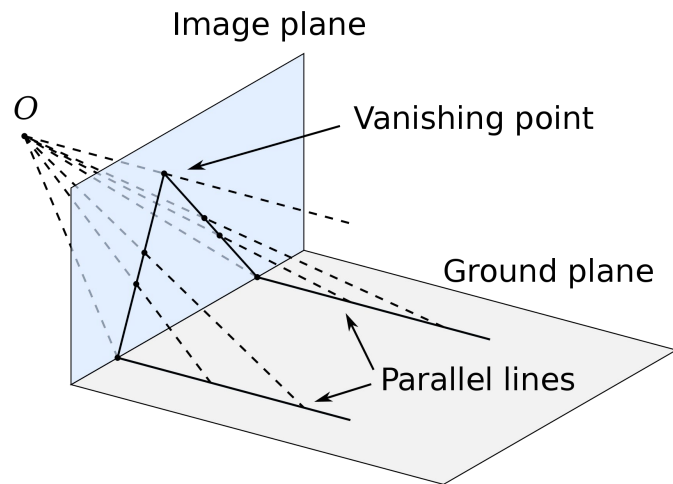


$$P' = K \begin{bmatrix} R & T \end{bmatrix} P_w$$

$$K = \begin{bmatrix} f & 0 & c_x \\ 0 & f & c_y \\ 0 & 0 & 1 \end{bmatrix}$$

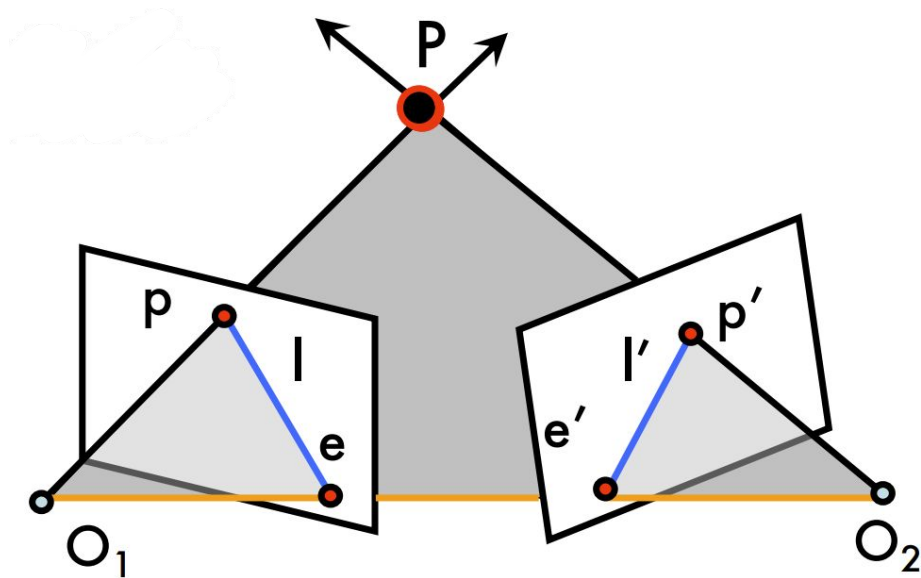
# Single View Metrology

Under projective transformation, parallel lines converge to a vanishing point:



We used this for camera calibration in your homework!

# Epipolar Geometry



## Essential matrix:

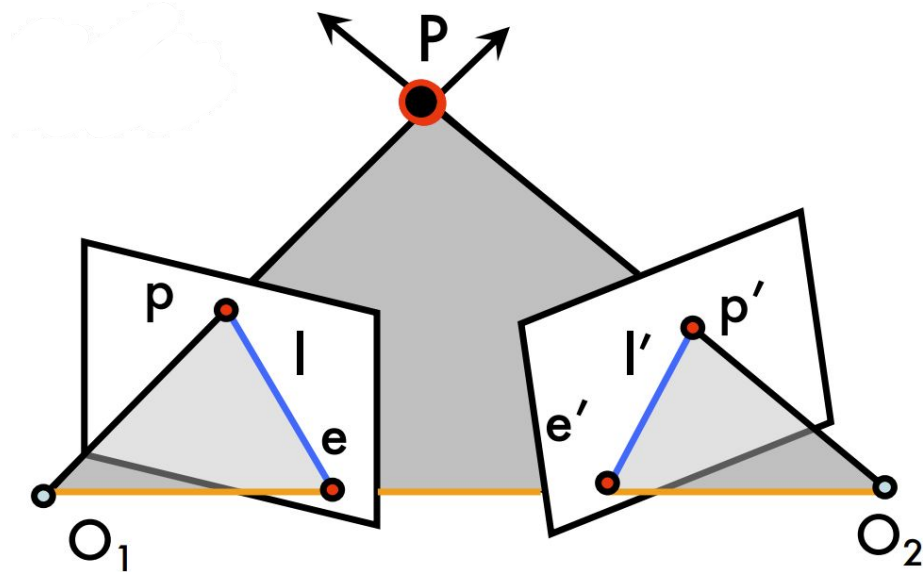
A point  $\rightarrow$  epipolar line mapping  
for canonical cameras ( $K = I$ )

$$l' = E^T p$$

$$l = E p'$$

$$p^T E p' = 0$$

# Epipolar Geometry



Given a 2D point in one camera, a correspondence in the other must lie on an epipolar line

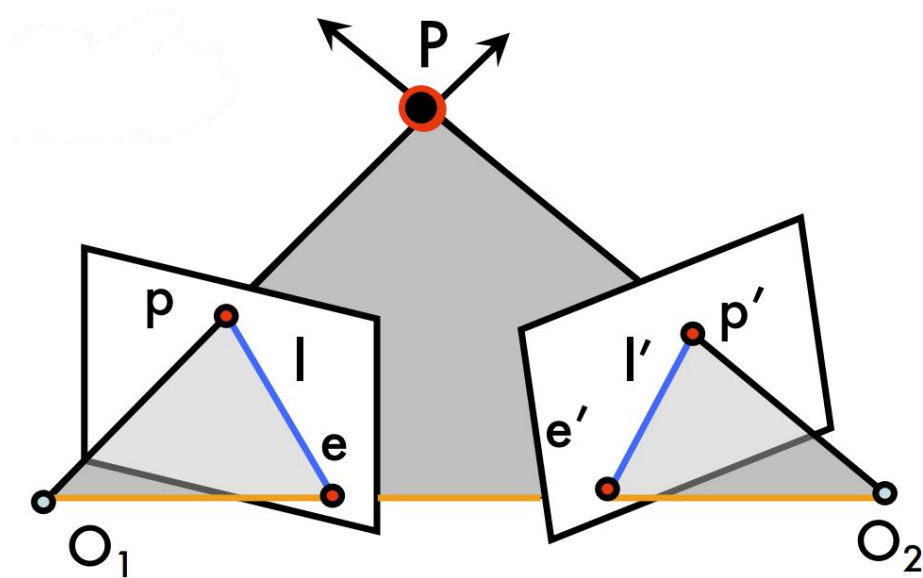
If  $\mathbf{p}'$  is known, we can compute  $\mathbf{l}$  and search for  $\mathbf{p}$  using:

$$\mathbf{l}^T \mathbf{p} = 0$$

If  $\mathbf{p}$  is known, we can compute  $\mathbf{l}'$  and search for  $\mathbf{p}'$  using:

$$\mathbf{l}'^T \mathbf{p}' = 0$$

# Epipolar Geometry



## Fundamental matrix:

A point  $\rightarrow$  epipolar line mapping  
for general projective cameras

$$l' = F^T p$$

$$l = F p'$$

$$p^T F p' = 0$$



# Epipolar Geometry

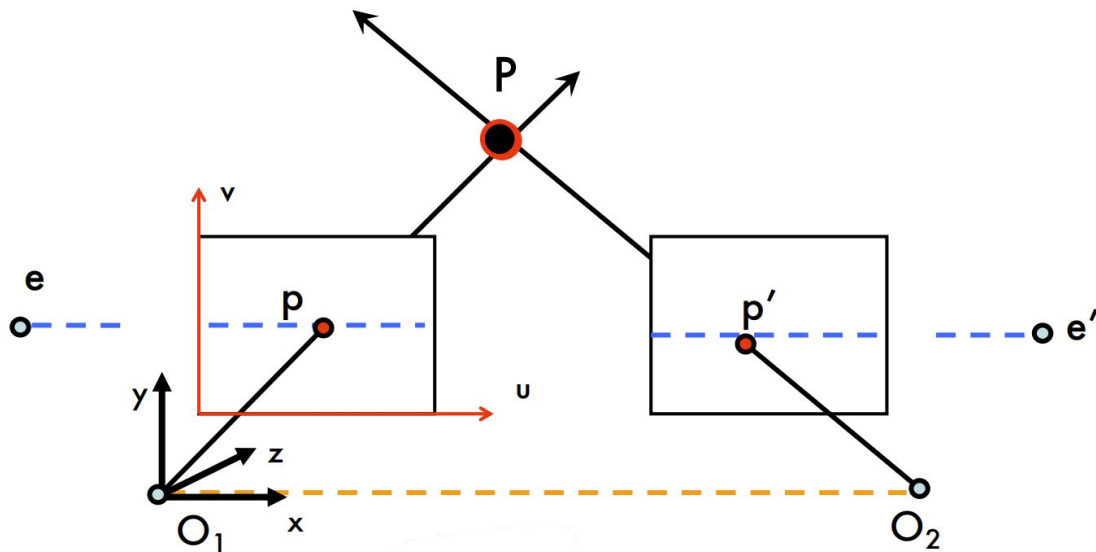
Computing the fundamental matrix with the 8-point algorithm:

$$\mathbf{p}^T \mathbf{F} \mathbf{p}' = 0$$

$$\begin{bmatrix} u_1 u'_1 & v_1 u'_1 & u'_1 & u_1 v'_1 & v_1 v'_1 & v'_1 & u_1 & v_1 & 1 \\ u_2 u'_2 & v_2 u'_2 & u'_2 & u_2 v'_2 & v_2 v'_2 & v'_2 & u_2 & v_2 & 1 \\ u_3 u'_3 & v_3 u'_3 & u'_3 & u_3 v'_3 & v_3 v'_3 & v'_3 & u_3 & v_3 & 1 \\ u_4 u'_4 & v_4 u'_4 & u'_4 & u_4 v'_4 & v_4 v'_4 & v'_4 & u_4 & v_4 & 1 \\ u_5 u'_5 & v_5 u'_5 & u'_5 & u_5 v'_5 & v_5 v'_5 & v'_5 & u_5 & v_5 & 1 \\ u_6 u'_6 & v_6 u'_6 & u'_6 & u_6 v'_6 & v_6 v'_6 & v'_6 & u_6 & v_6 & 1 \\ u_7 u'_7 & v_7 u'_7 & u'_7 & u_7 v'_7 & v_7 v'_7 & v'_7 & u_7 & v_7 & 1 \\ u_8 u'_8 & v_8 u'_8 & u'_8 & u_8 v'_8 & v_8 v'_8 & v'_8 & u_8 & v_8 & 1 \end{bmatrix} \begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix} = 0$$

=> Solve with SVD,  
then project to rank 2

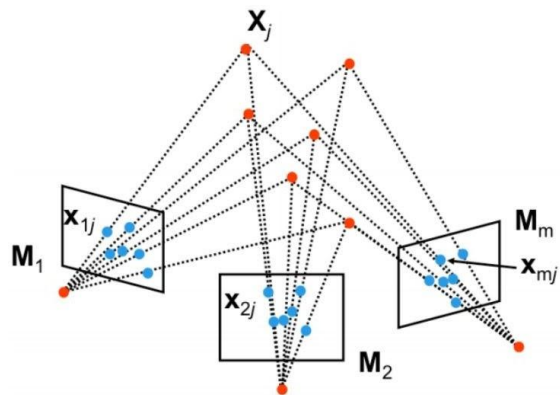
# Epipolar Geometry



## Parallel images planes or rectification:

simplifies correspondence problem, moves epipoles to infinity

# Structure from Motion



Determining structure *and* motion

You've implemented a few algorithms for this!

- Factorization, triangulation

# Factorization Method

$$\mathbf{D} = \begin{bmatrix} \hat{\mathbf{x}}_{11} & \hat{\mathbf{x}}_{12} & \cdots & \hat{\mathbf{x}}_{1n} \\ \hat{\mathbf{x}}_{21} & \hat{\mathbf{x}}_{22} & \cdots & \hat{\mathbf{x}}_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \hat{\mathbf{x}}_{m1} & \hat{\mathbf{x}}_{m2} & \cdots & \hat{\mathbf{x}}_{mn} \end{bmatrix} = \begin{bmatrix} \mathbf{A}_1 \\ \mathbf{A}_2 \\ \vdots \\ \mathbf{A}_m \end{bmatrix} \begin{bmatrix} \mathbf{X}_1 & \mathbf{X}_2 & \cdots & \mathbf{X}_n \end{bmatrix} \quad [\text{Eq. 10}]$$

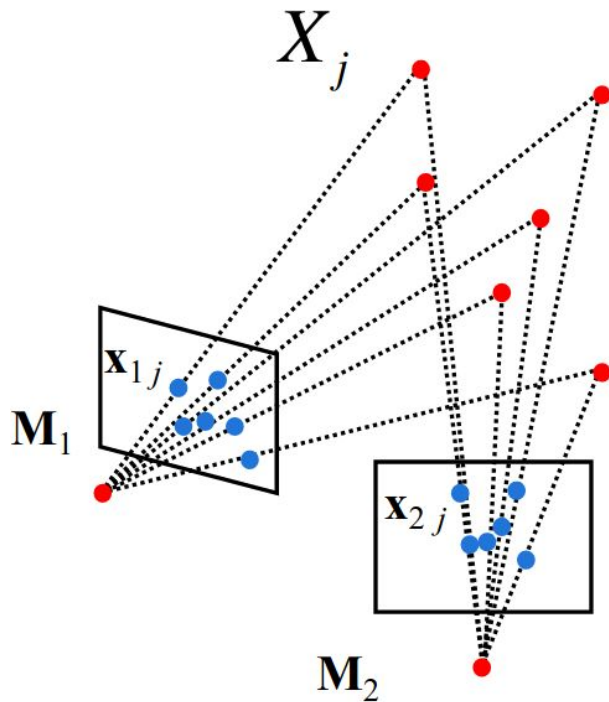
$(2m \times n)$   $(2m \times 3)$   $\mathbf{M}$   $\text{points } (3 \times n)$   $\mathbf{S}$

Assume all points are visible

# Algebraic approach

- Compute fundamental matrix  $F$
- Use  $F$  to estimate projective cameras
- Use these cameras to triangulate and estimate points in 3D

Works with 2 views



# Bundle Adjustment

Non-linear method for refining structure and motion

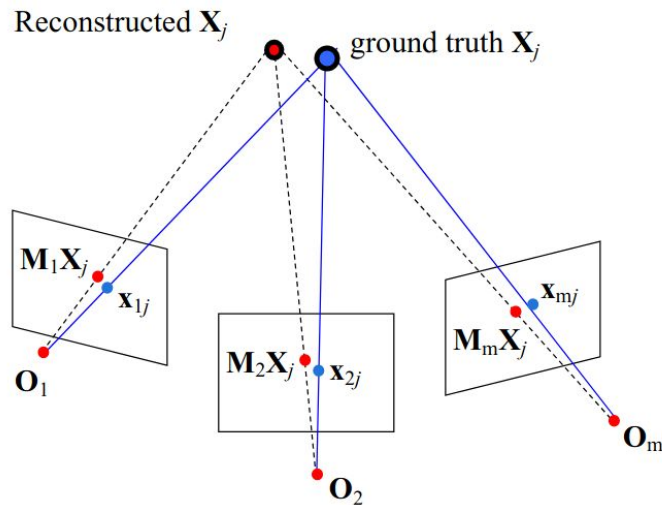
Goal: minimize reprojection error

Advantages

- Handle large number of views
- Handle missing data

Limitations

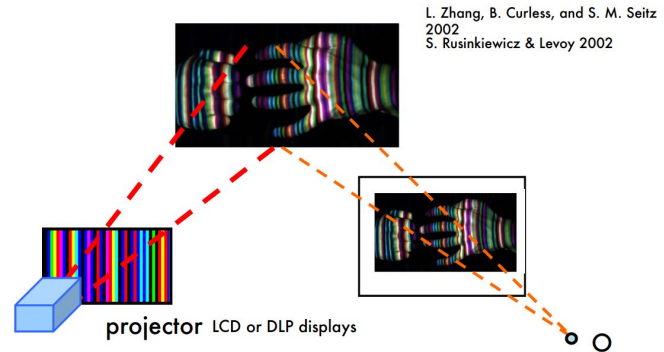
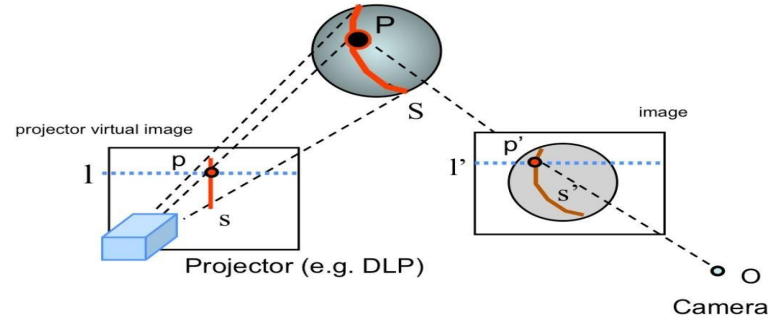
- Large minimization problem
- Require good initialization



# Active Stereo

## Active Stereo

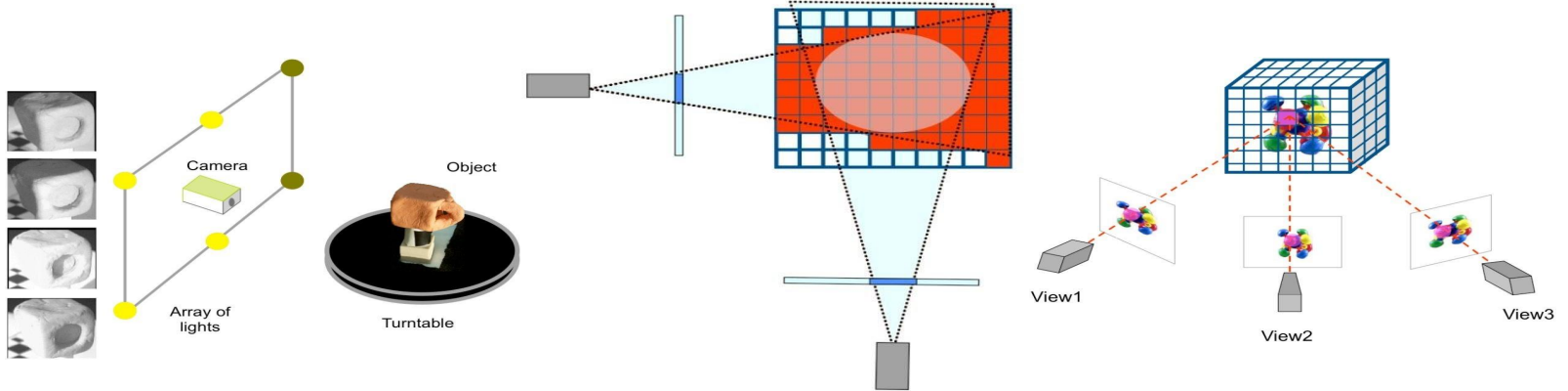
- Replaces one camera in a stereo pair with a projector
- Solves matching problem



# Volumetric Stereo

## Volumetric Stereo

- Space carving
  - Cannot carve concavity
- Shadow carving
- Voxel coloring

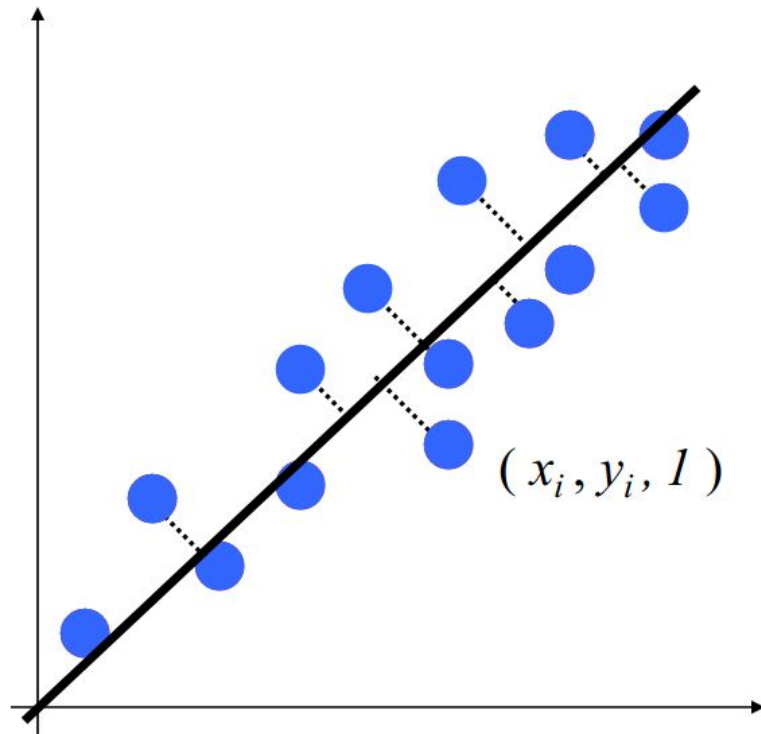




# Least square

Find a line to minimize the sum of squared distance to the points

$$E = \sum_{i=1}^n (ax_i + by_i + d)^2$$



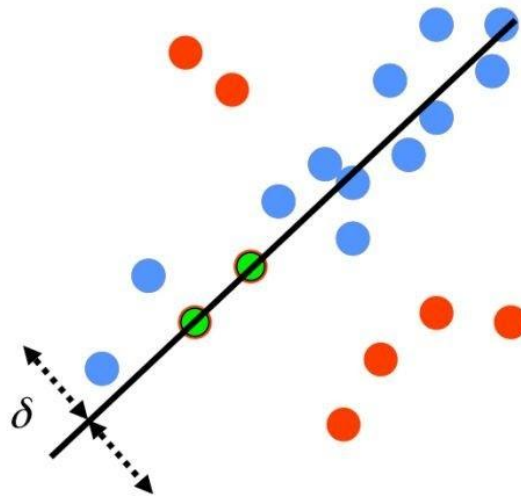
# RANSAC

***R**andom **s**ample **c**onsensus*

For fitting a model to noisy data!

Iterative approach:

- Sample a subset of points
- Fit our model
- Count the total # of inliers that match this model
- Repeat

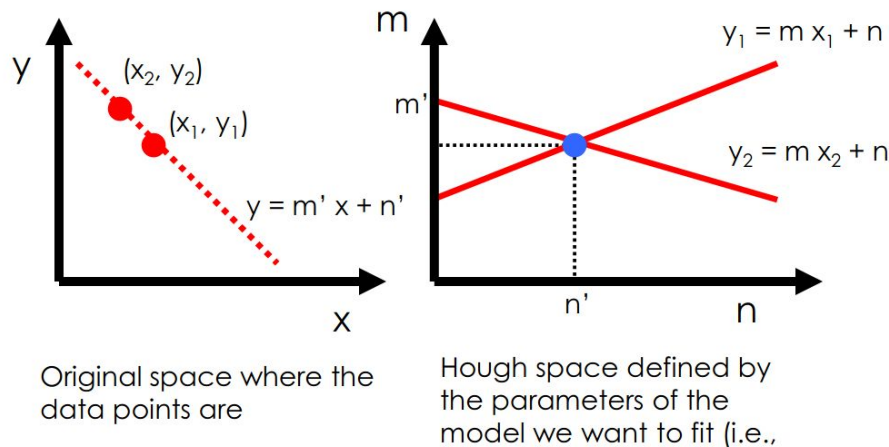


# Hough Transforms

Key idea for line fitting:

- Map points in  $(x,y)$  to a line in our Hough space
- Each point in our Hough space represents a line in our  $(x,y)$  space
- Intersection of lines in hough space = line

- + Polar line representation
- + Discretization and voting

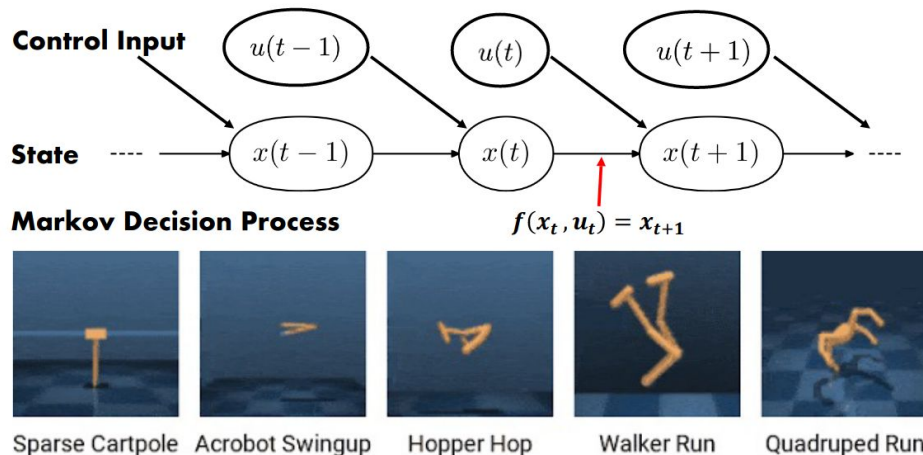


# Representation Learning

**State:** Quantity that describes the most important aspect of a dynamical system at time  $t$

**Representation:** data format of input or output including a low-dimensional representation of sensor data

- Compact
- Explanatory
- Disentangled
- Hierarchical
- Makes subsequent problem easier

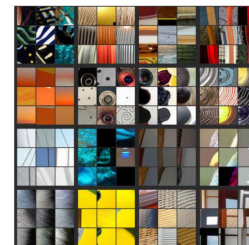
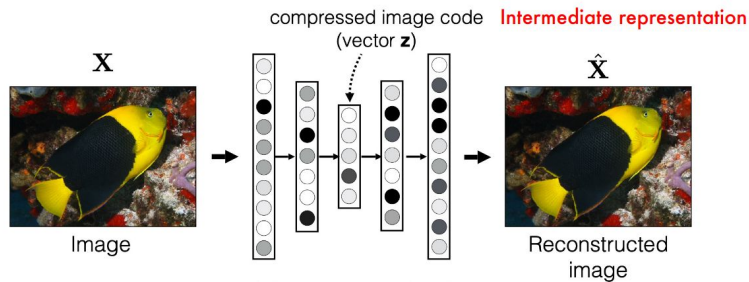


# Representation

## Learned representation

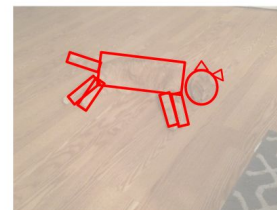
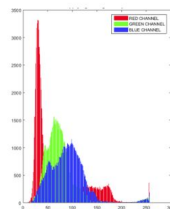
- Supervised
- Unsupervised
- Self-supervised

Visualizing learned representation



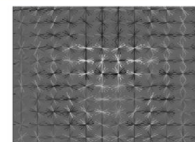
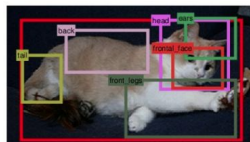
## Interpretable representation

Color Histograms



Model based Shapes

Deformable Part based Models (DPM)



Histogram of Gradients (HOG)