

# PS3 Review Session

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CS231A  
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# Overview

Space carving

Single Object Recognition via SIFT

Histogram of Oriented Gradients (HOG)

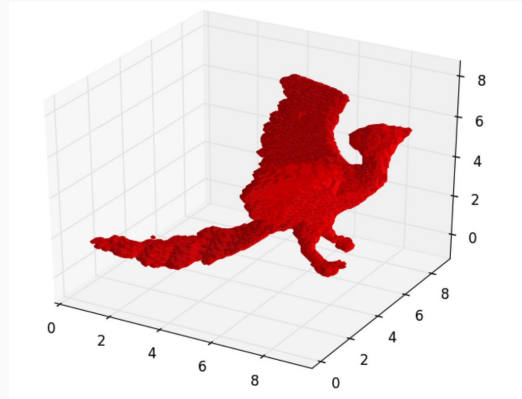
# Space Carving

Objective:

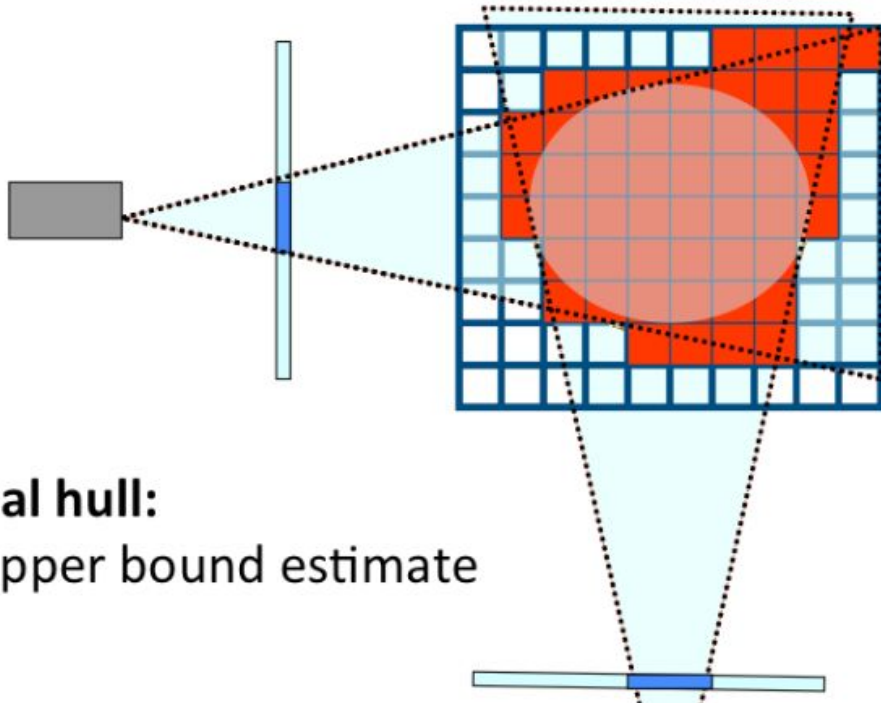
- Implement the process of space carving.

Lectures:

- Active Stereo & Volumetric Stereo

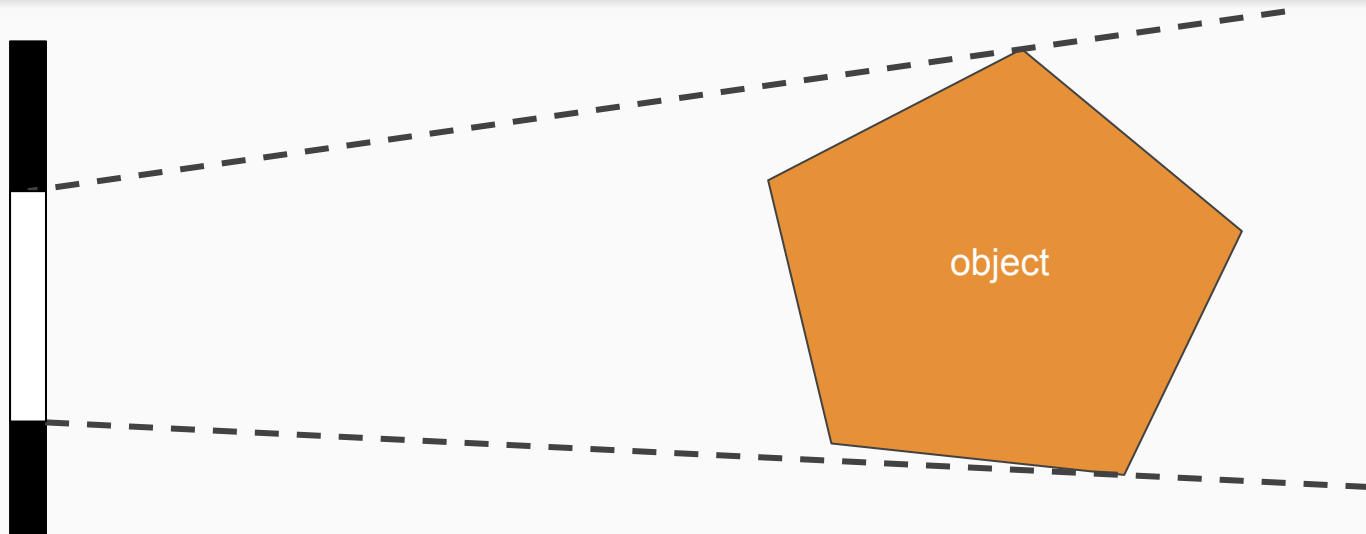


## Review: Space Carving

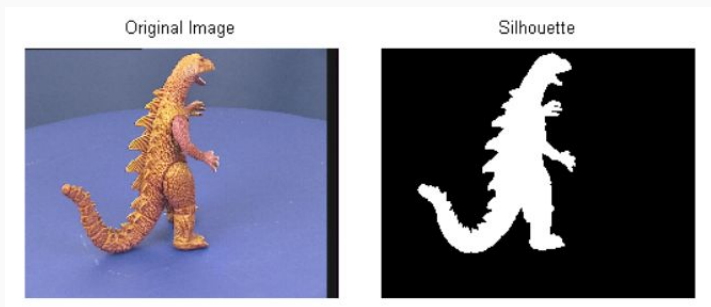


**Visual hull:**  
an upper bound estimate

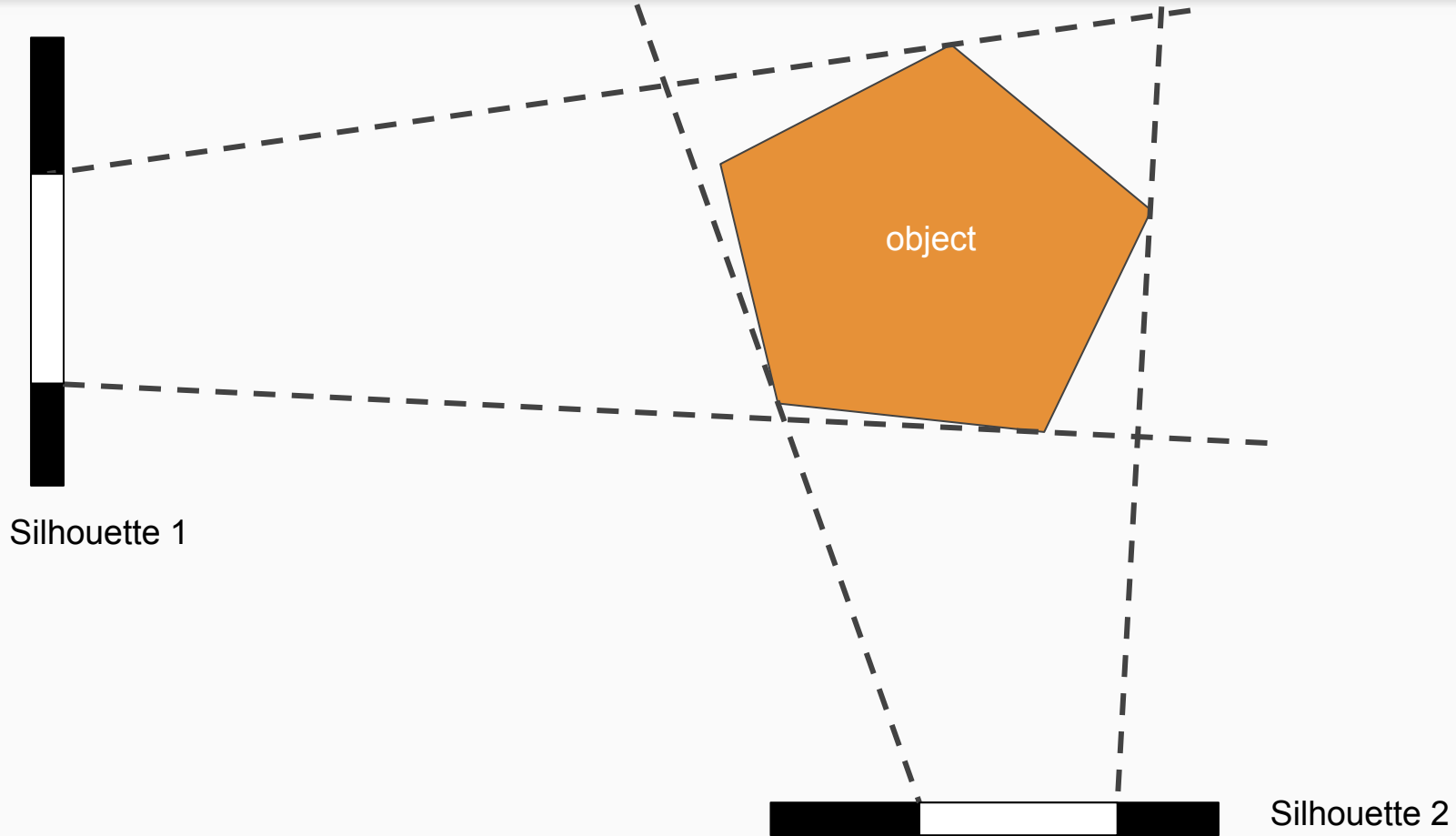
# Review: Space Carving



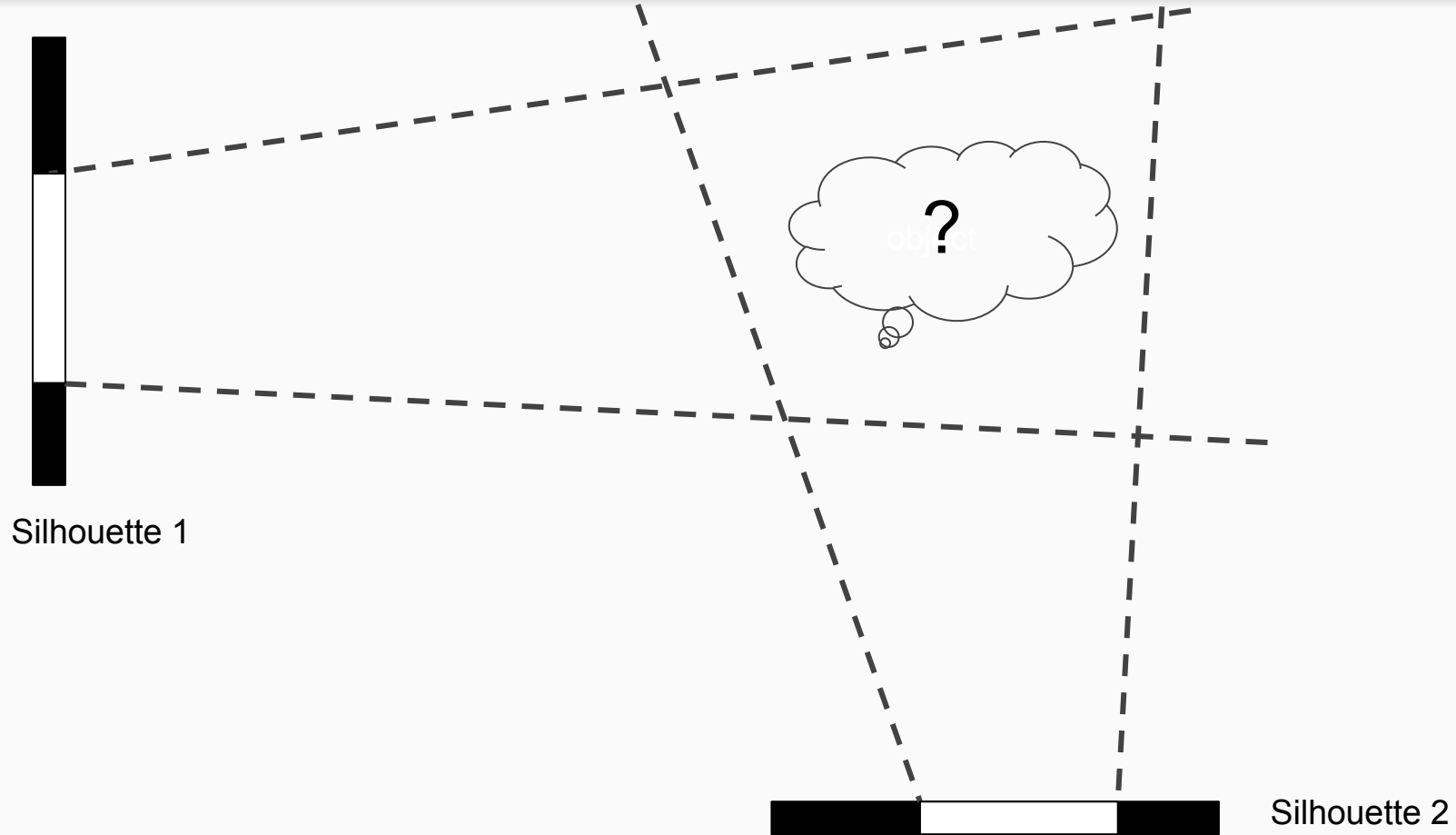
Silhouette 1



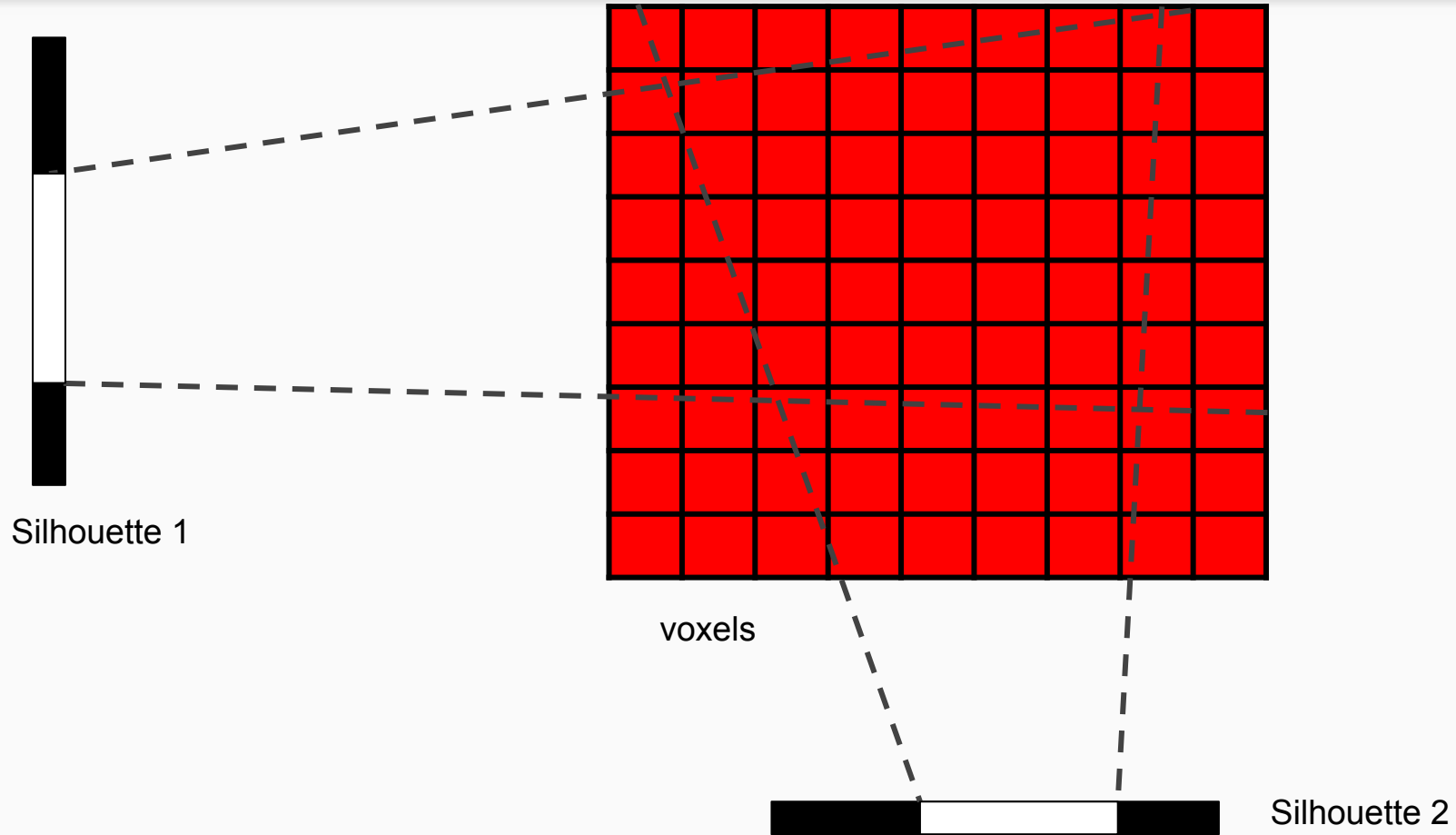
## Review: Space Carving



# Goal of Space Carving

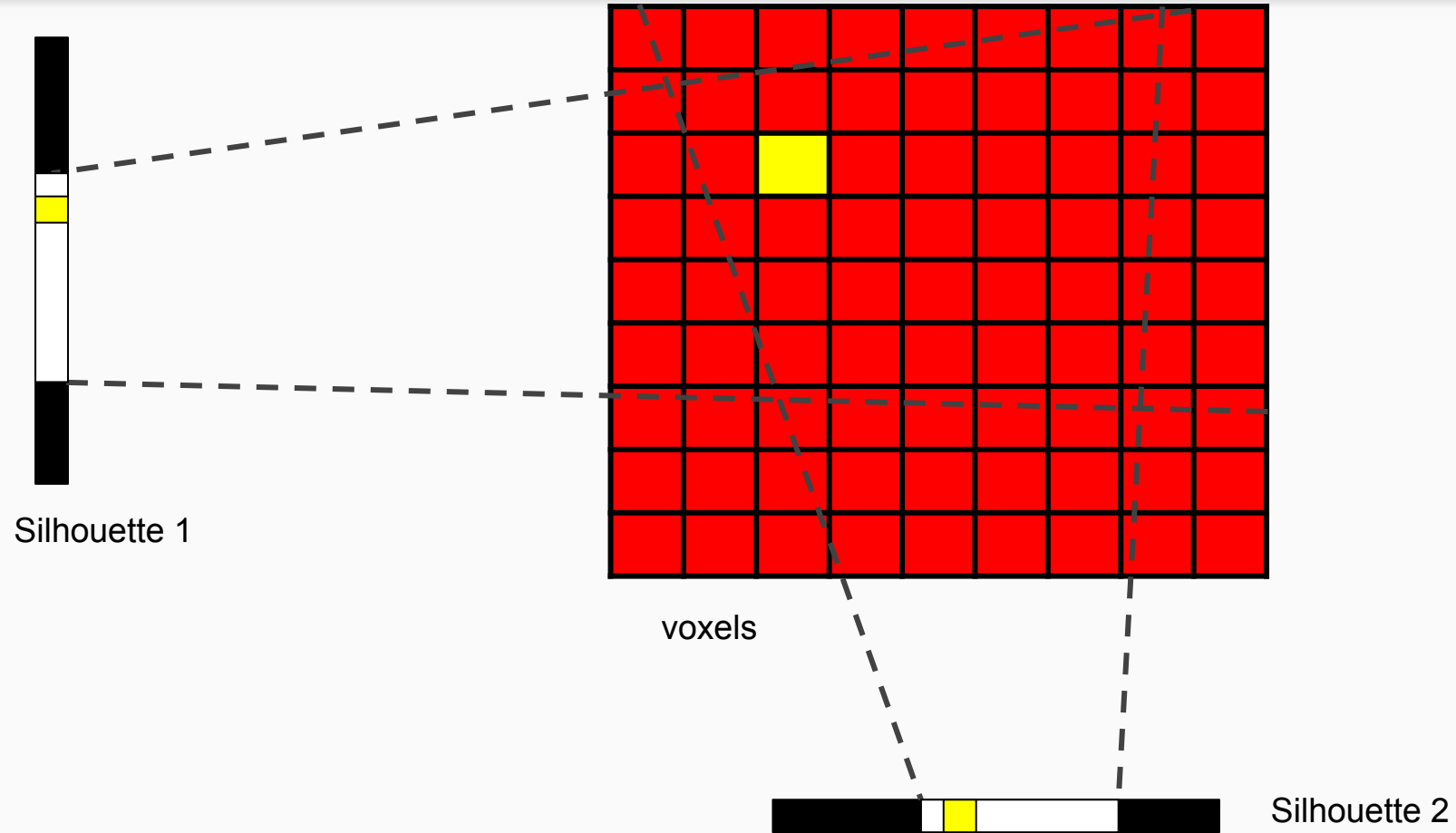


## Review: Space Carving

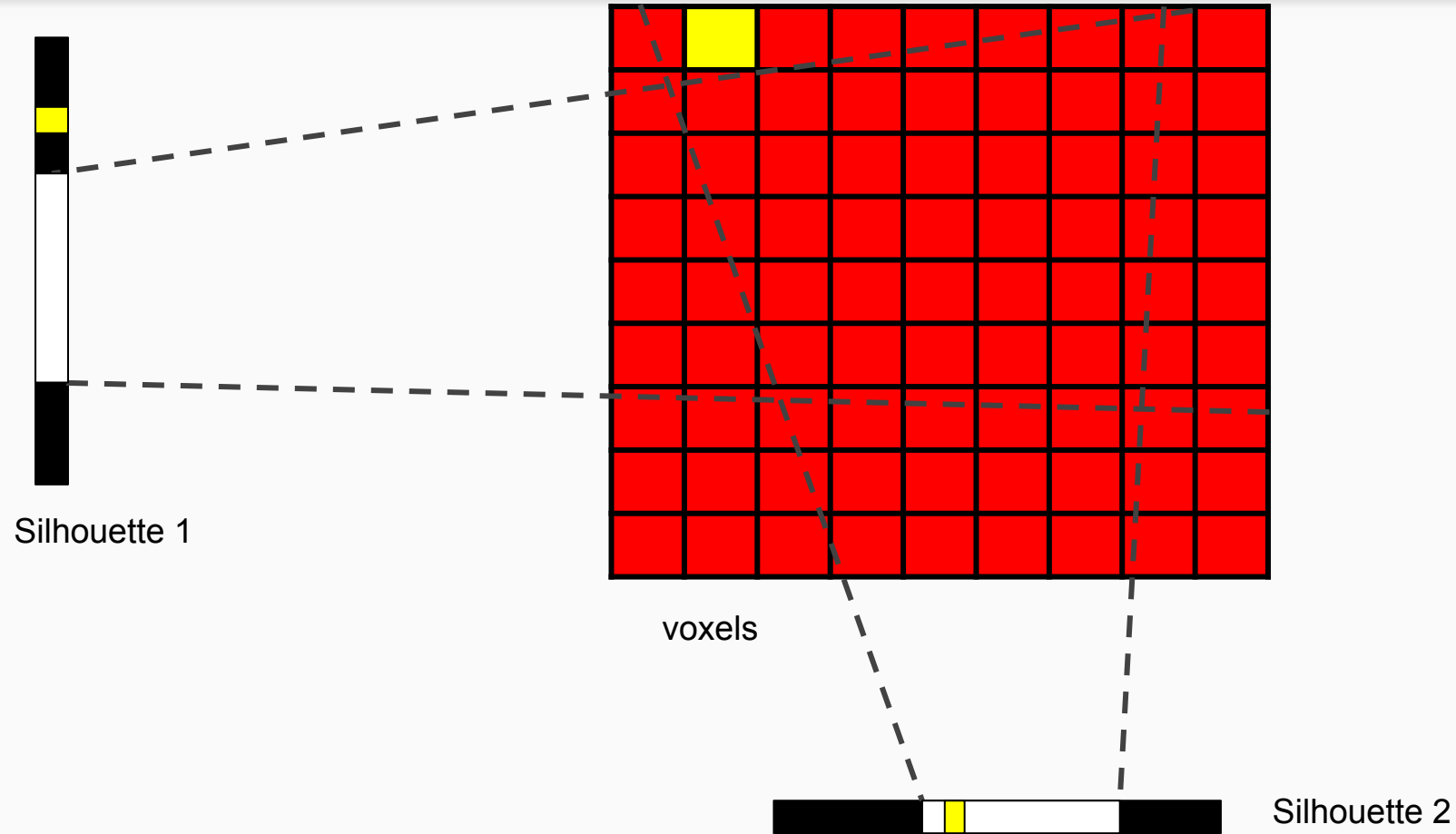




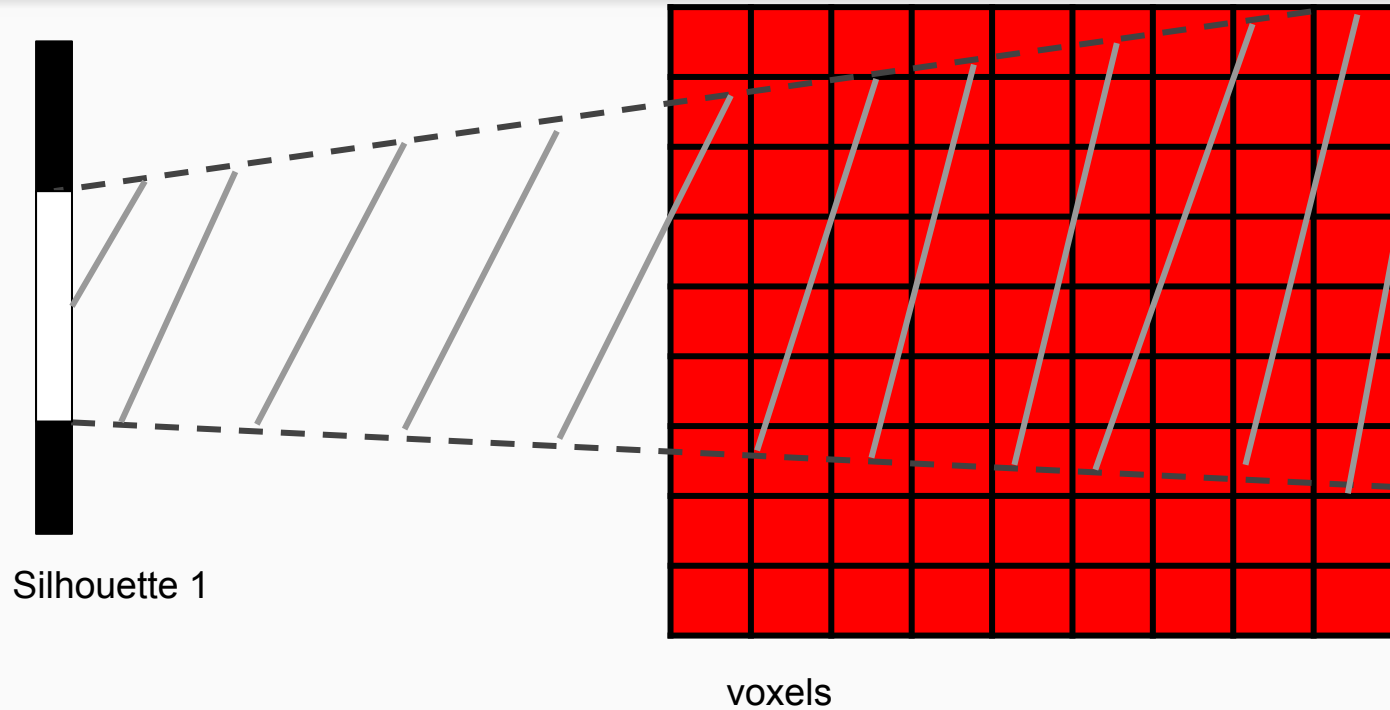
# Review: Space Carving



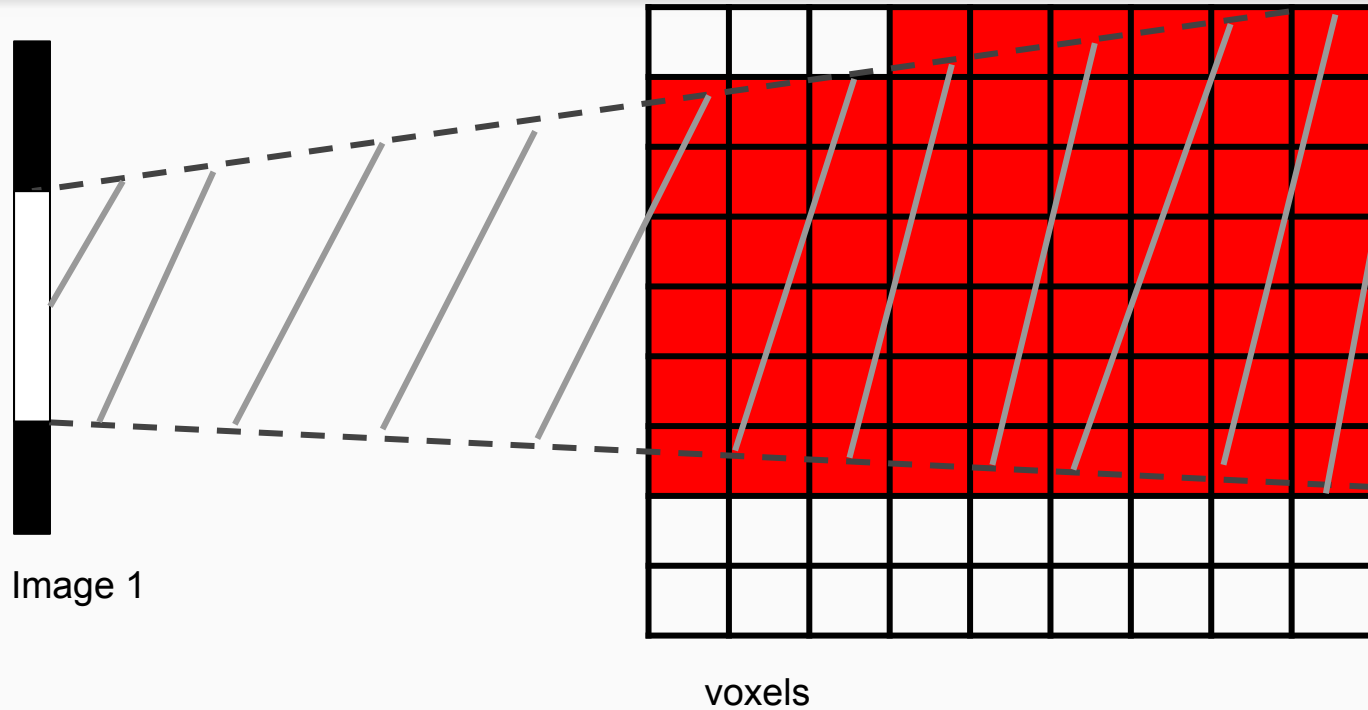
# Review: Space Carving



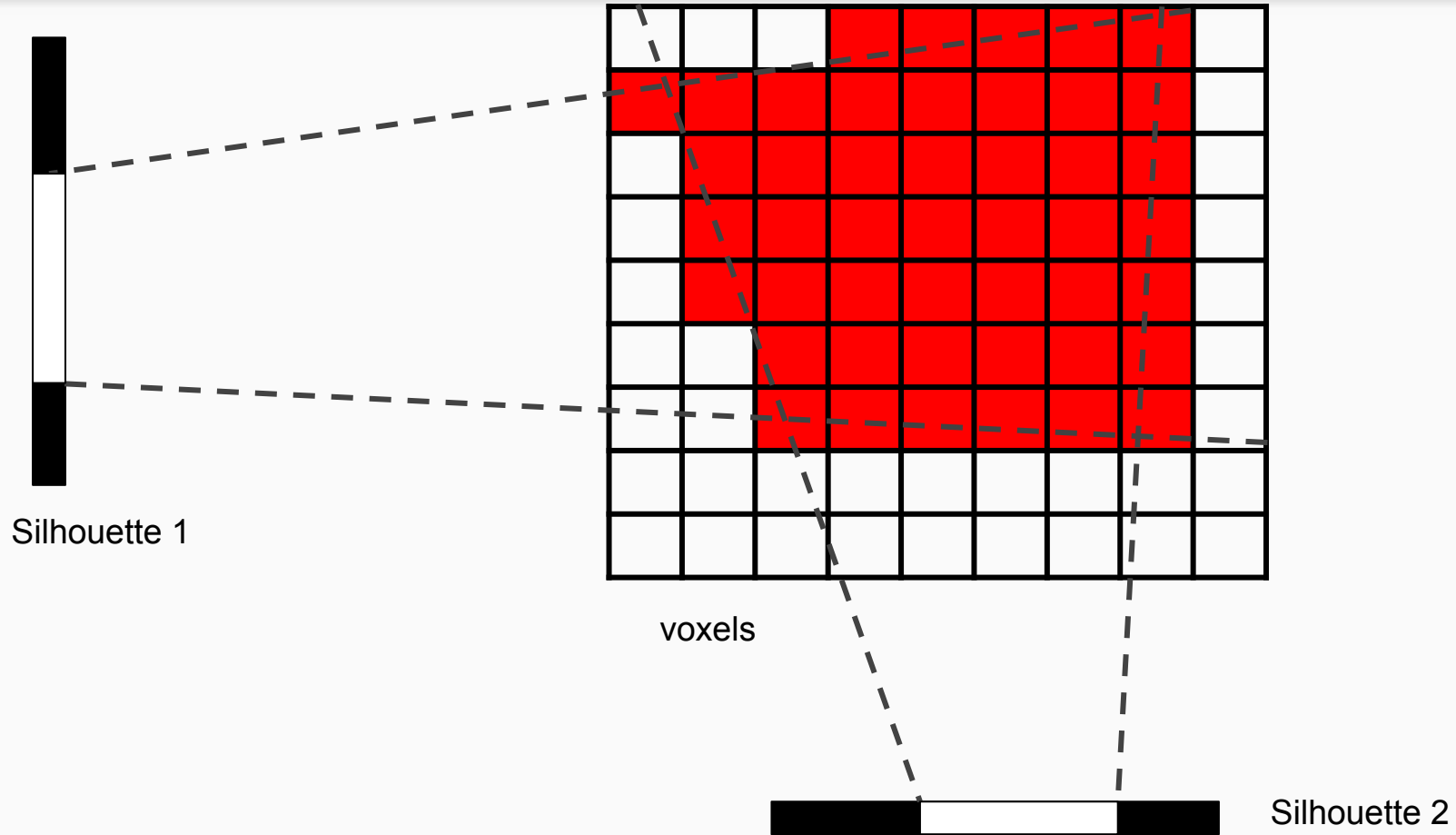
## Review: Space Carving



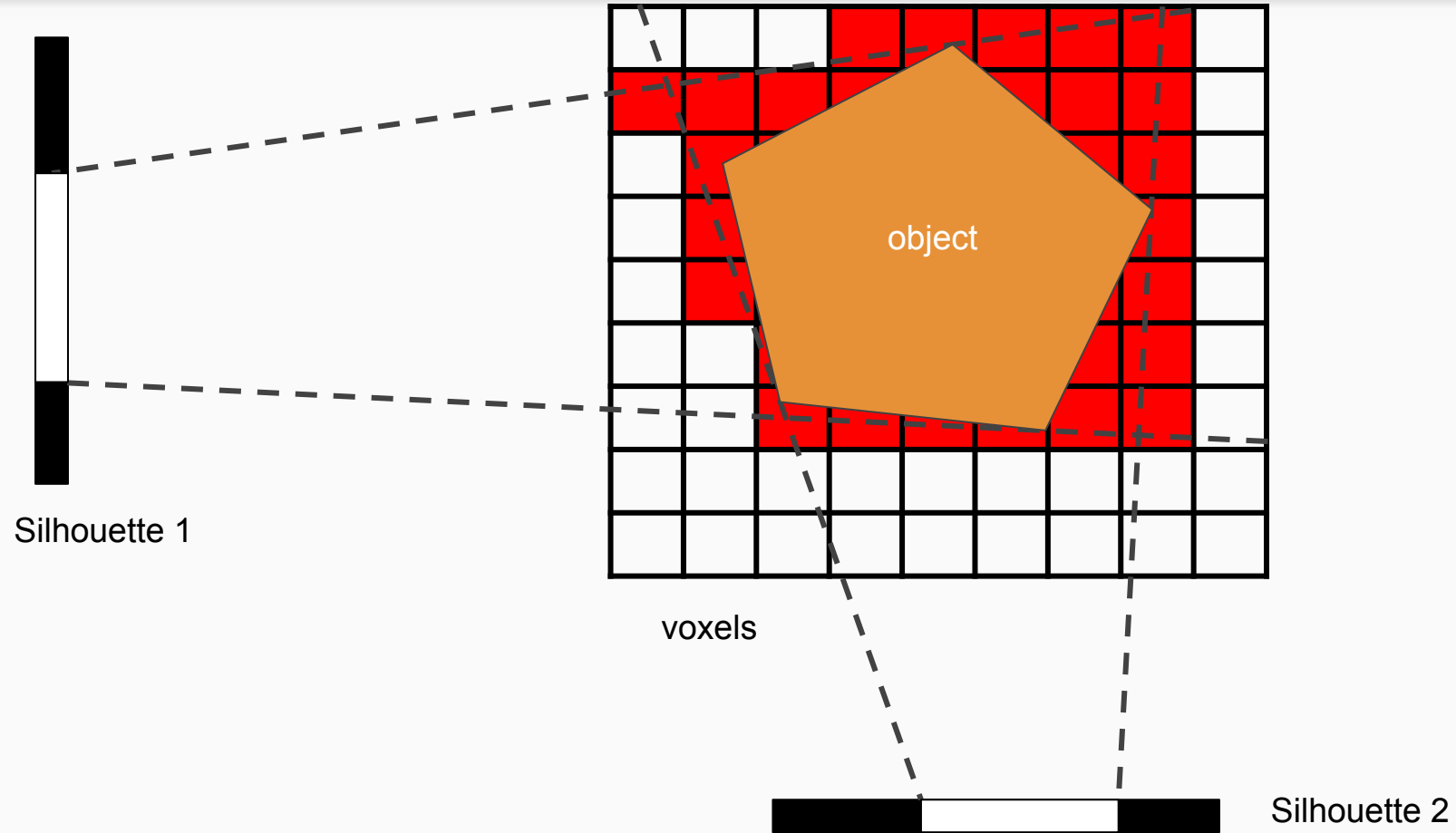
## Review: Space Carving



## Review: Space Carving



## Review: Space Carving



# Space carving - overview

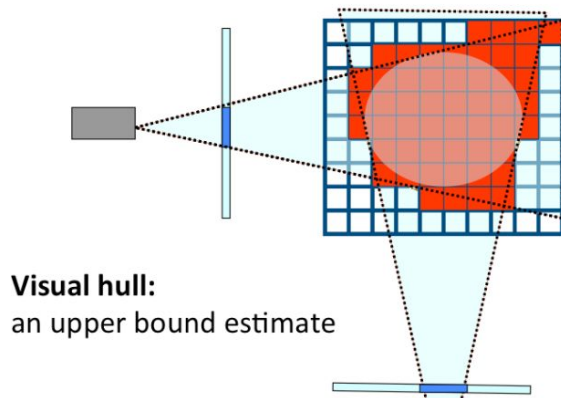
## Steps:

- Estimate silhouettes of images (could be based on some heuristics, e.g. color)
- Form the initial voxels as a cuboid
- Iterate over cameras and remove the voxels which project to the dark part of each silhouette

# Space carving - (a) (b) (c)

## Steps:

- Estimate silhouettes of images (could be based on some heuristics, e.g. color)
- Form the initial voxels as a cuboid
  - You may find these functions useful: `np.meshgrid`, `np.repeat`, `np.tile`
- Iterate over cameras and remove the voxels which project to the dark part of each silhouette
  - Question: What will the voxels look like after the first, second, ... iteration?

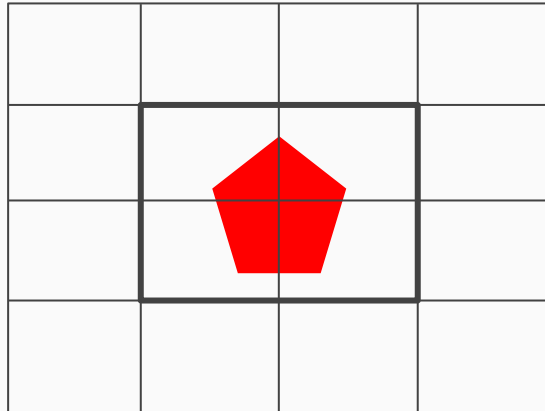




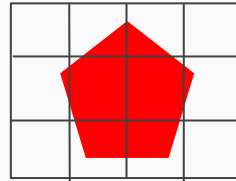
## Space carving - (d)

Steps:

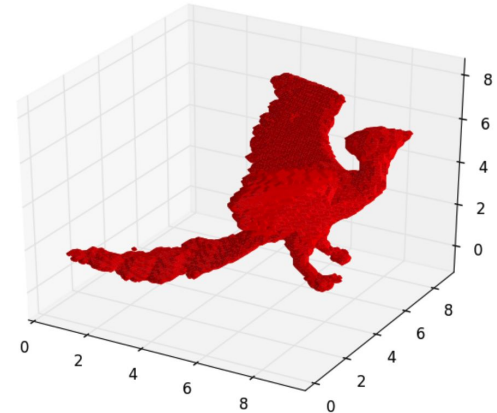
- Estimate silhouettes of images (could be based on some heuristics, e.g. color)
- Form the initial voxels as a cuboid
  - **Question: What will the cuboid look like after each iteration?**
- Iterate over cameras and remove the voxels which project to the dark part of each silhouette



Coarse  
Carving



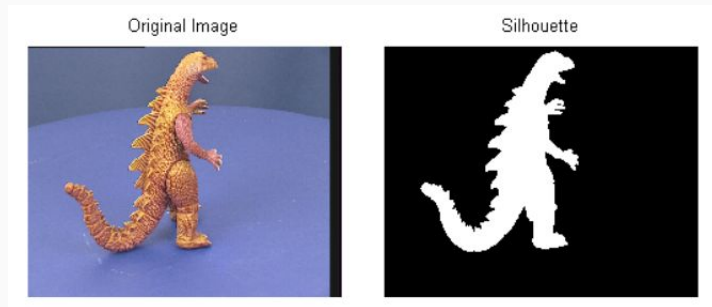
Final Output



## Space carving - (e)

### Steps:

- Estimate silhouettes of images (could be based on some heuristics, e.g. color)
  - Problem: The quality of silhouettes is not perfect.
  - The silhouette from each camera is not perfect, but the result is ok. Why?
  - Experiment: Use only a few of the silhouettes.
- Form the initial voxels as a cuboid
- Iterate over cameras and remove the voxels which project to the dark part of each silhouette



# Single Object Recognition Via SIFT - overview

## Objective:

- Understand how to use SIFT features for object recognition.
- Implement the RANSAC algorithm.
- Implement the Hough Transform algorithm.
- (Implementation of SIFT is not required.)

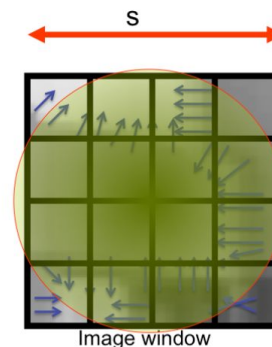
## Lectures:

- Fitting and Matching
- Detectors and Descriptors

## SIFT descriptor

David G. Lowe. "[Distinctive image features from scale-invariant keypoints.](#)" *IJCV* 60 (2), 04

- Alternative representation for image regions
- Location and characteristic scale  $s$  given by DoG detector



- 1 Compute gradient at each pixel
- 2  $N \times N$  spatial bins
- 3 Compute an histogram  $h_i$  of  $M$  orientations for each bin  $i$
- 4 Concatenate  $h_i$  for  $i=1$  to  $N^2$  to form a  $1 \times MN^2$  vector  $H$
- 5 Gaussian center-weighting
- 6 Normalize to unit norm

Typically  $M = 8$ ;  $N = 4$   
 $H = 1 \times 128$  descriptor

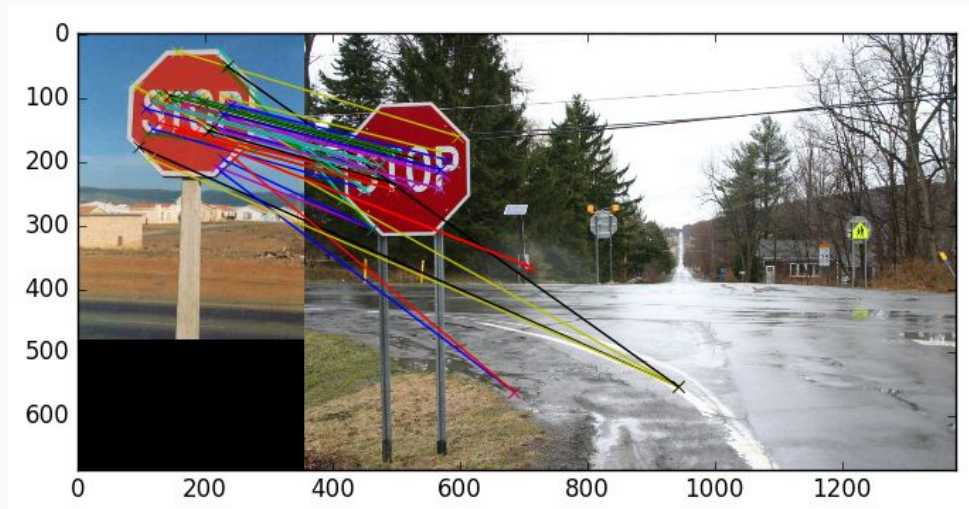
# Single Object Recognition Via SIFT - (a)

We've implemented SIFT descriptor for you, and your task is using it for object recognition.

Read: **Section 7. Application to object recognition** in Lowe's SIFT paper

<http://www.cs.ubc.ca/~lowe/papers/ijcv04.pdf>

Be sure to understand what does the threshold mean and how to use it.

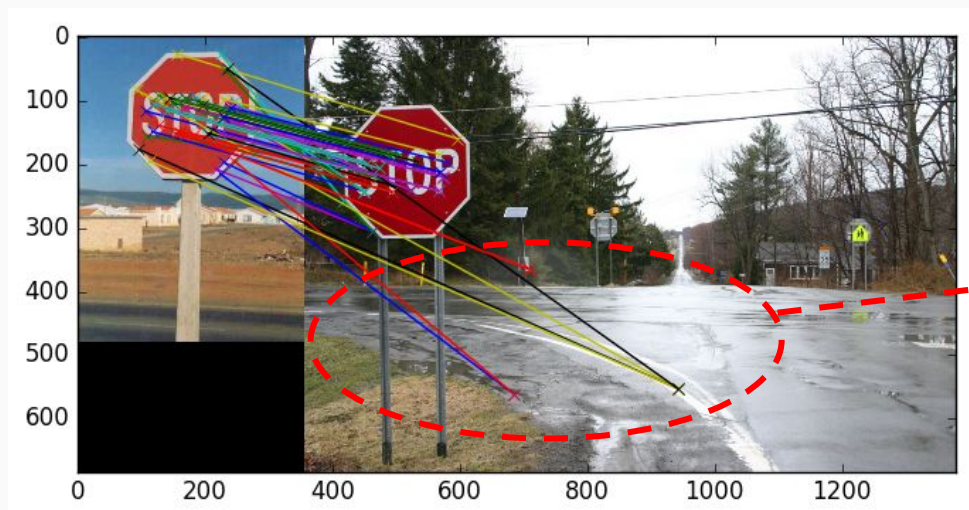


# Single Object Recognition Via SIFT - (b)

RANSAC to refine matching

Basic idea: use RANSAC to fit a homography matrix  $H$  between two set of key points. Only keep the inliers for matching, remove the outliers.

Question: how many pairs of correspondences do we need in each iteration?



# Single Object Recognition Via SIFT - (c)

Theoretical properties about RANSAC: see lecture notes / slides

Understand the relations between:

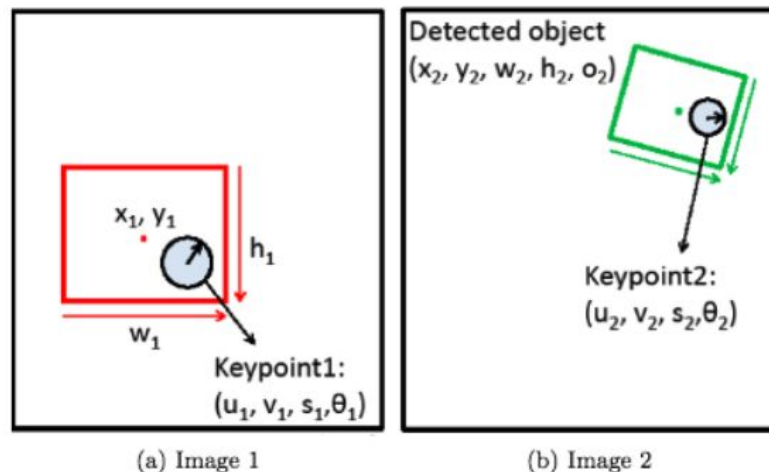
- $N$ : number of samples
- $e$ : outlier ratio
- $s$ : minimum number of data points to fit the model
- $p$ : probability guaranteed

# Single Object Recognition Via SIFT - (d) (e)

Find quantitative relation between two bounding boxes

Input ( $u_1, v_1, s_1, \theta_1$ ;  $u_2, v_2, s_2, \theta_2$ ;  $x_1, y_1, w_1, h_1$ )    Output ( $x_2, y_2, w_2, h_2, o_2$ )

Keypoint1                      Keypoint2                      Bbx1                      Bbx2 & Orientation



# Histogram of Oriented Gradients (HOG) - Overview

## Objective:

- Implement HOG features.

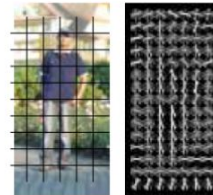
## Lectures:

- Detectors and Descriptors

## HoG = Histogram of Oriented Gradients

Navneet Dalal and Bill Triggs, Histograms of Oriented Gradients for Human Detection, CVPR05

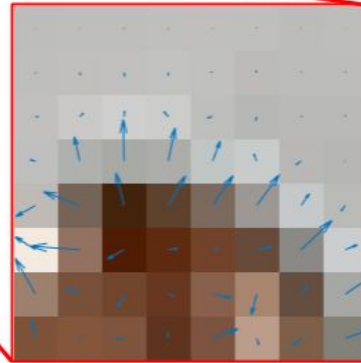
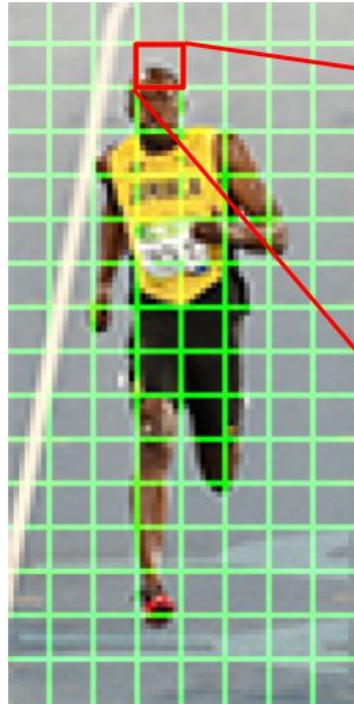
- Like SIFT, but...
  - Sampled on a dense, regular grid around the object
  - Gradients are contrast normalized in overlapping blocks





# Histogram of Oriented Gradients (HOG) - Online Reference

<https://www.learnopencv.com/histogram-of-oriented-gradients/>



2	3	4	4	3	4	2	2
5	11	17	13	7	9	3	4
11	21	23	27	22	17	4	6
23	99	165	135	85	32	26	2
91	155	133	136	144	152	57	28
98	196	76	38	26	60	170	51
165	60	60	27	77	85	43	136
71	13	34	23	108	27	48	110

**Gradient Magnitude**

80	36	5	10	0	64	90	73
37	9	9	179	78	27	169	166
87	136	173	39	102	163	152	176
76	13	1	168	159	22	125	143
120	70	14	150	145	144	145	143
58	86	119	98	100	101	133	113
30	65	157	75	78	165	145	124
11	170	91	4	110	17	133	110

**Gradient Direction**

# Histogram of Oriented Gradients (HOG) - Overview

## Steps:

- Preprocessing the image.
- Calculate the gradient image.
- Calculate the gradient histogram of each cell.
- Normalize each block.
- Calculate the HOG feature vector.

# Histogram of Oriented Gradients (HOG) - (a)

Steps:

- Preprocessing the image.
- Calculate the gradient image.
- Calculate the gradient histogram of each cell.
- Normalize each block.
- Calculate the HOG feature vector.

# Histogram of Oriented Gradients (HOG) - (a)

Compute the gradients of image (angles, magnitudes)

The way that the angles and magnitude per pixel are computed as follows:  
Given the following pixel grid

```
P1 P2 P3
P4 P5 P6
P7 P8 P9
```

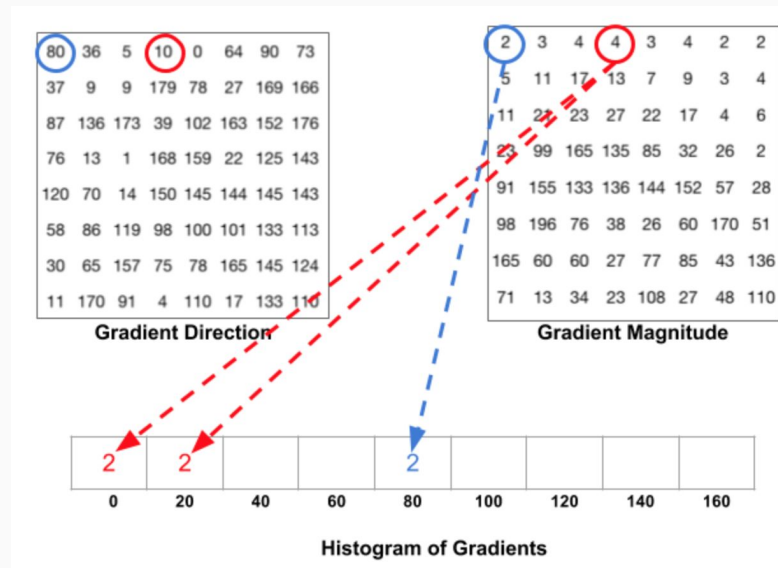
We compute the angle on P5 as  $\arctan(dy/dx) = \arctan(P2-P8 / P4-P6)$ .

The magnitude is simply  $\sqrt{(P4-P6)^2 + (P2-P8)^2}$

# Histogram of Oriented Gradients (HOG) - (b)

## Steps:

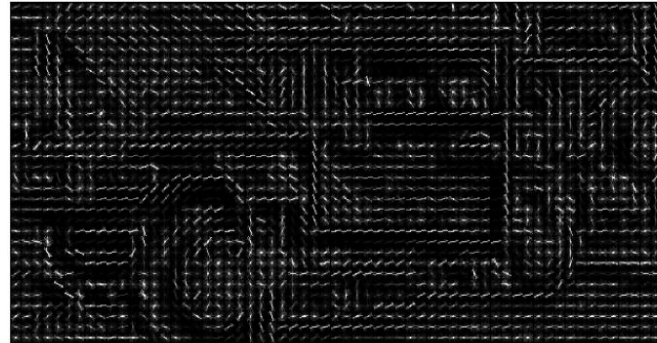
- Preprocessing the image.
- Calculate the gradient image.
- Calculate the gradient histogram of each cell.
  - The output shape is (nbins,).
- Normalize each block.
- Calculate the HOG feature vector.



# Histogram of Oriented Gradients (HOG) - (c)

## Steps:

- Preprocessing the image.
- Calculate the gradient image.
- Calculate the gradient histogram of each cell.
- **Normalize each block.**
  - We use a stride of 50% of the block size.
- **Calculate the HOG feature vector.**



# Thank You

