



CS 556: Computer Vision

Lecture 2

Prof. Sinisa Todorovic

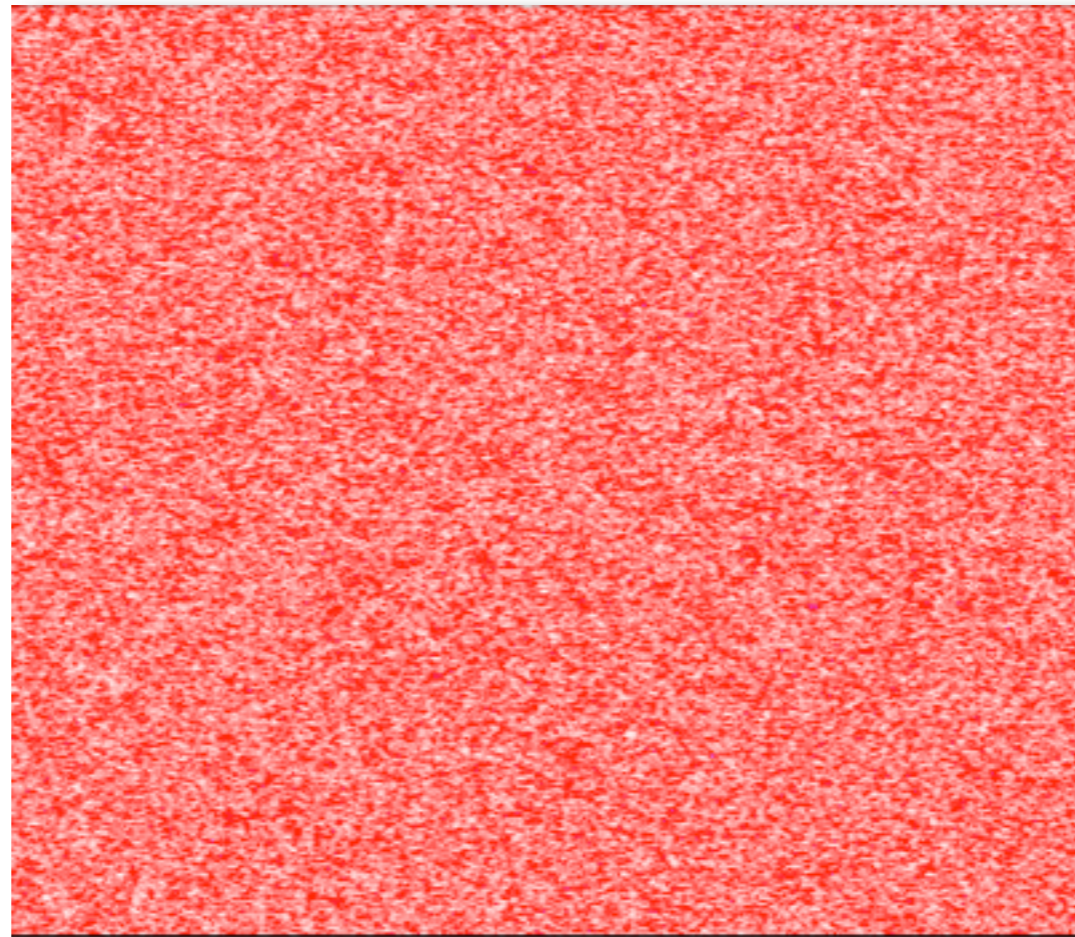
sinisa@eecs.oregonstate.edu

OSU Oregon State University

Outline

- Properties of low-level image structure
- From image formation to image interpretation
- Computational paradigms of computer vision

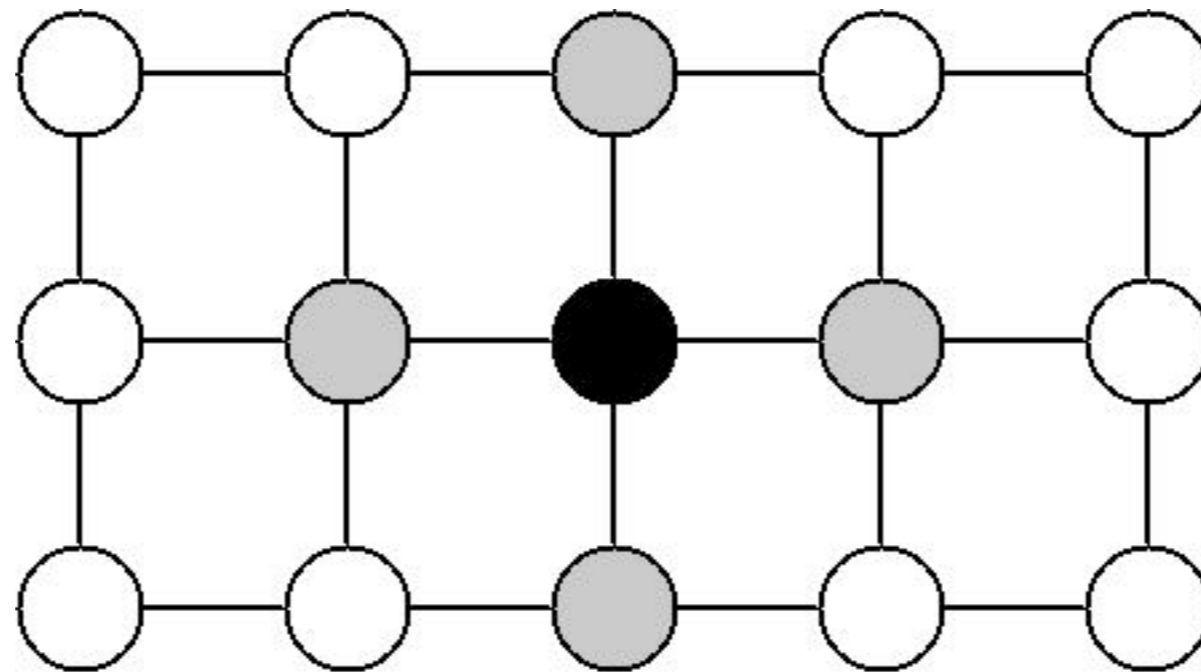
Images are not Collections of Random Pixels



Images are characterized
by structure

Image Structure

- Pixels, 4-adjacency, 8-adjacency, m-adjacency



4-adjacency

Image Structure

- Path -- directed, undirected, loop

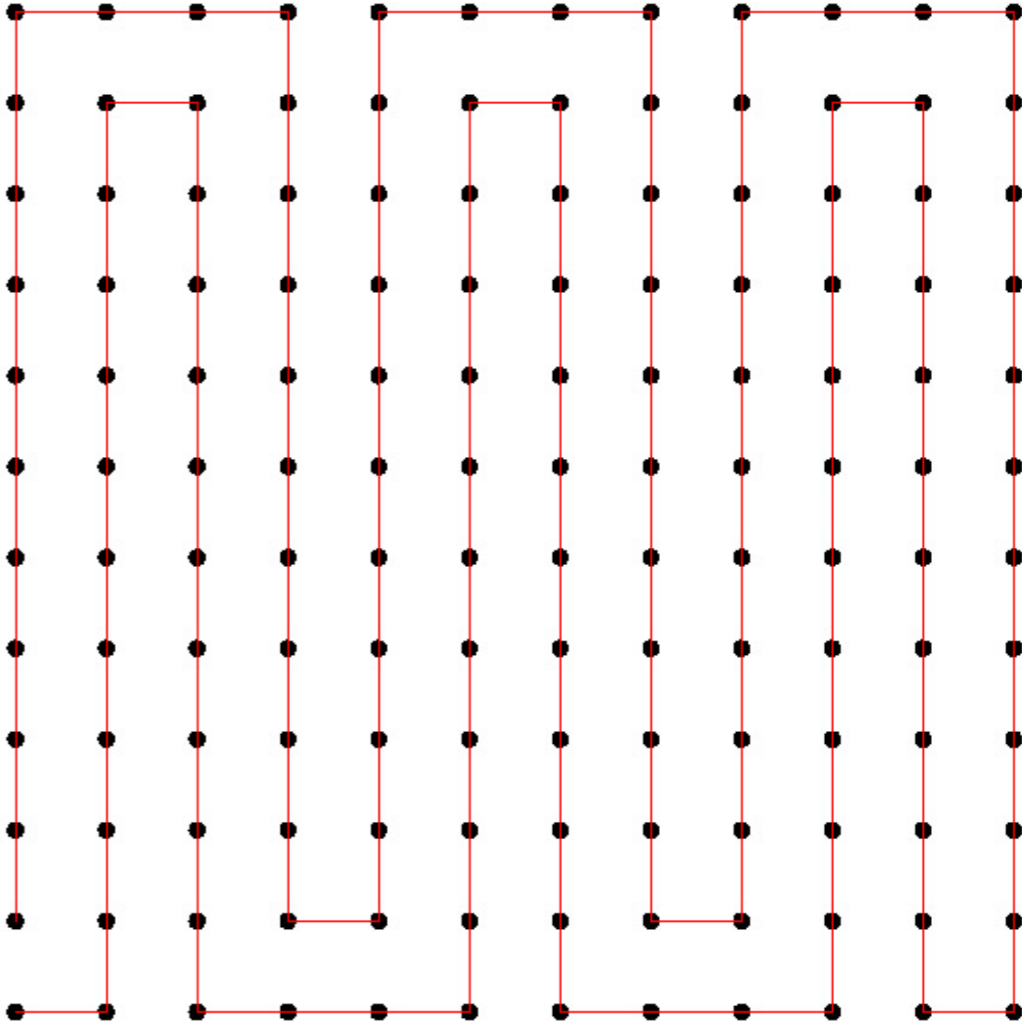
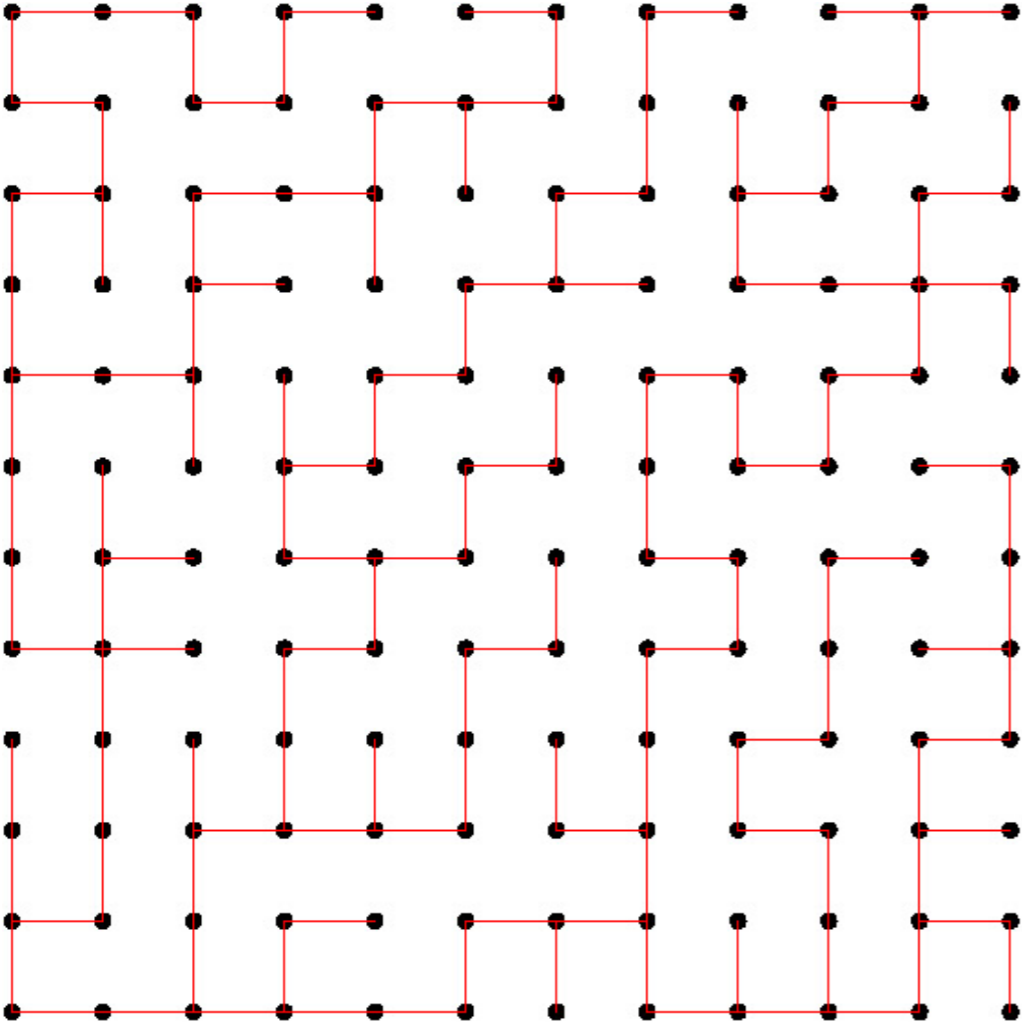


Image Structure

- Edge = Connected pixels with high gradient values



Image Structure

- T-junctions



Image Structure

- T-junctions

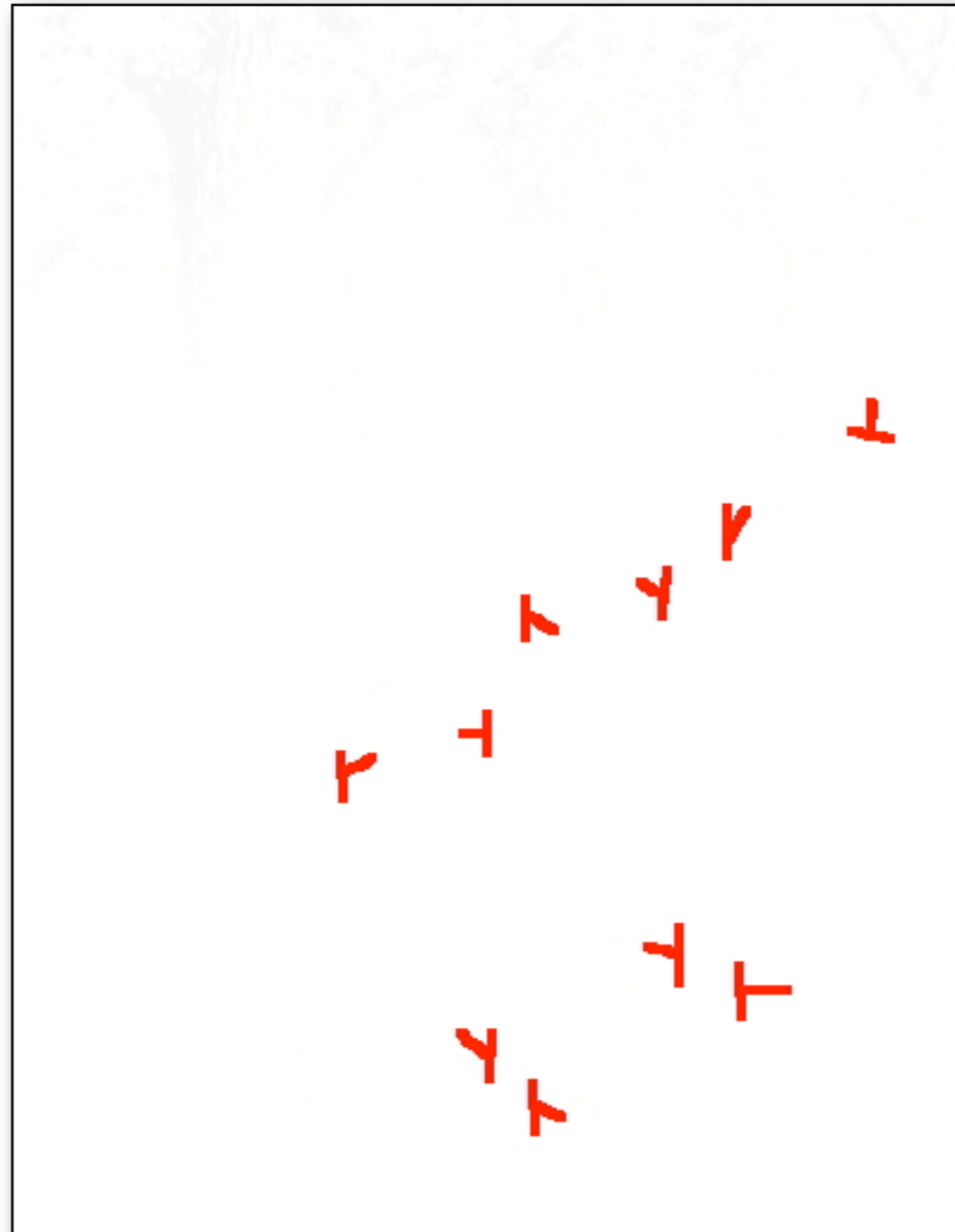


Image Structure

- T-junctions

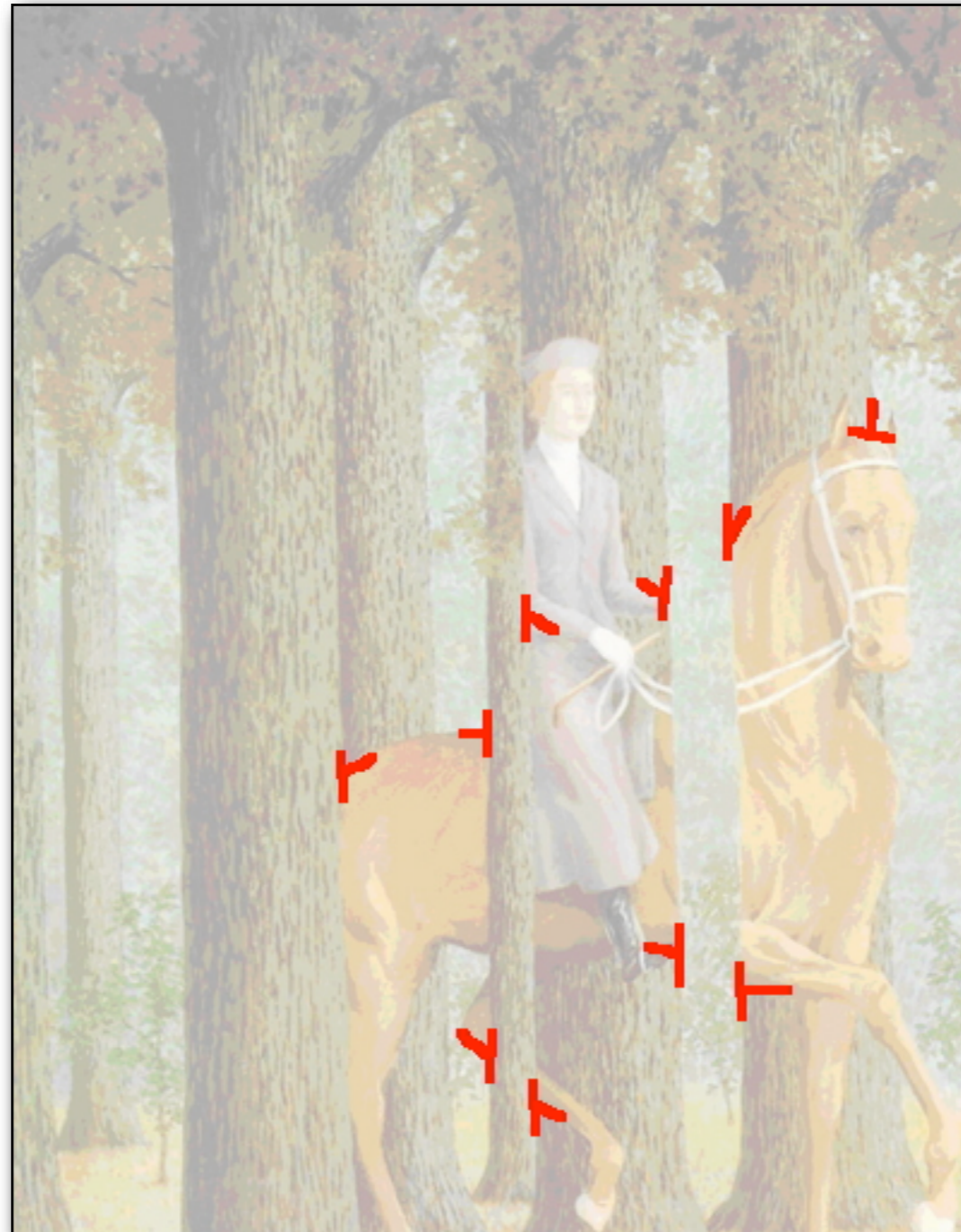


Image Structure

- Interest points = corners, textured patches

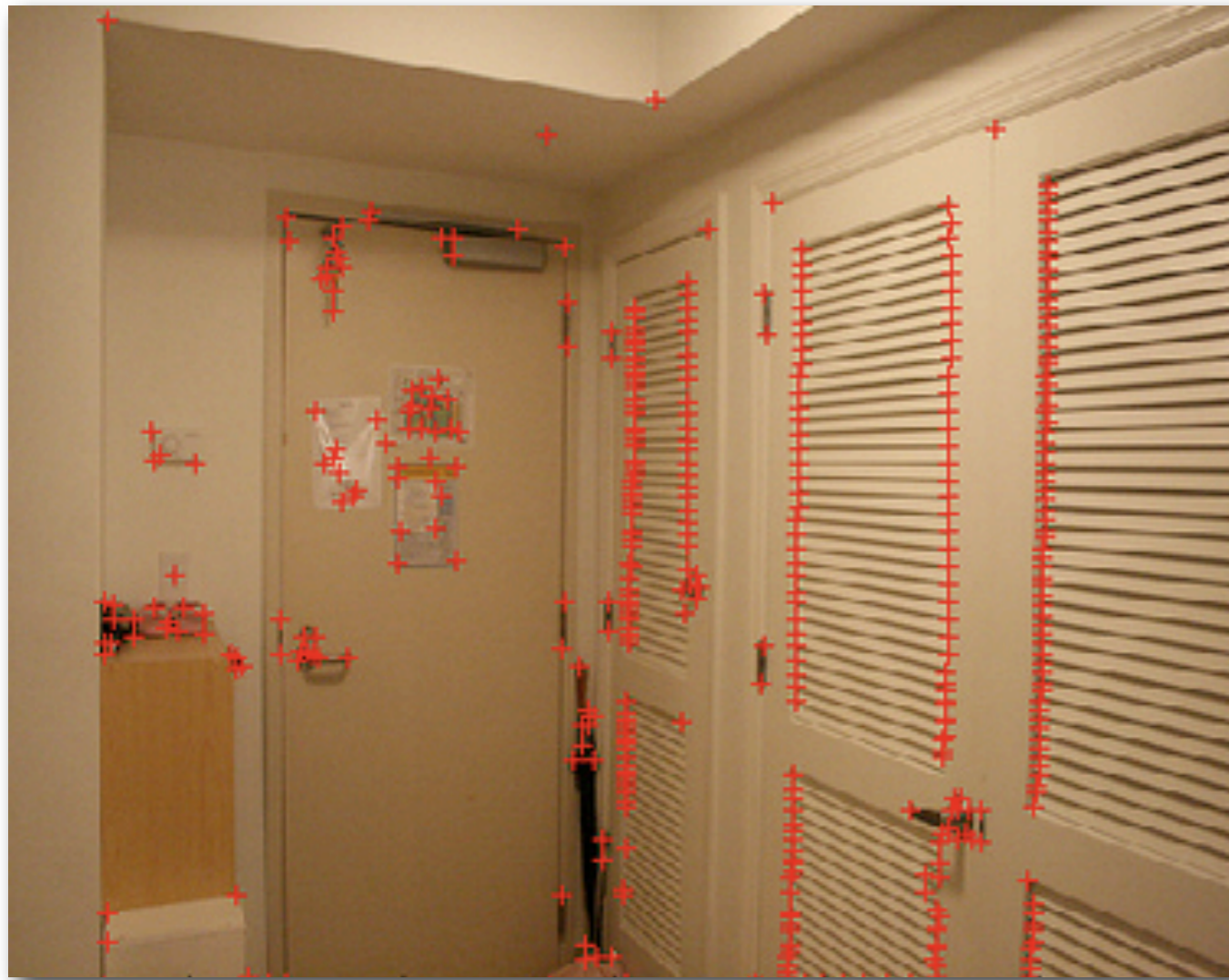


Image Structure

- Specularity = Highlights



Image Structure

- Lambertian surface = isotropic reflectance
- Specular surface = zero reflectance except at an angle



Image Structure

- Region = Connected set of pixels
- Region boundary, inner and outer contour

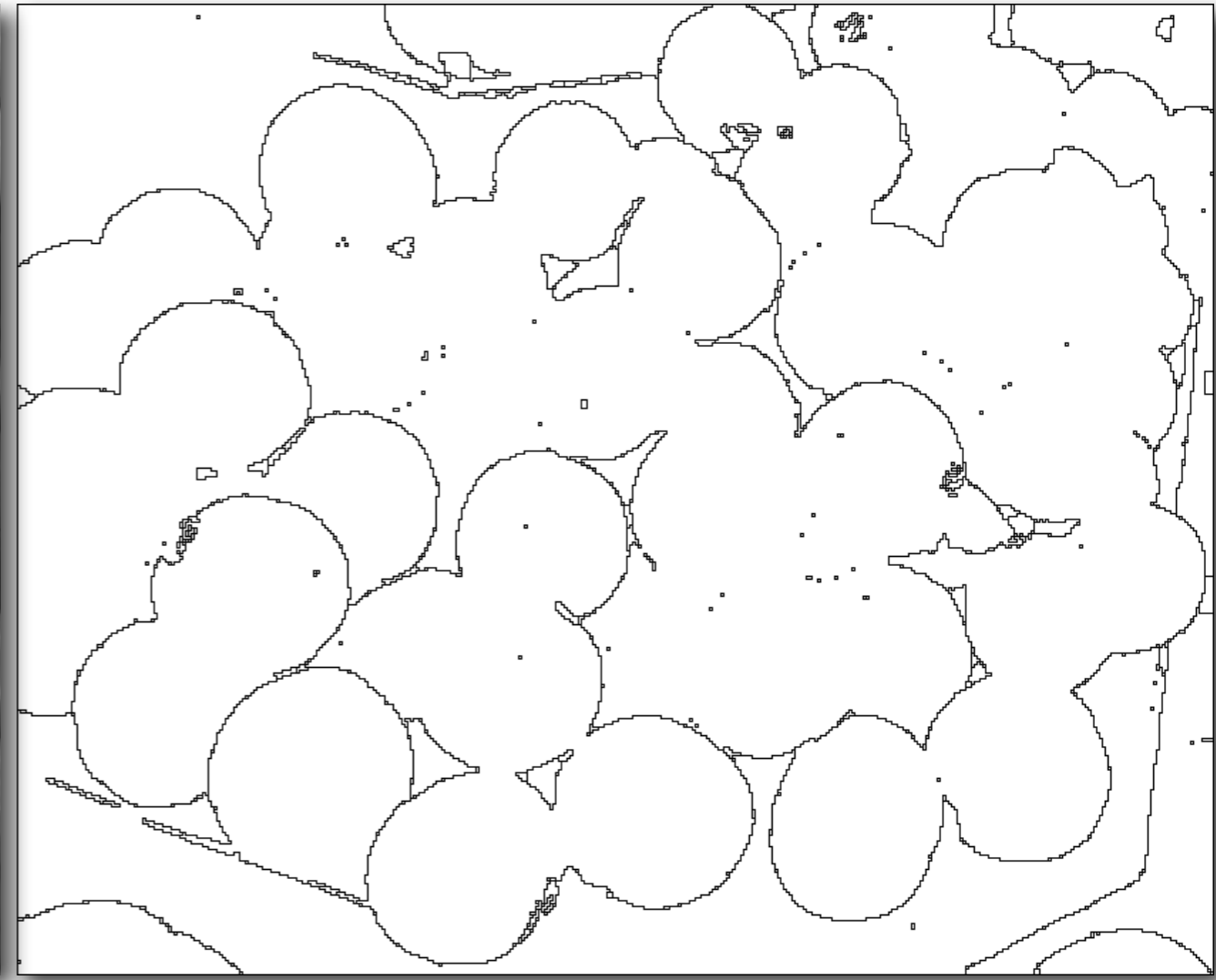


Image Structure

- Texture



Image Structure

- Dynamic texture



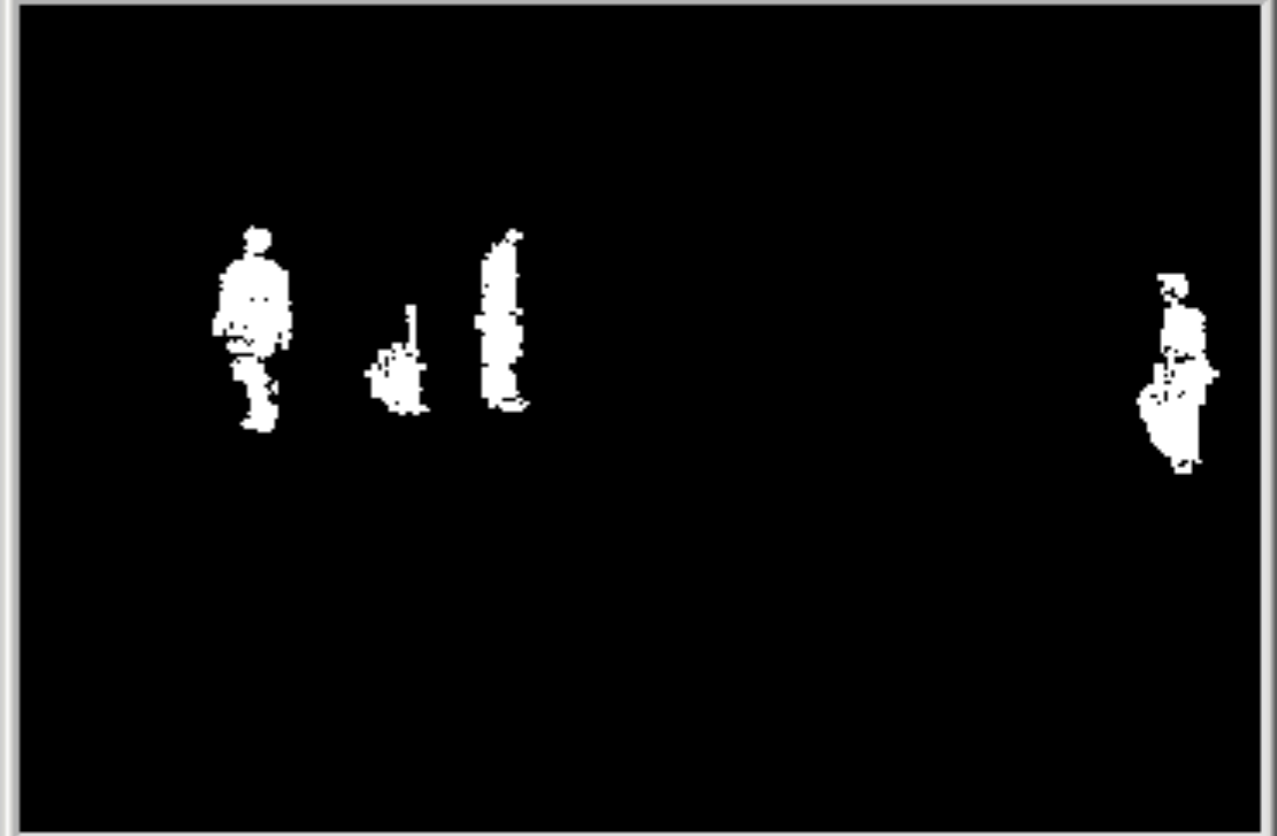
Image Structure

- Foreground - Background

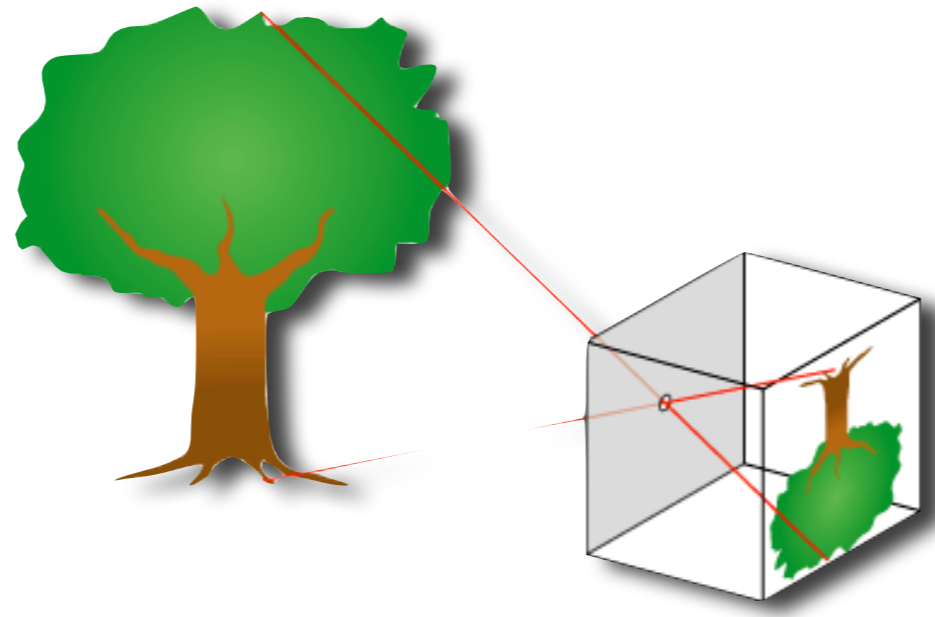
Input Image



Binary Detection



From Image Formation to Image Interpretation



From Image Formation to Image Interpretation

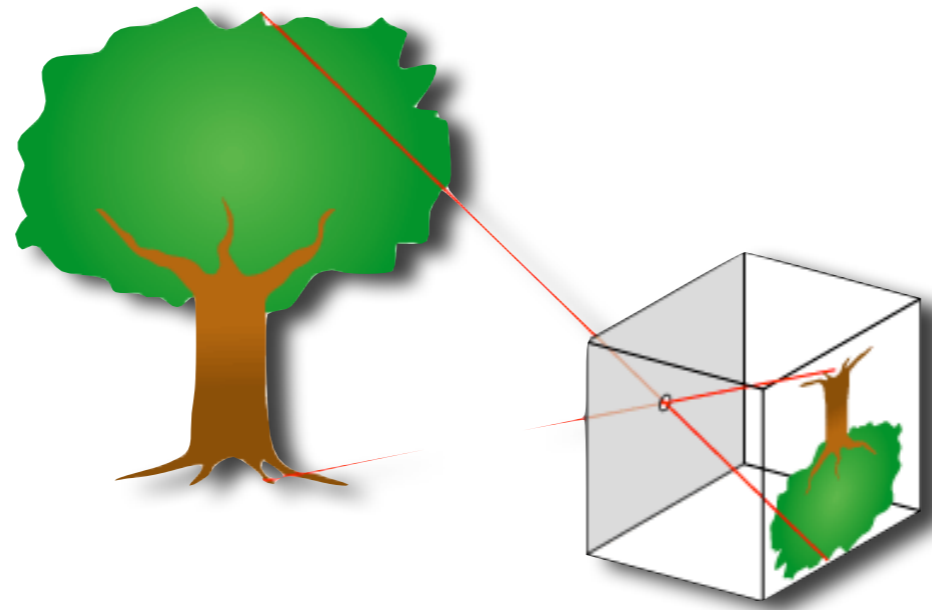


Image interpretation is usually defined in terms of

- Objects, scenes, activities, events
- Spatiotemporal relations between objects, scenes, ...

From Image Formation to Image Interpretation

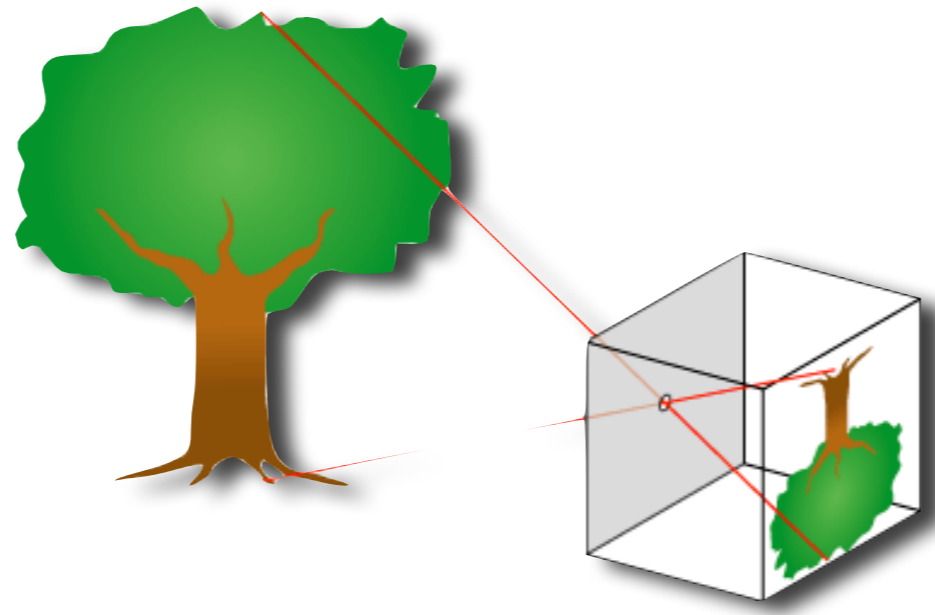
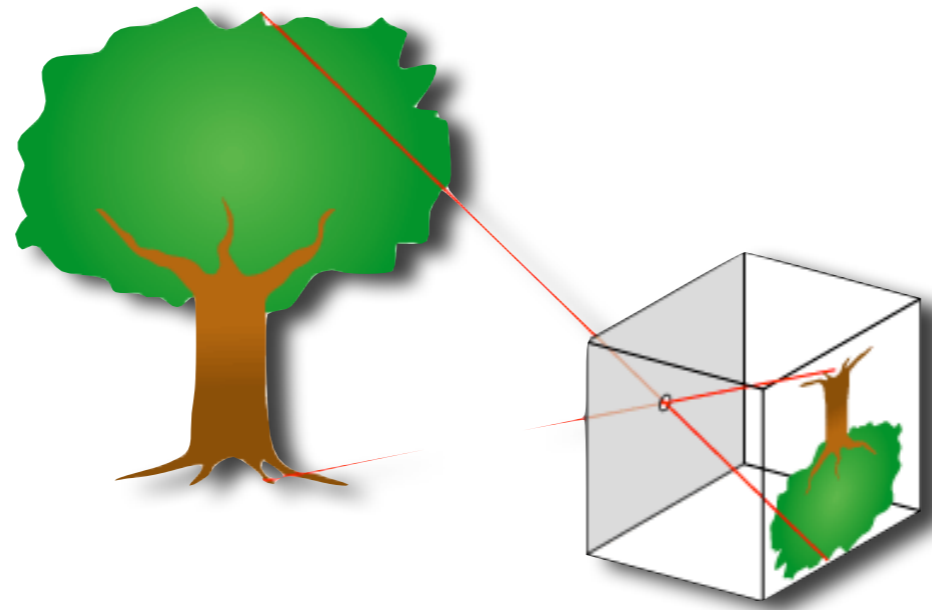


Image properties depend on:

- Imaging conditions
- Visual properties of the 3D world

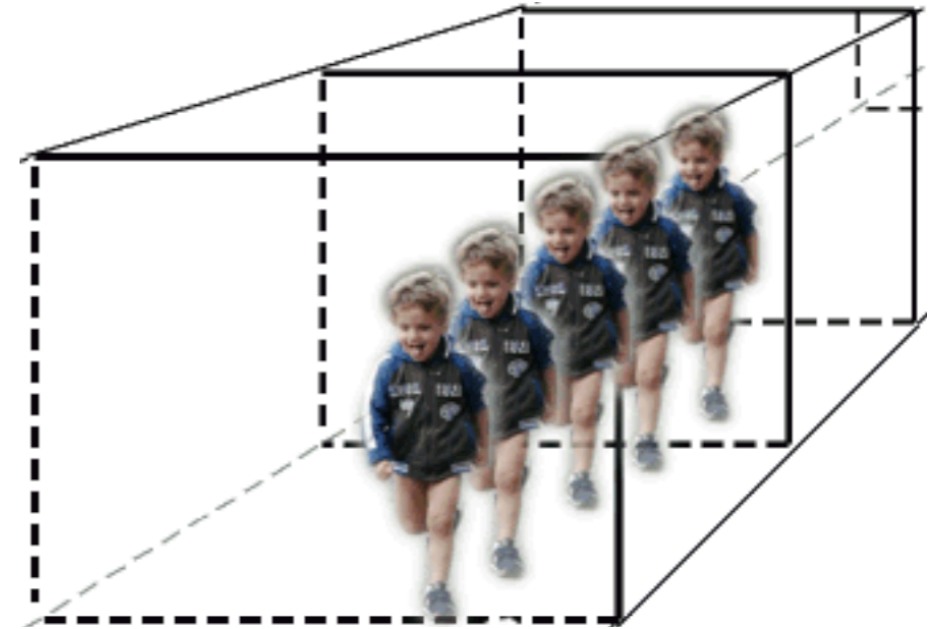
From Image Formation to Image Interpretation



- Image acquisition parameters:
 - Camera distance, viewpoint, motion
 - Camera intrinsic parameters (e.g., lens)
 - Illumination or Brightness
 - Occlusion and Clutter

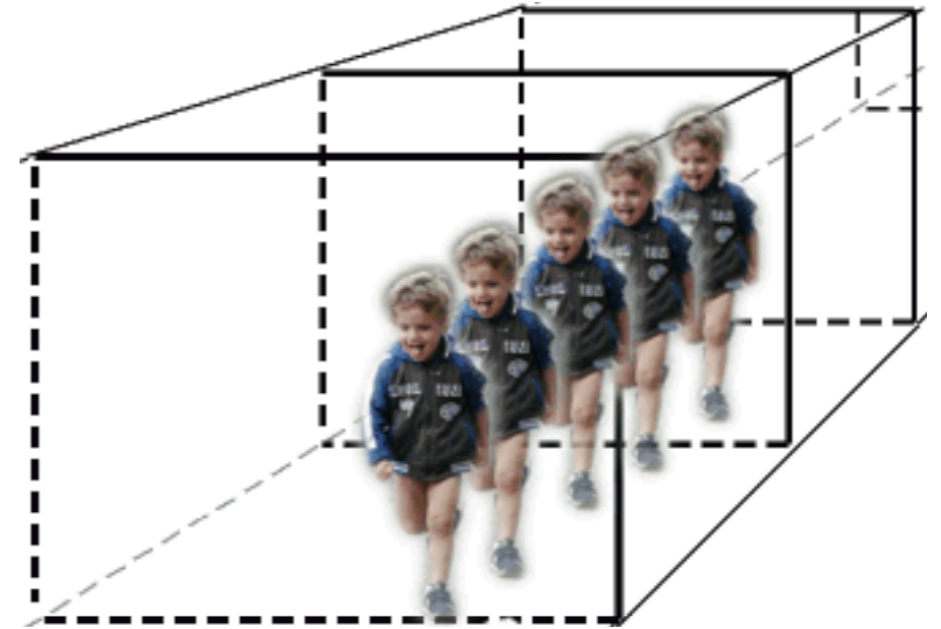
Are there general constraints
the real world poses
on image properties?

Properties of Real World Objects



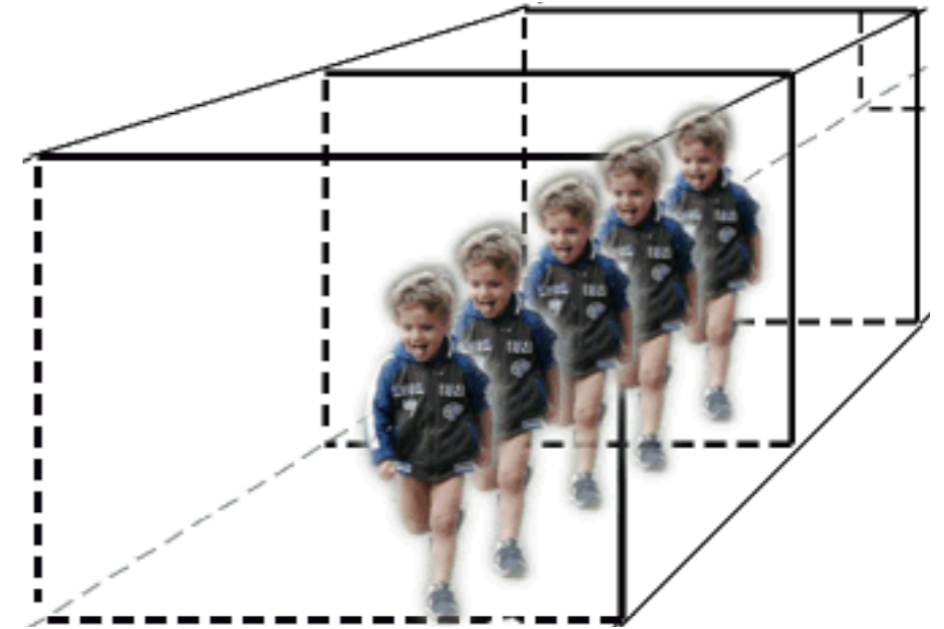
Properties of Real World Objects

- Occupy finite 3D volume



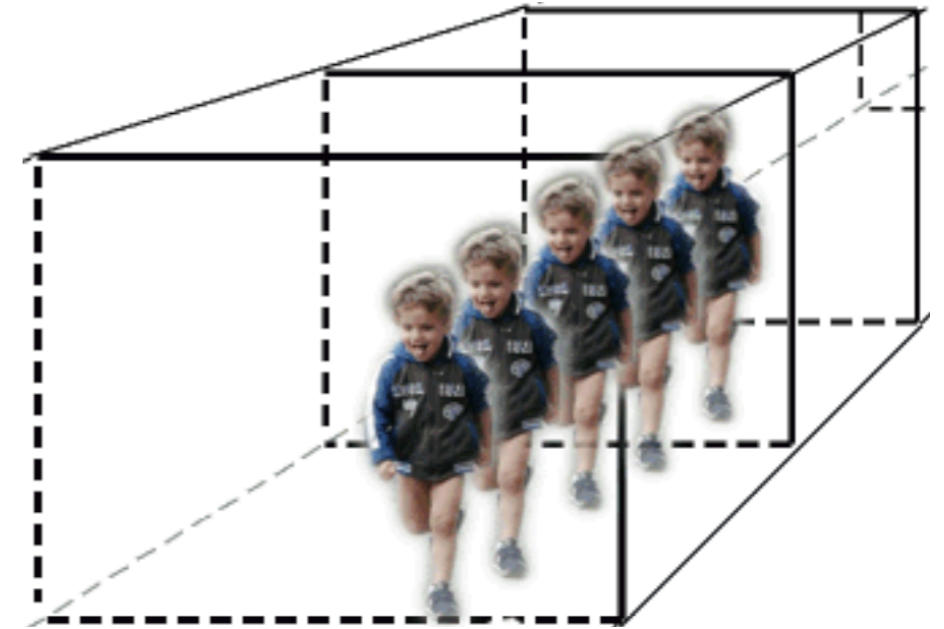
Properties of Real World Objects

- Occupy finite 3D volume
- Cohesive -- Contiguous & Continuous



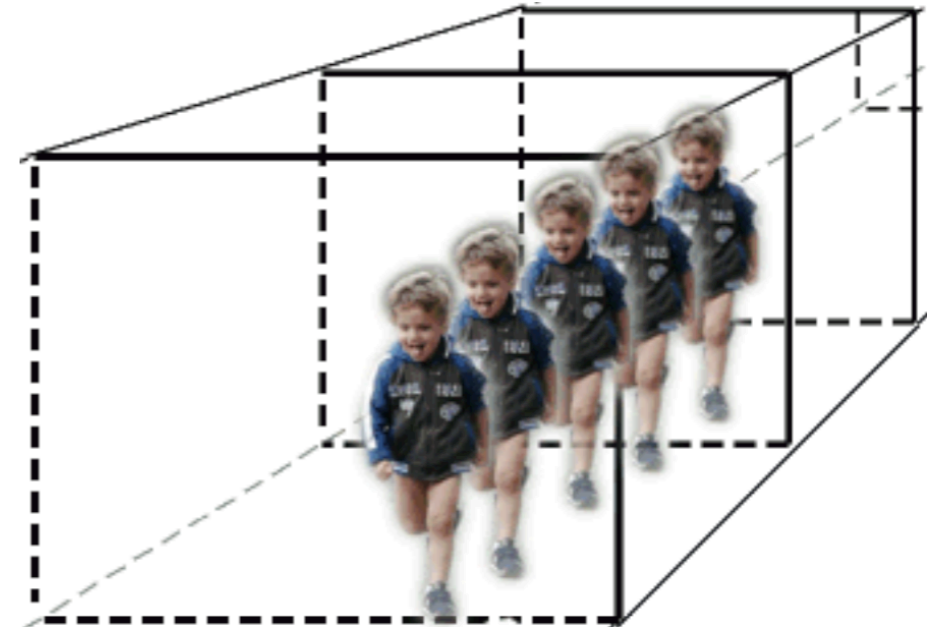
Properties of Real World Objects

- Occupy finite 3D volume
- Cohesive -- Contiguous & Continuous
- Have locally smooth shape



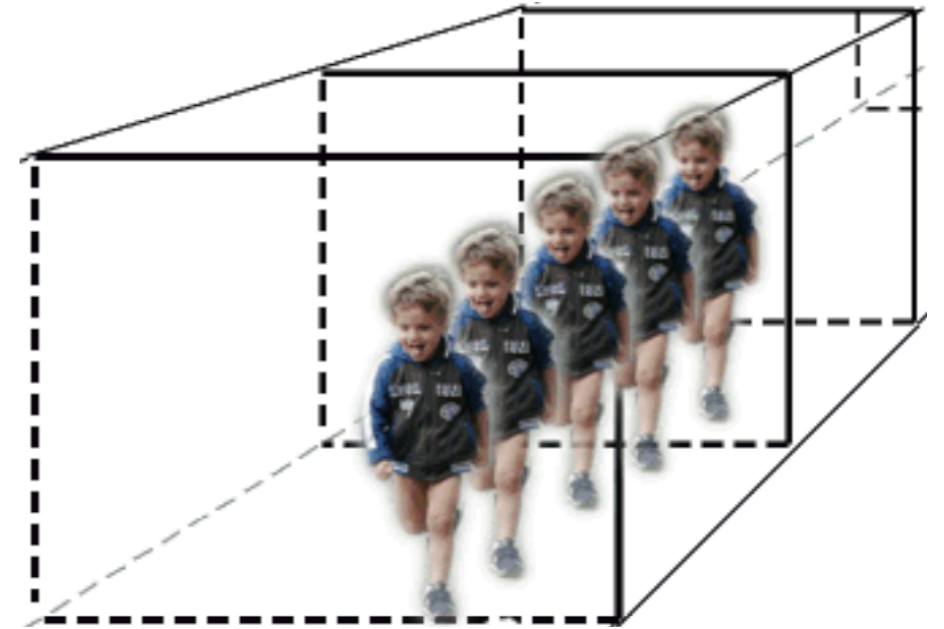
Properties of Real World Objects

- Occupy finite 3D volume
- Cohesive -- Contiguous & Continuous
- Have locally smooth shape
- Made locally of the same material



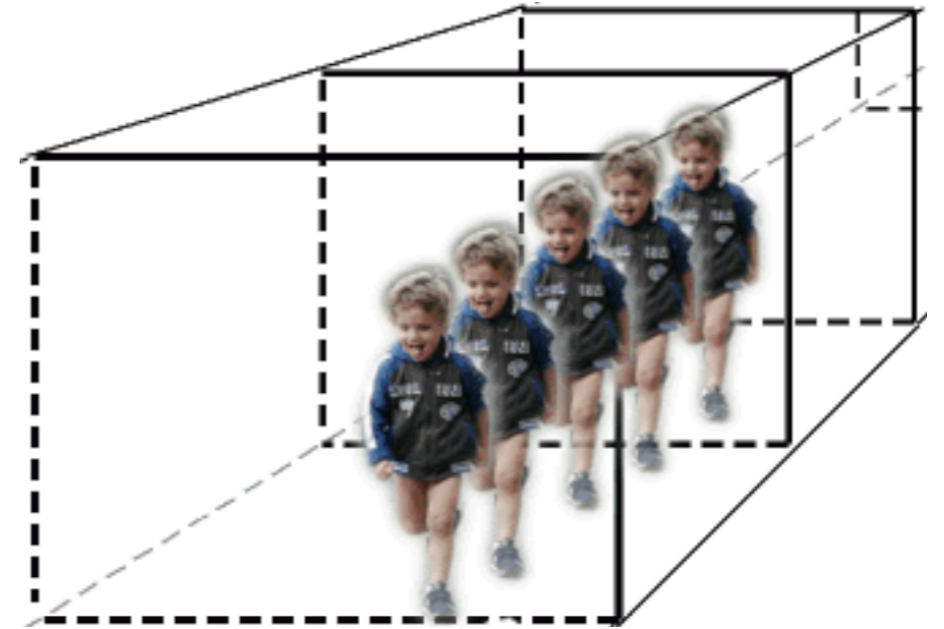
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- Made locally of the same material
- Opaque or partially transparent



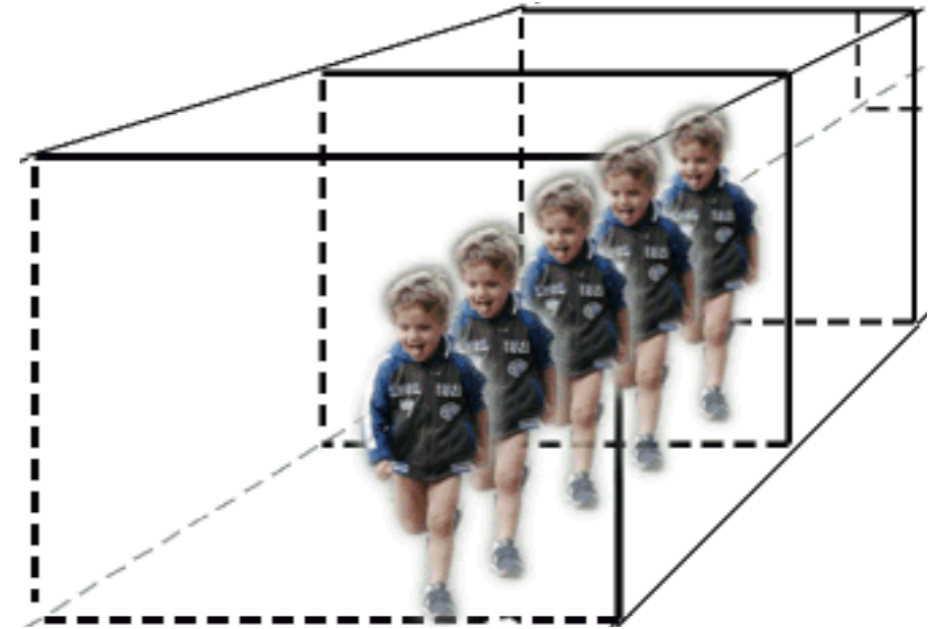
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- Occupy finite 3D volume
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- Have certain photometric properties (e.g., reflectance, specularity)



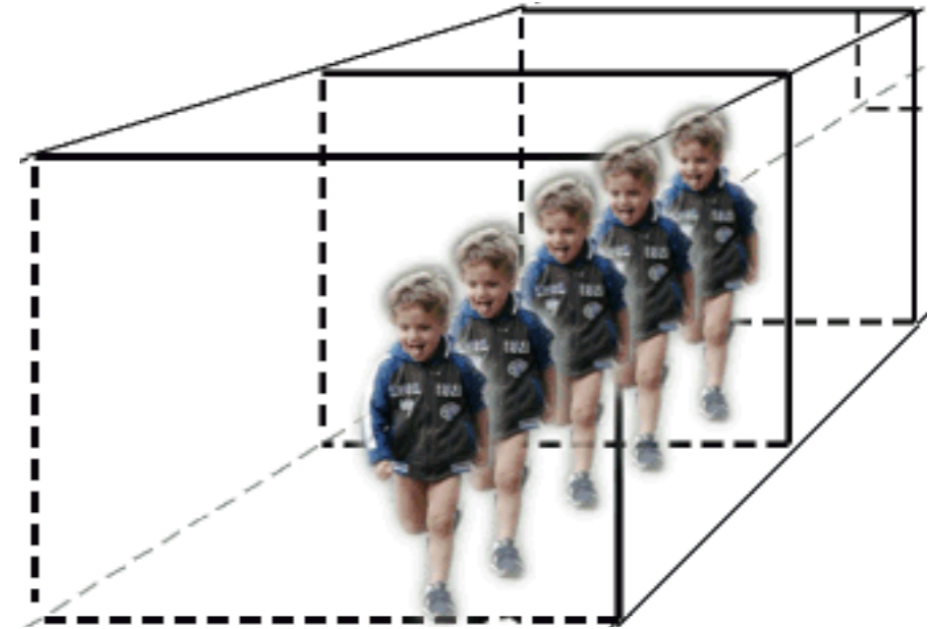
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- Occupy distinct locations



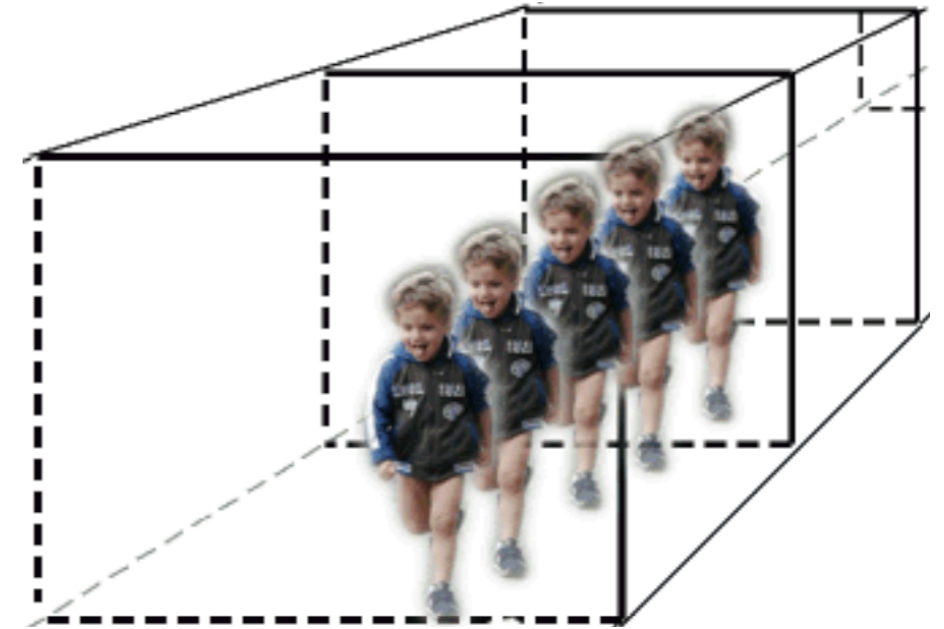
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- Occupy distinct locations
- Form characteristic spatial configurations with other 3D objects



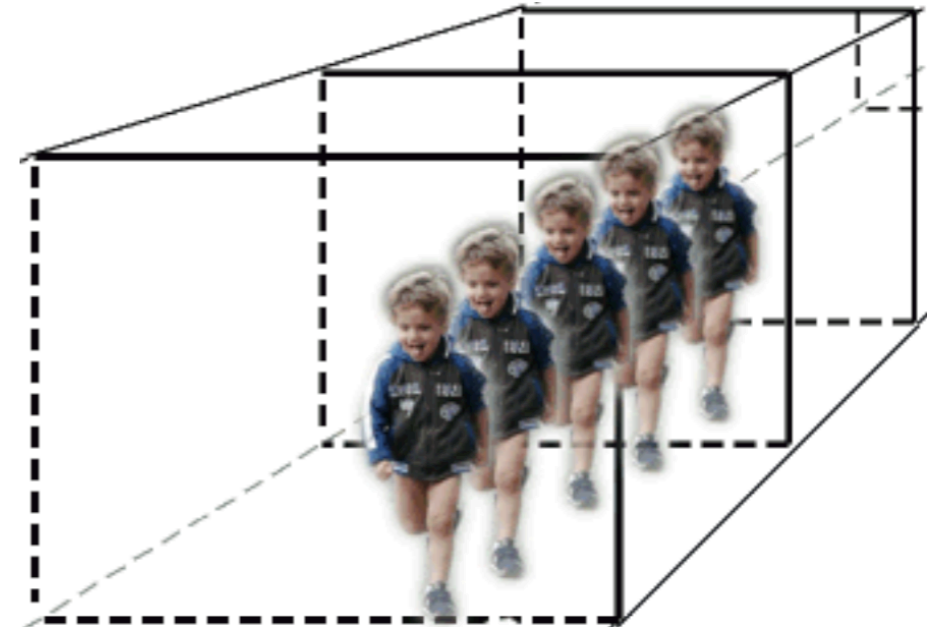
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- Have certain photometric properties (e.g., reflectance, specularities)
- Occupy distinct locations
- Form characteristic spatial configurations with other 3D objects
- Form characteristic temporal configurations with other 3D objects
- Consist of parts that are valid 3D objects in their own right



What low-level image properties capture
the general constraints of real world?

Image Formation + Real World Properties → 2D Objects

Finite 3D volume + Cohesive → 2D objects = Regions

- Region boundaries coincide with 2D object boundaries
- Characteristic size and shape of 2D objects

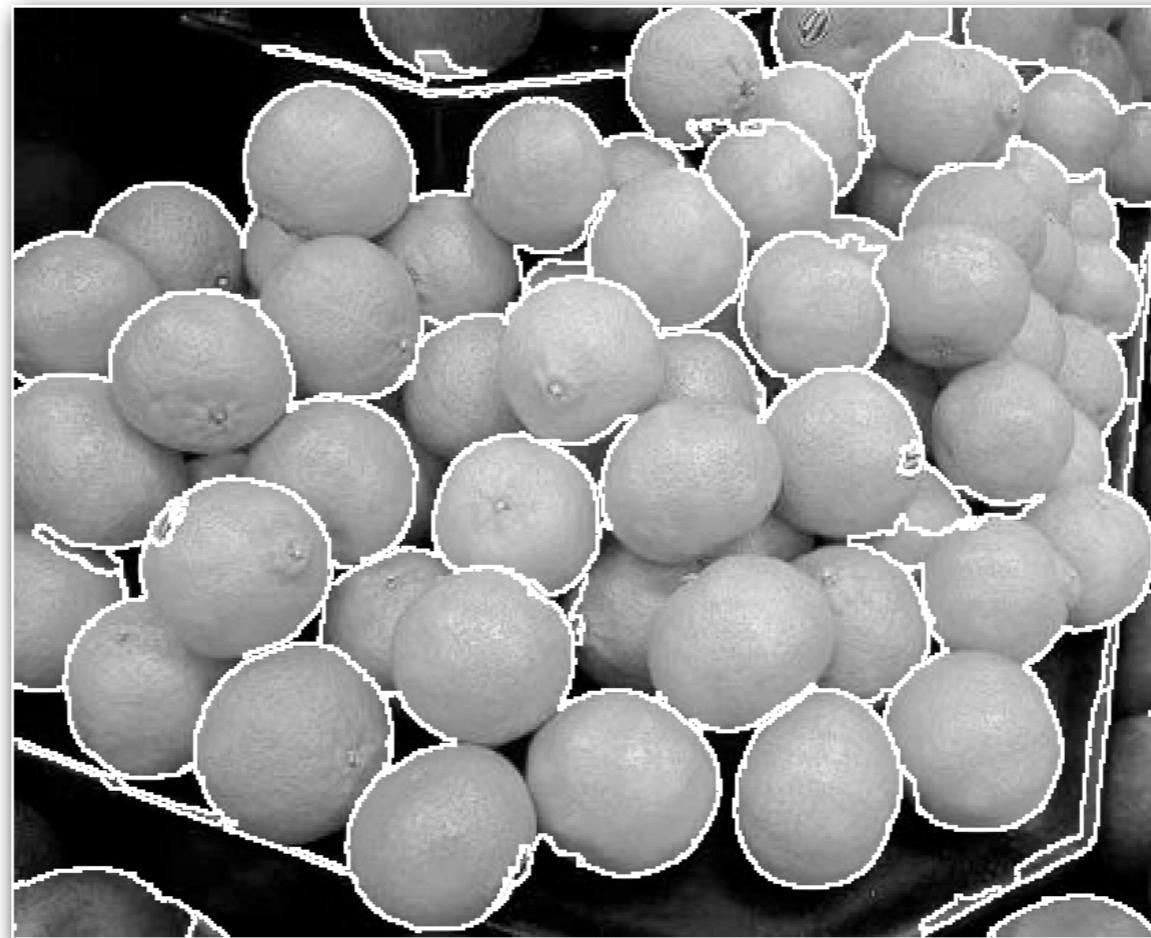


Source: N. Ahuja

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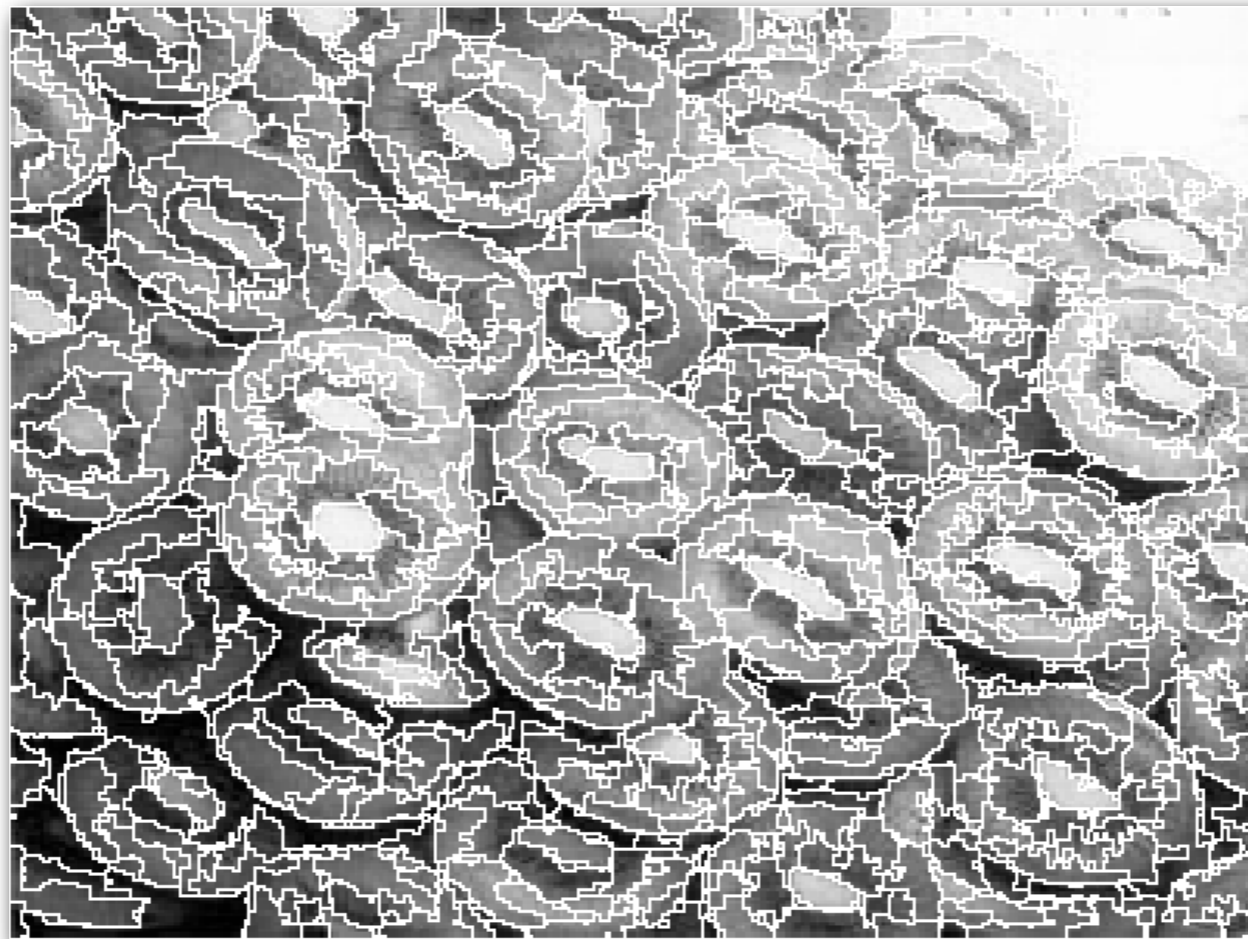
- Opaqueness + Photometry → Color, brightness of 2D objects
- 3D local smoothness → Locally smooth shape, color of 2D objects
- 3D locations → 2D spatial layout, partial occlusion
- Compositionality → Smaller regions embedded within larger ones



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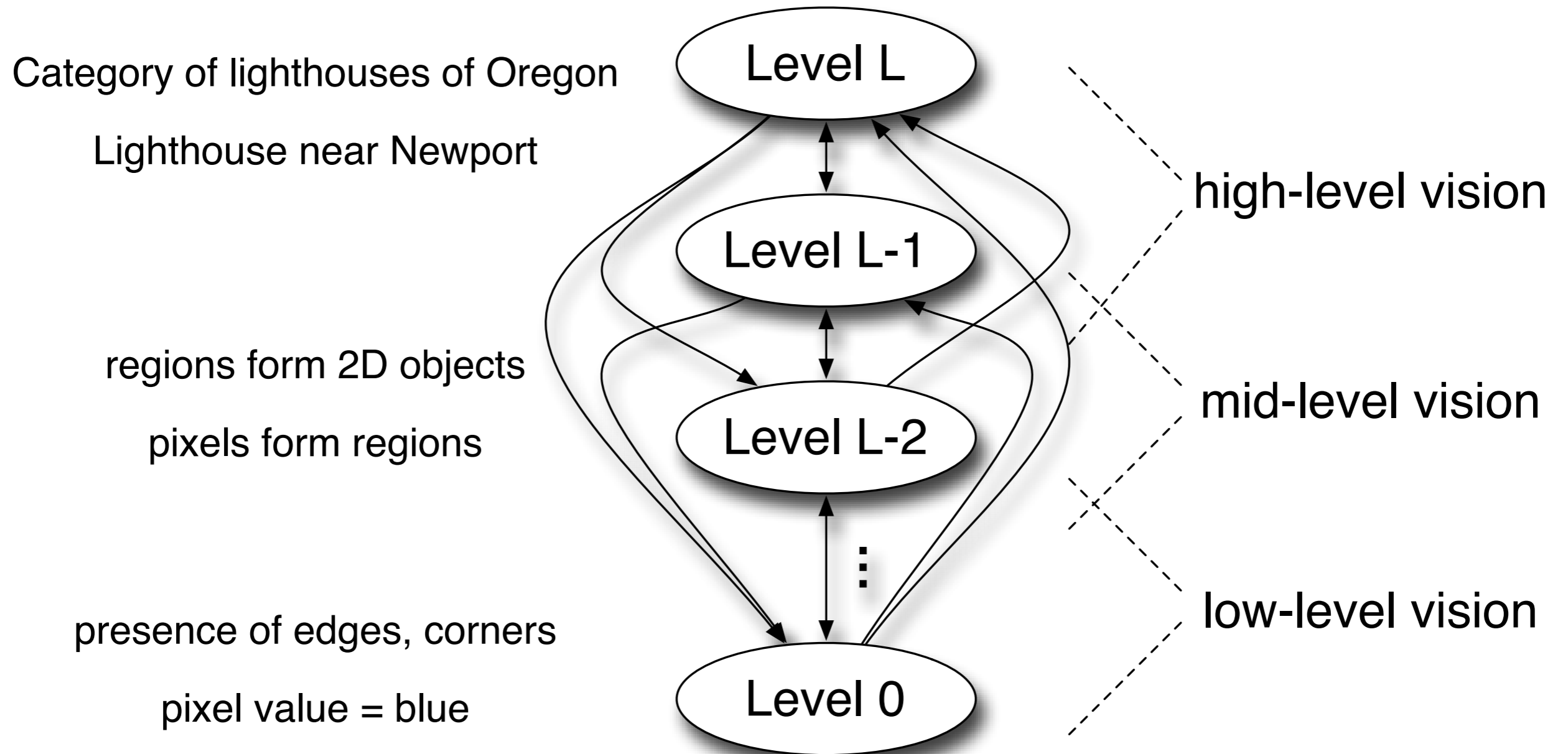
From Low-Level Image Properties to Interpretation



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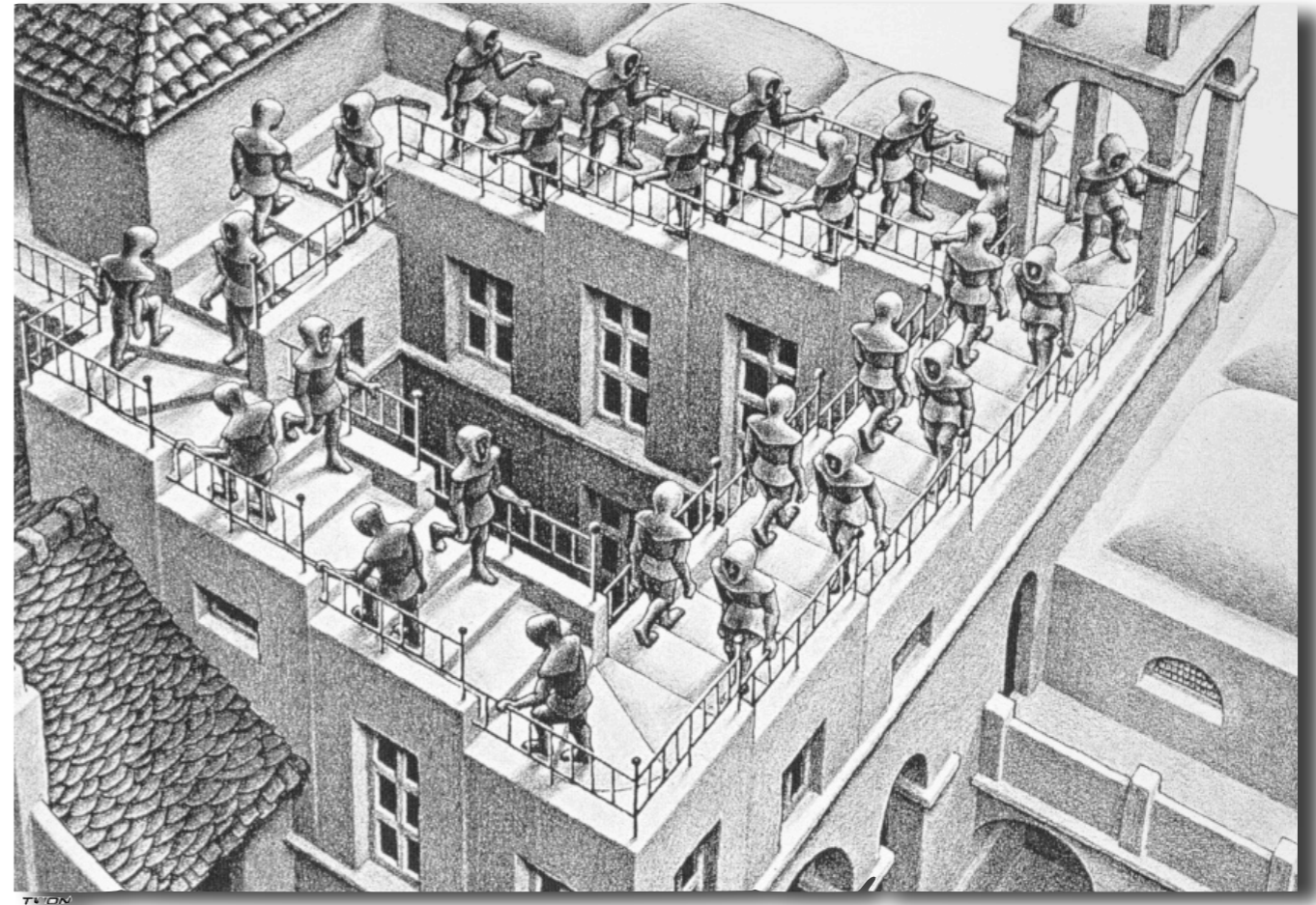
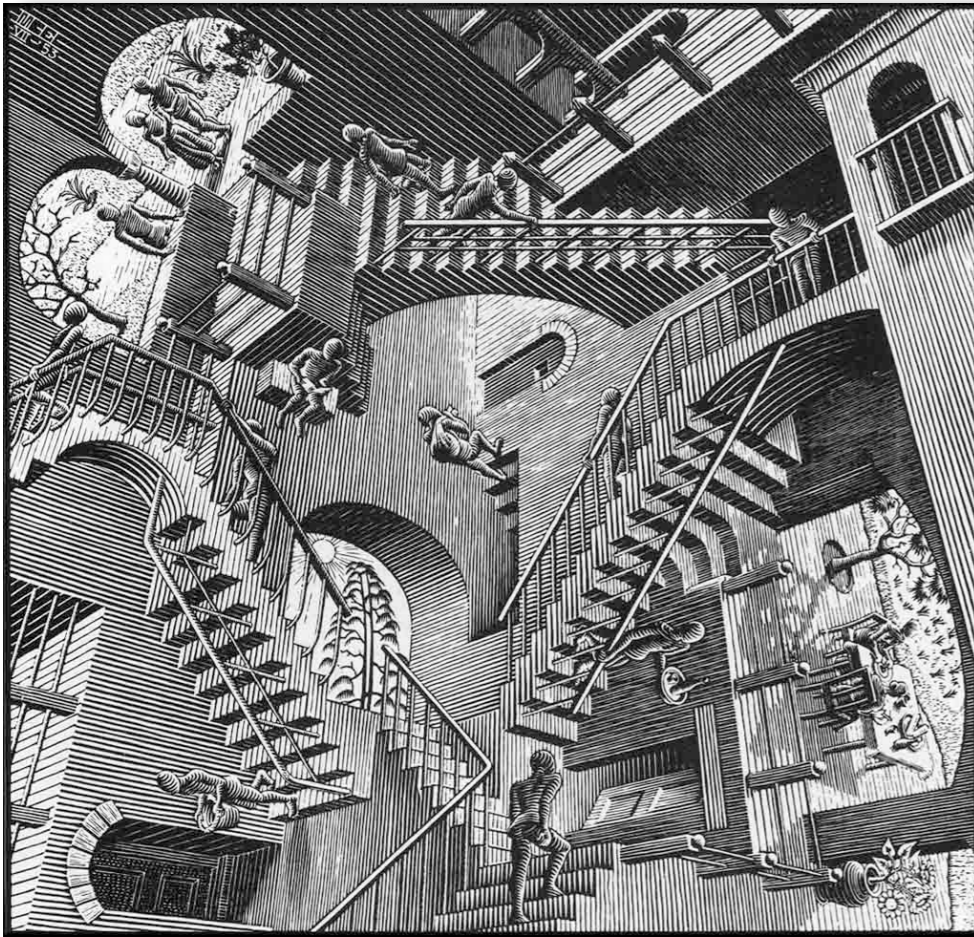


Interpretation = Traversal Across Knowledge Levels



- Bottom-up traversal -- e.g., perceptual grouping
- Top-down traversal -- e.g., context

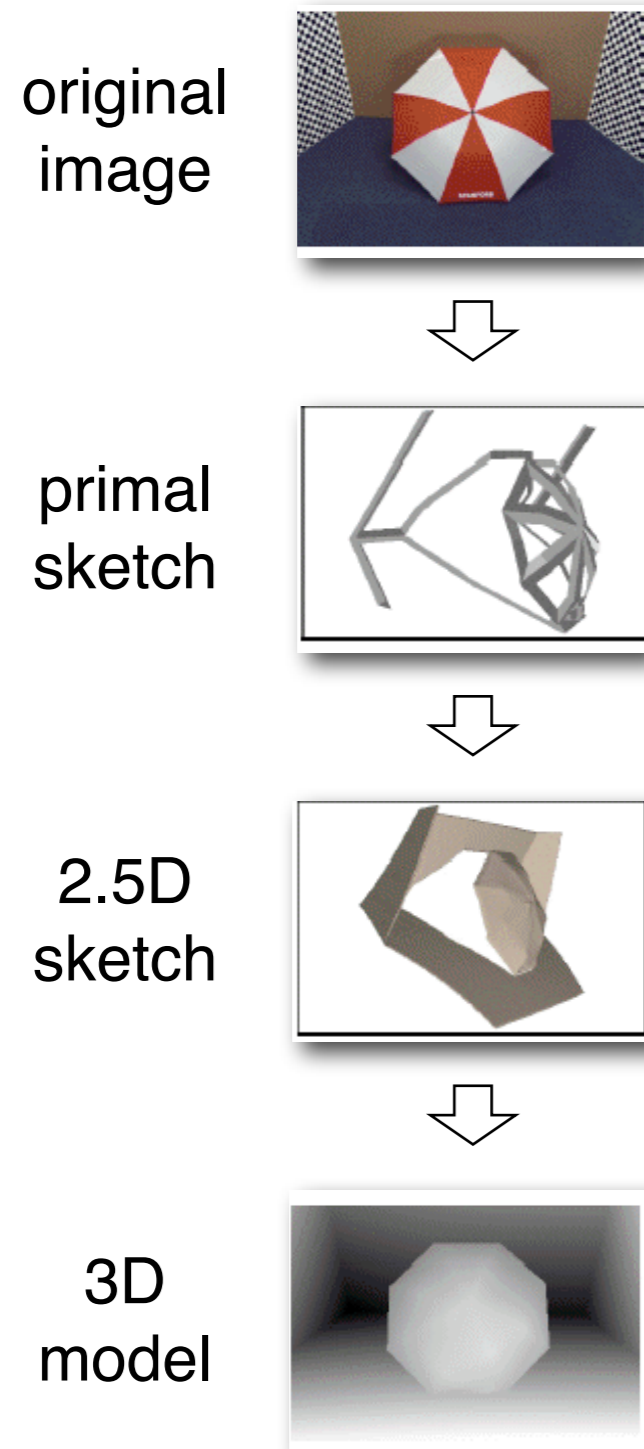
Bottom-up Reasoning Important, But...



- A sense made from analyzing low-level constraints, imposed by real world, last only until cognitive scrutiny takes over
- Good syntax does not imply good semantics

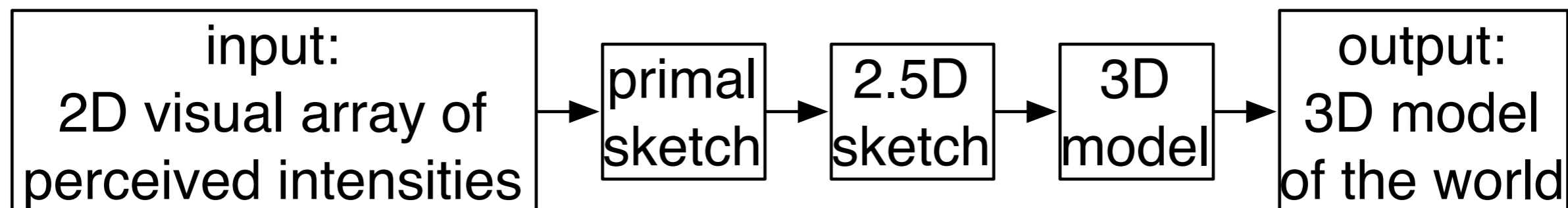
How to conduct image interpretation?

David Marr's Paradigm -- Feedforward Approach



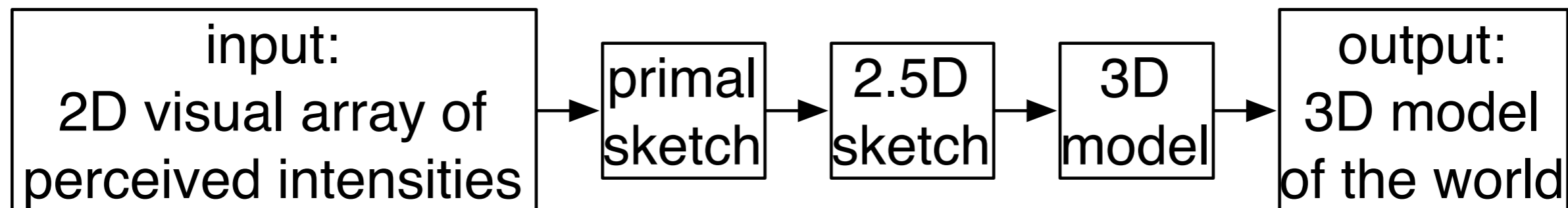
- Pros:
 - Simple
- Cons:
 - Progression often not clear
 - Early mistakes unrecoverable
 - No synergy between the stages

David Marr's Paradigm



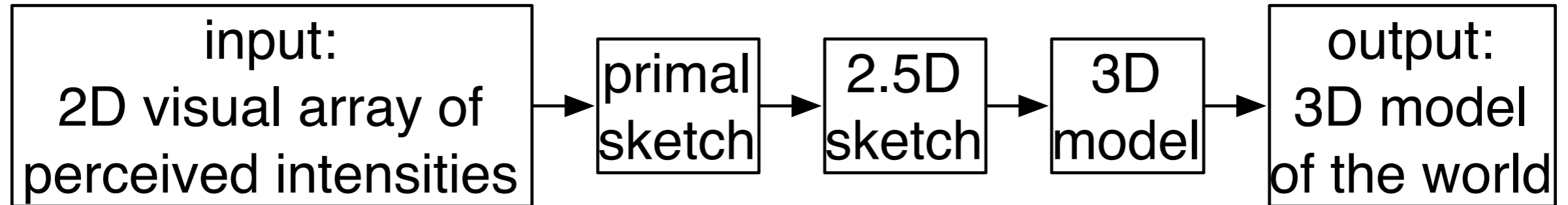
- Human vision = Stageswise processing
- From an image, we can reason what is where in the world
⇒ We must have representations of visual information
- Recognition matches 3D models from a catalogue to the extracted 3D representation in output

David Marr's Paradigm



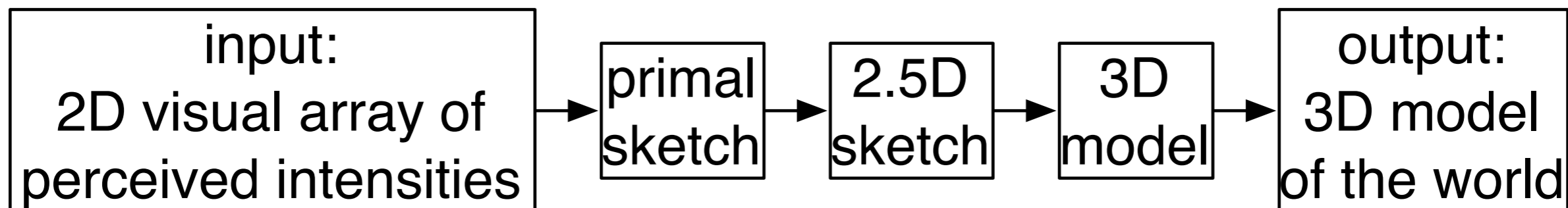
- Perceived intensities depend on:
 geometry reflectance
 illumination viewpoint
- Perceived intensities \leftrightarrow Structure of the visible surfaces
- Primal sketch = Feature extraction (edges, corners, regions)
- Each feature is a 5-tuple:
 type position contrast
 orientation scale

David Marr's Paradigm

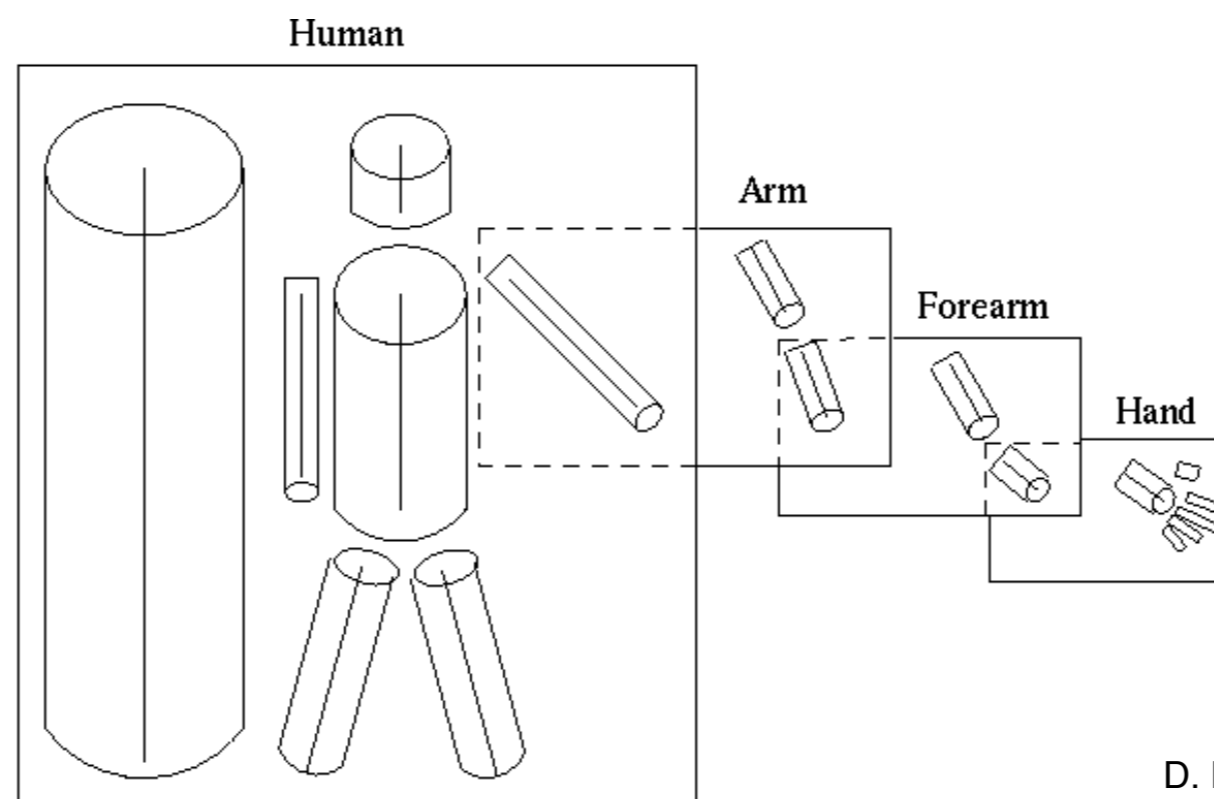


- 2.5D sketch =
 - Textures
 - Depth map
 - Surface orientation
 - Discontinuities
- Viewer centered coordinate system

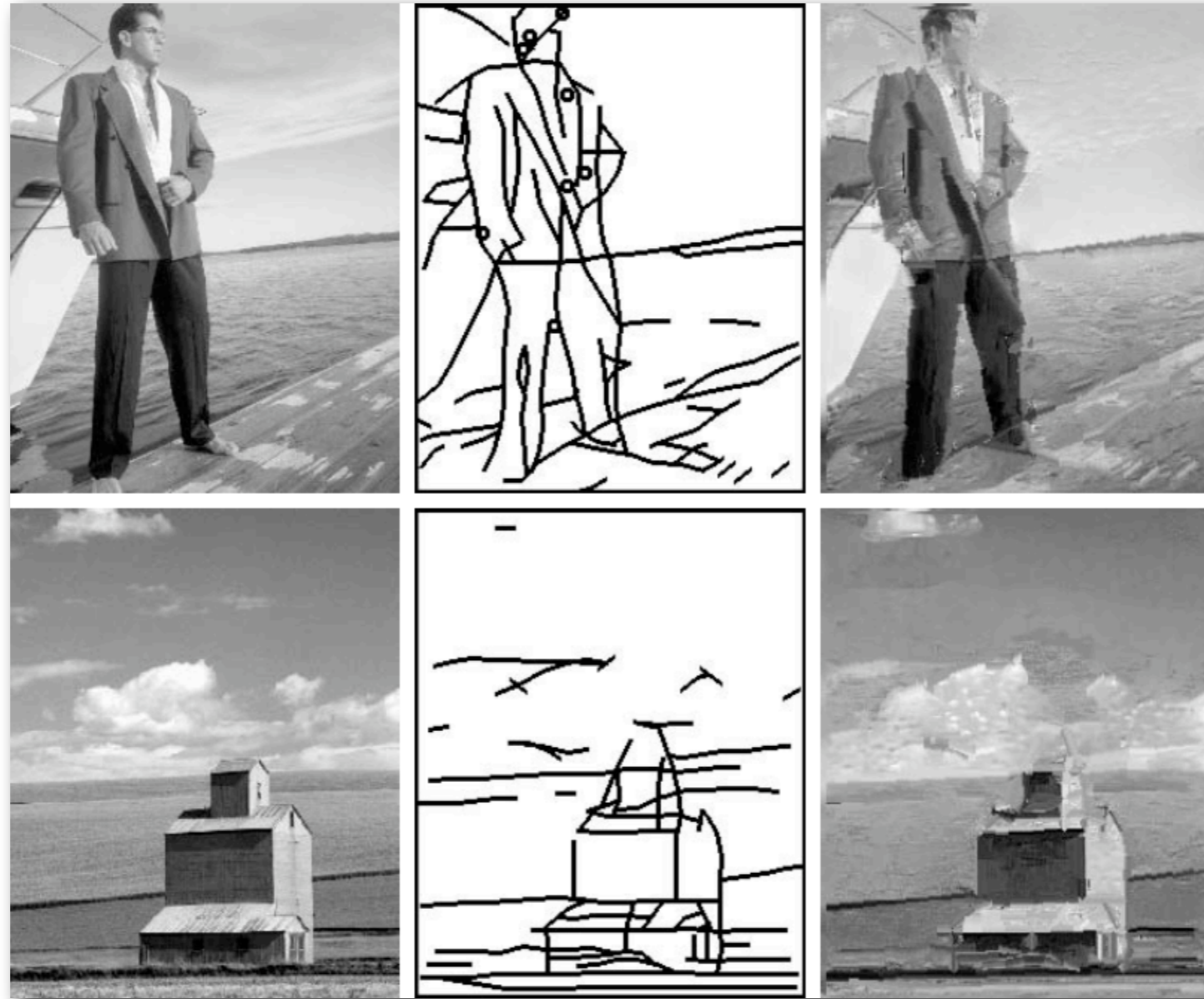
David Marr's Paradigm



- 3D model = Modular and hierarchical representation of 3D world
- Representation in terms of volumetric and surface primitives
- Hierarchy of 3D models wrt specificity of information they carry

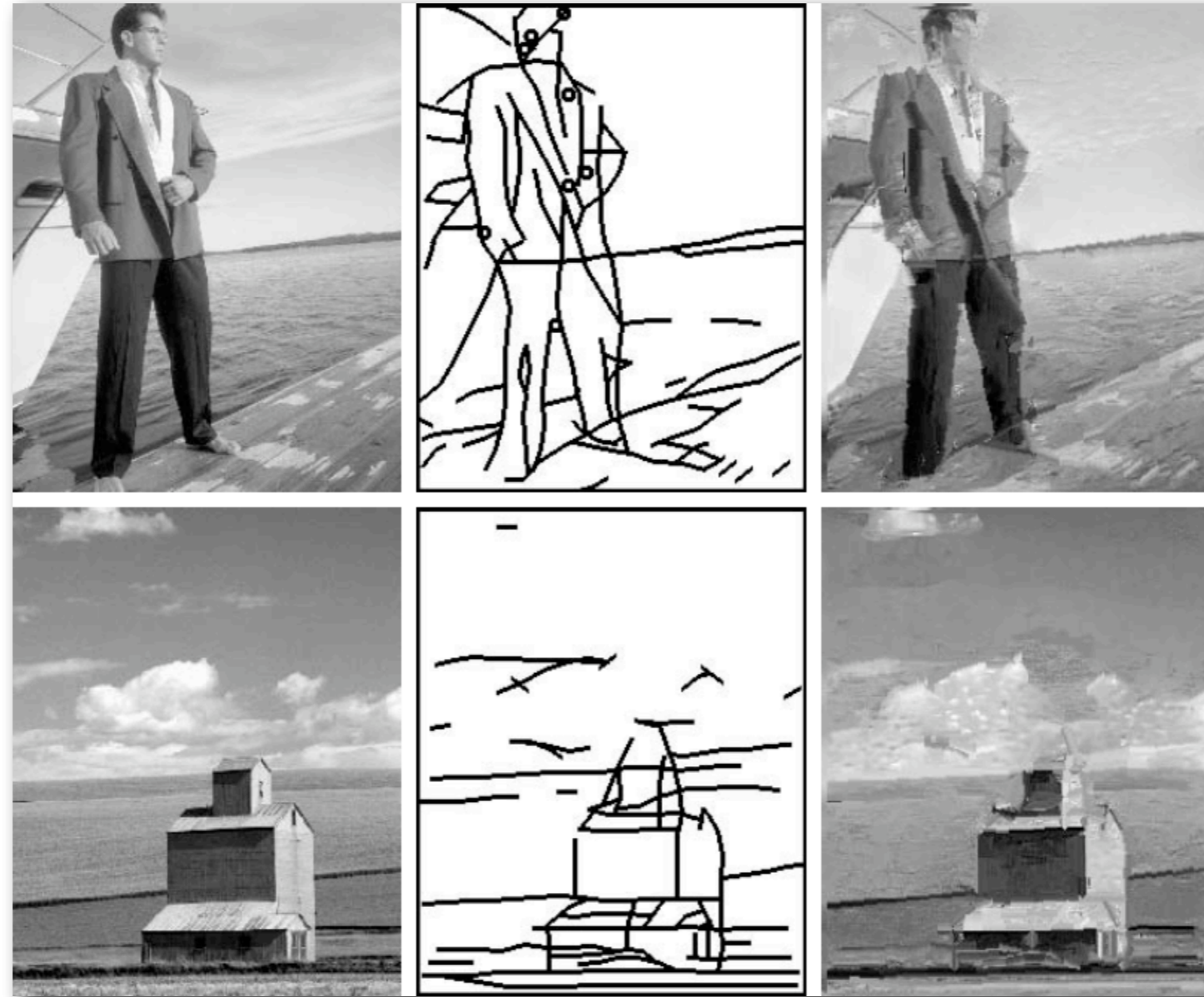


David Marr's Paradigm



original
image

David Marr's Paradigm



original
image

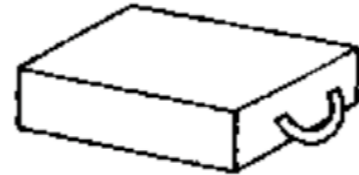
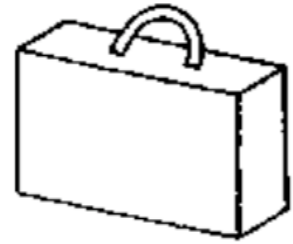
primal
sketch

2.5D
sketch

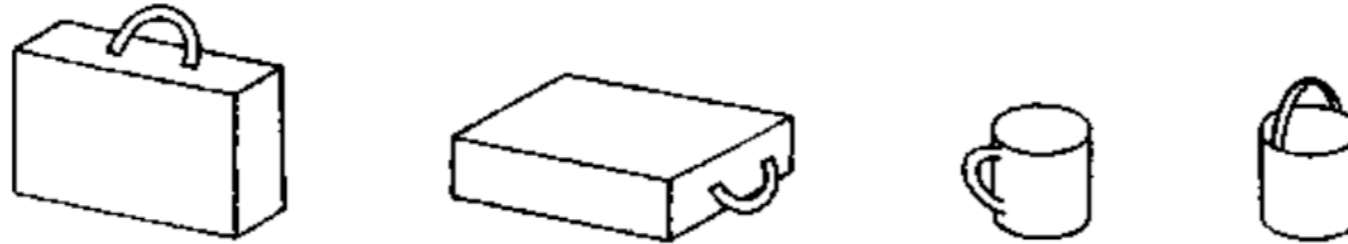
Source: S.C. Zhu et al.

Recognition by Components

Biederman: Recognition-by-Components

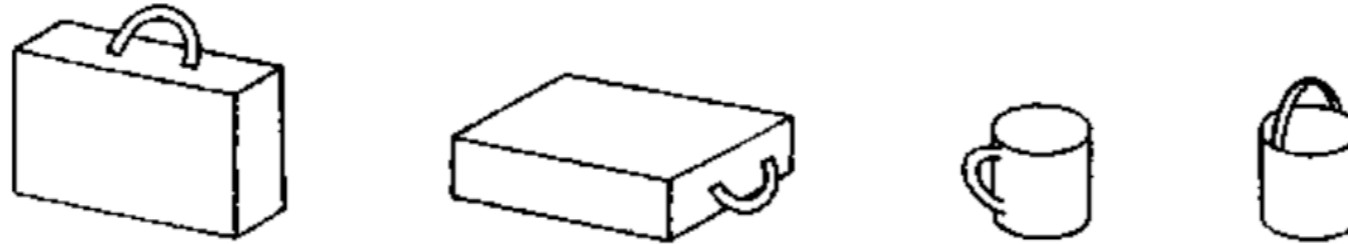


Biederman: Recognition-by-Components



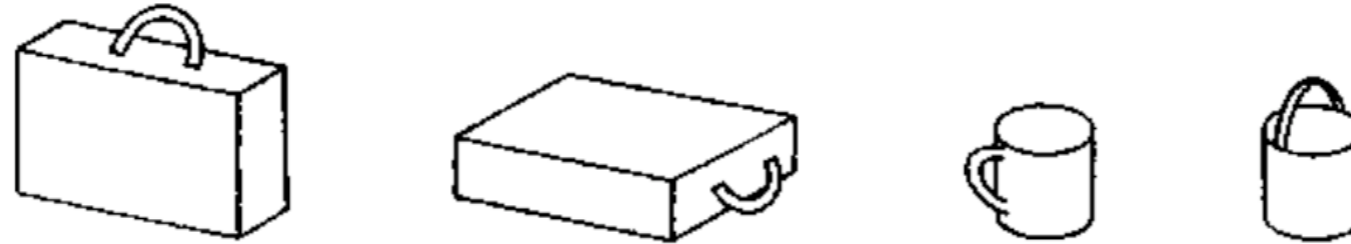
- Objects consist of generalized-cone components -- geons

Biederman: Recognition-by-Components



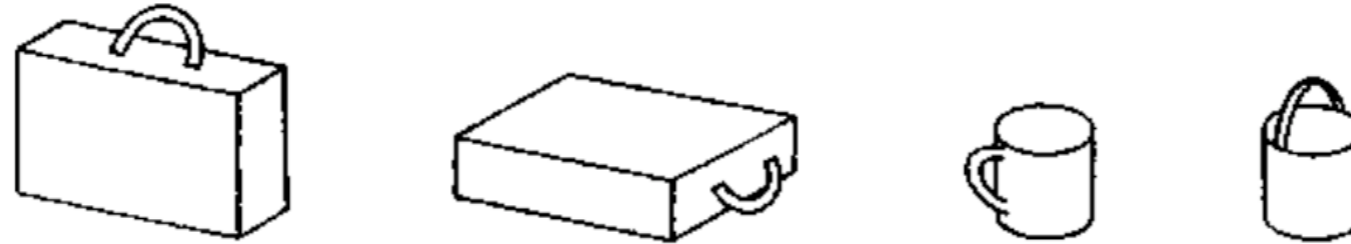
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- There is a finite (small) number of distinct geons, shared by objects

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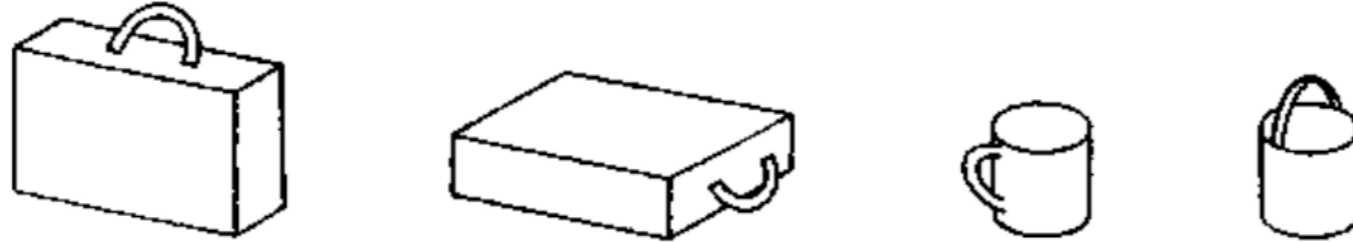
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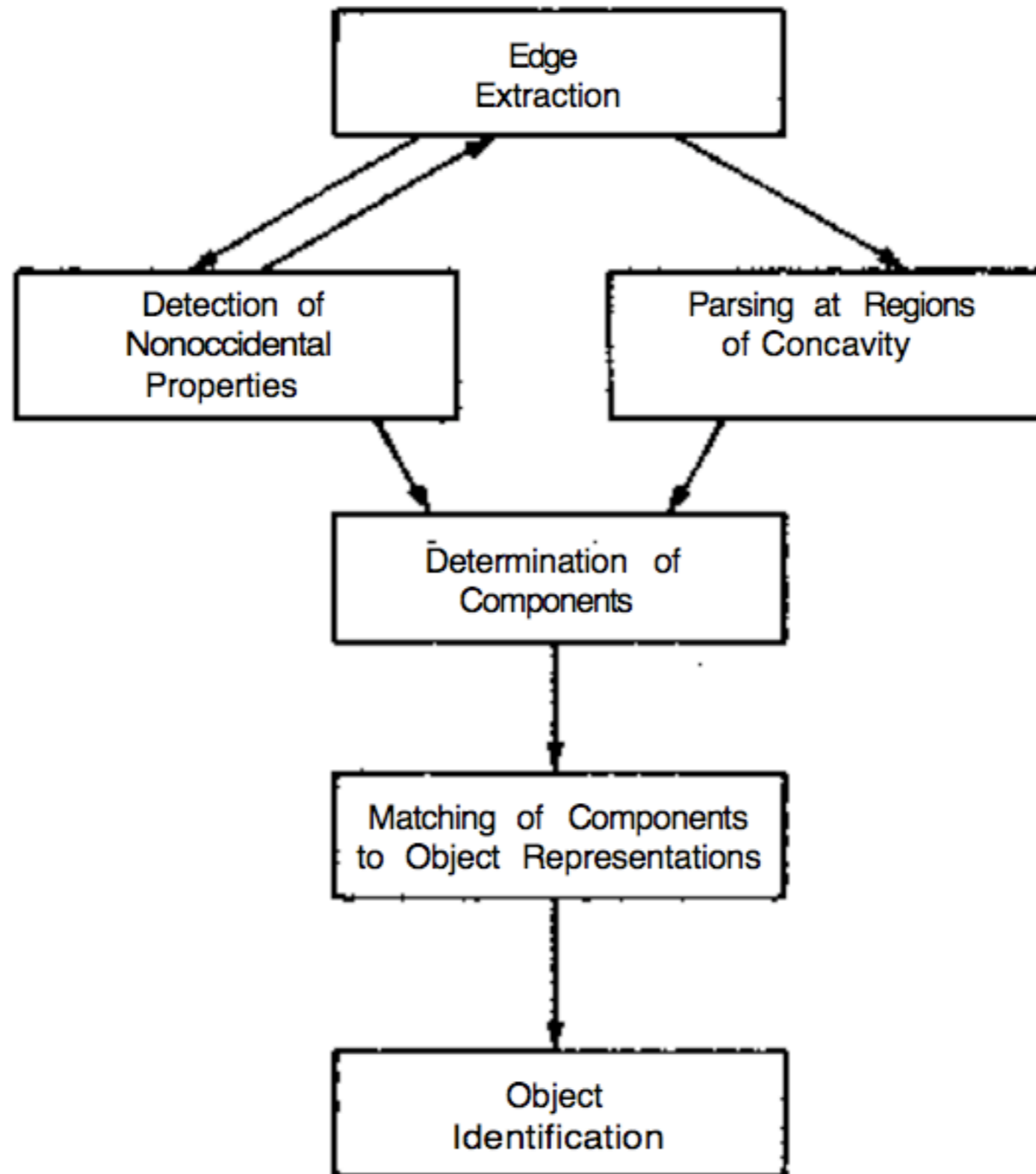
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- Geons can be robustly detected from properties of edges: collinearity, curvature, symmetry, parallelism, and co-termination

Biederman: Recognition-by-Components



- Objects consist of generalized-cone components -- geons
- There is a finite (small) number of distinct geons, shared by objects
- Different arrangements of the same geons produce different objects
- Geons can be robustly detected from properties of edges: collinearity, curvature, symmetry, parallelism, and co-termination
- Objects can be readily recognized from detected geons even if occluded, rotated in depth, deformed, etc.

Biederman: Recognition-by-Components



Intrinsic Images

Barrow & Tenenbaum -- Intrinsic Images



Input Image

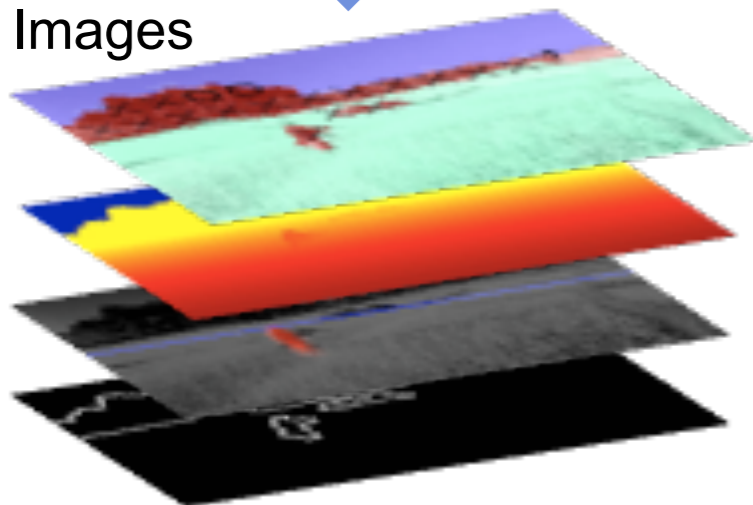


Scene Analysis Processes

Surface Orientation
Object/Viewpoint
Occlusion/Depth



Intrinsic
Images



- Intrinsic images = Confidence maps
- Pros:
 - Synergistic contextual interactions
 - Easily extended
 - Robust
 - Consistency enforced via soft constraints
- Cons:
 - Complex
 - How to fuse flawed and disparate intrinsic images into coherent interpretation?

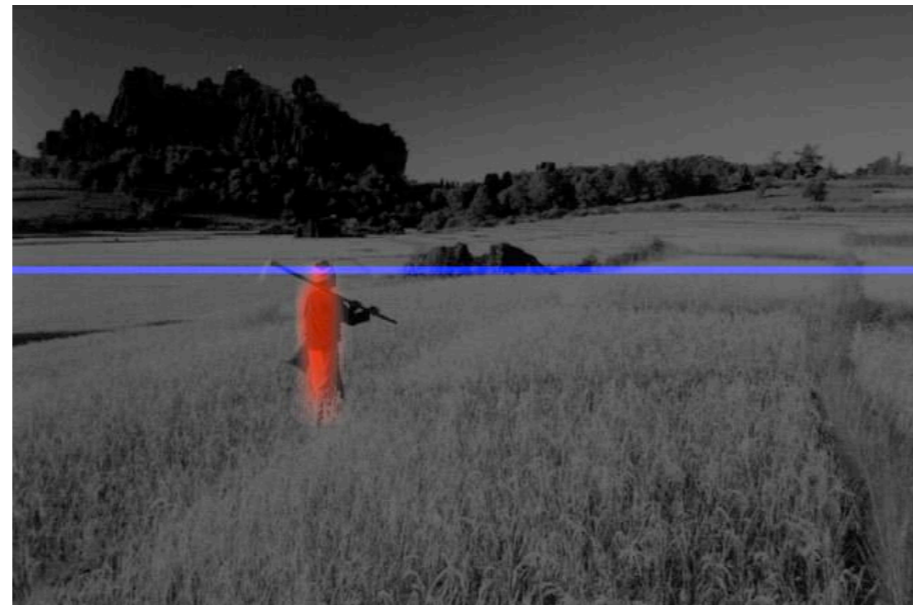
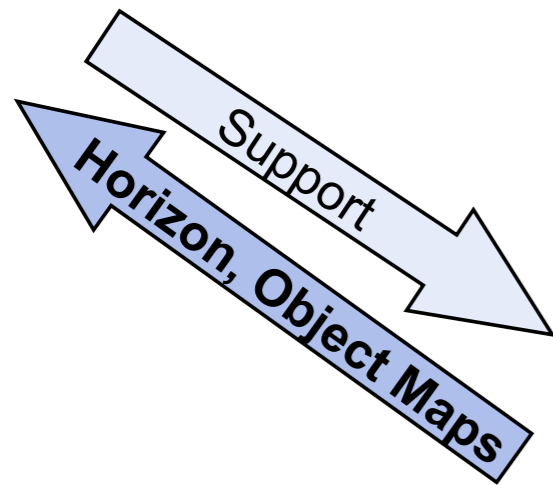
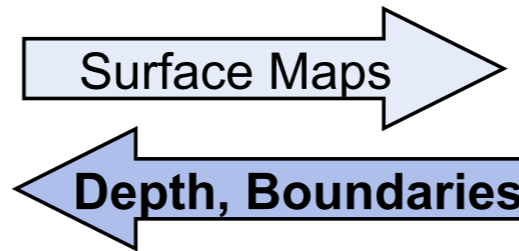
Barrow & Tenenbaum -- Intrinsic Images



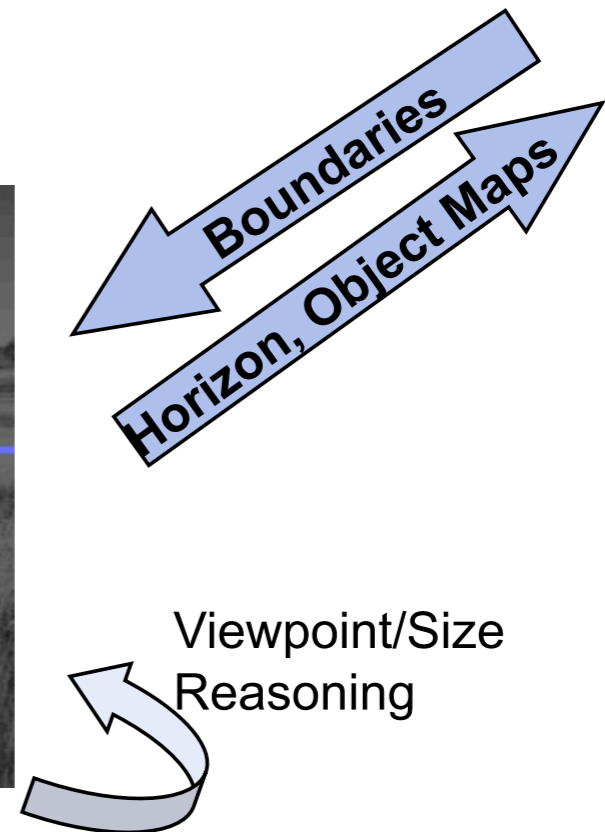
Surfaces



Occlusions



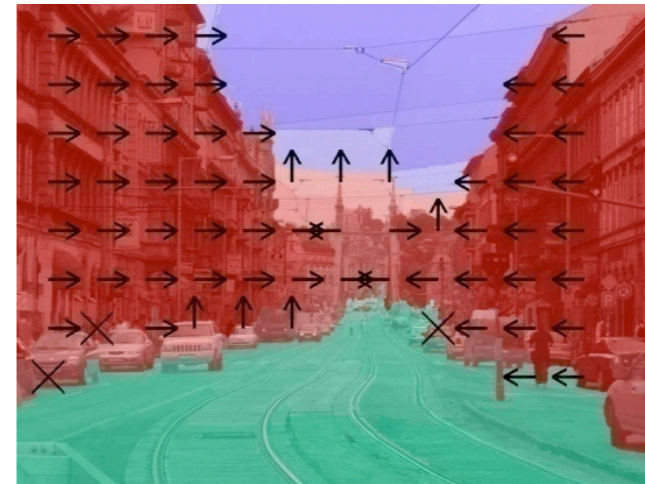
Objects and Viewpoint



Output of Individual Intrinsic Images



input



surface map



occlusion boundaries



object/horizon detection

Results by Fusing all Intrinsic Images



input



surface map



occlusion boundaries



object/horizon detection

Next Class

- Image features
 - Color
 - Edges
 - Interest points
- Homework 1