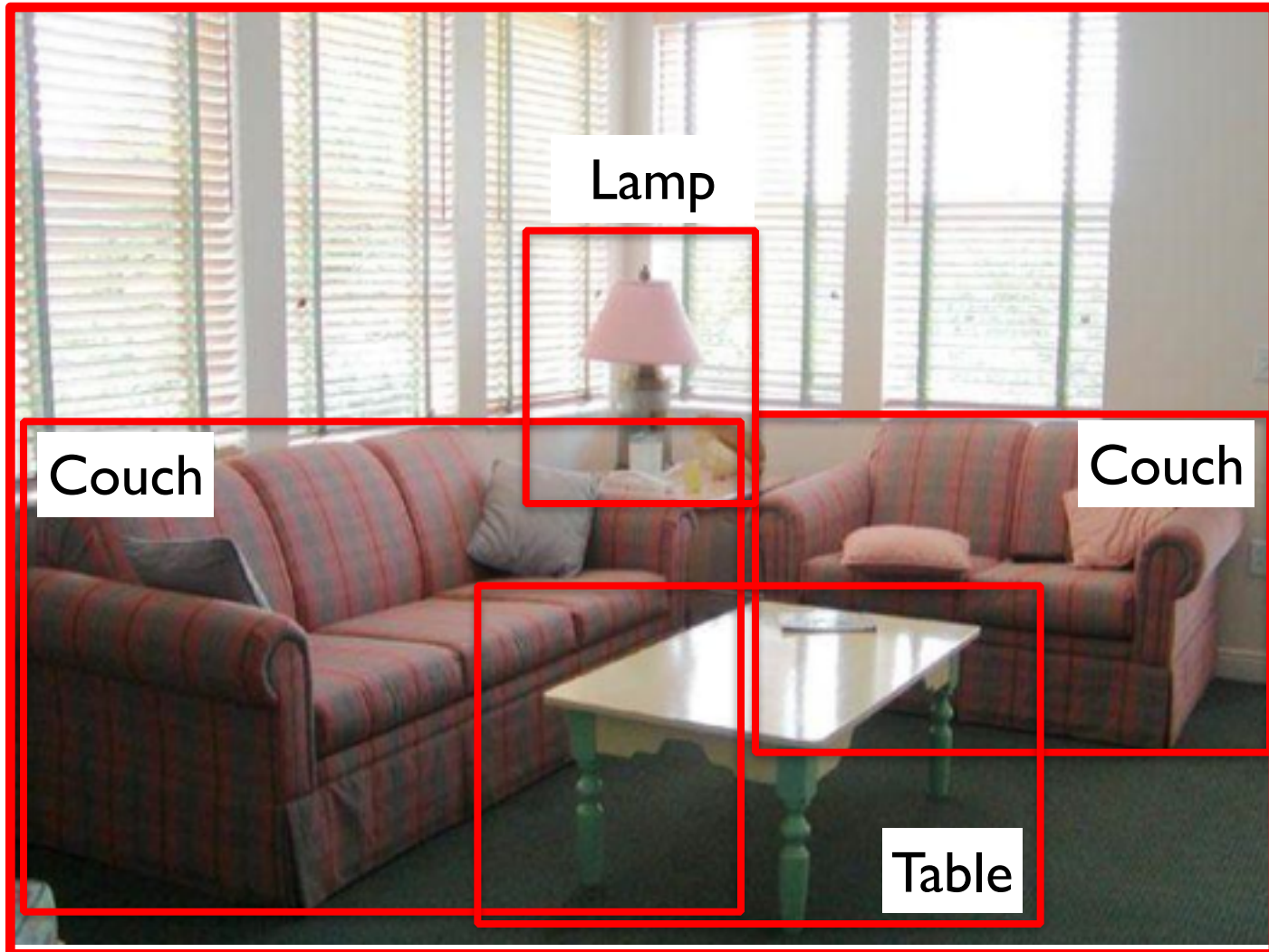


Lecture 2: Theories and History

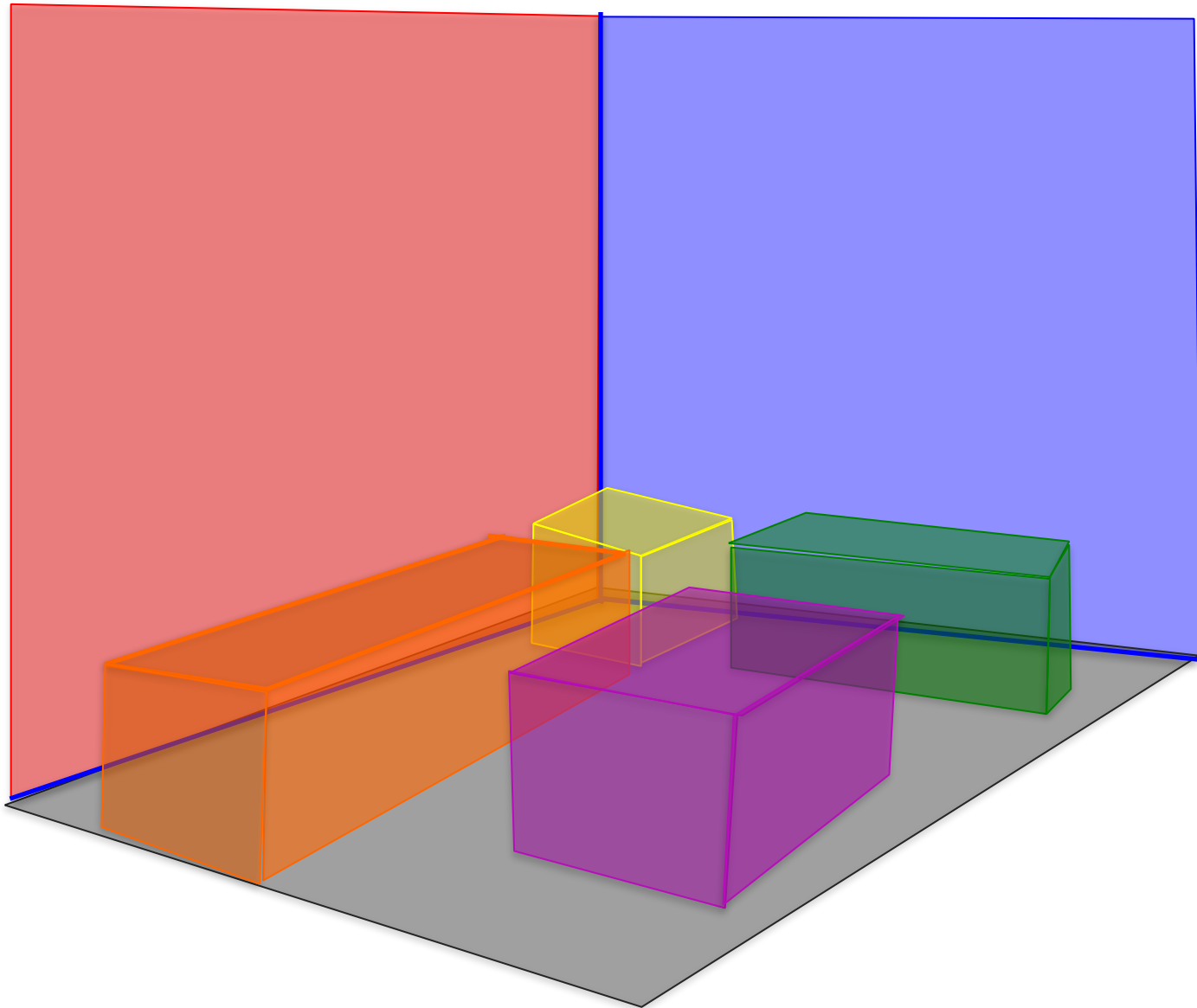
fundamental questions

What does it mean to understand?

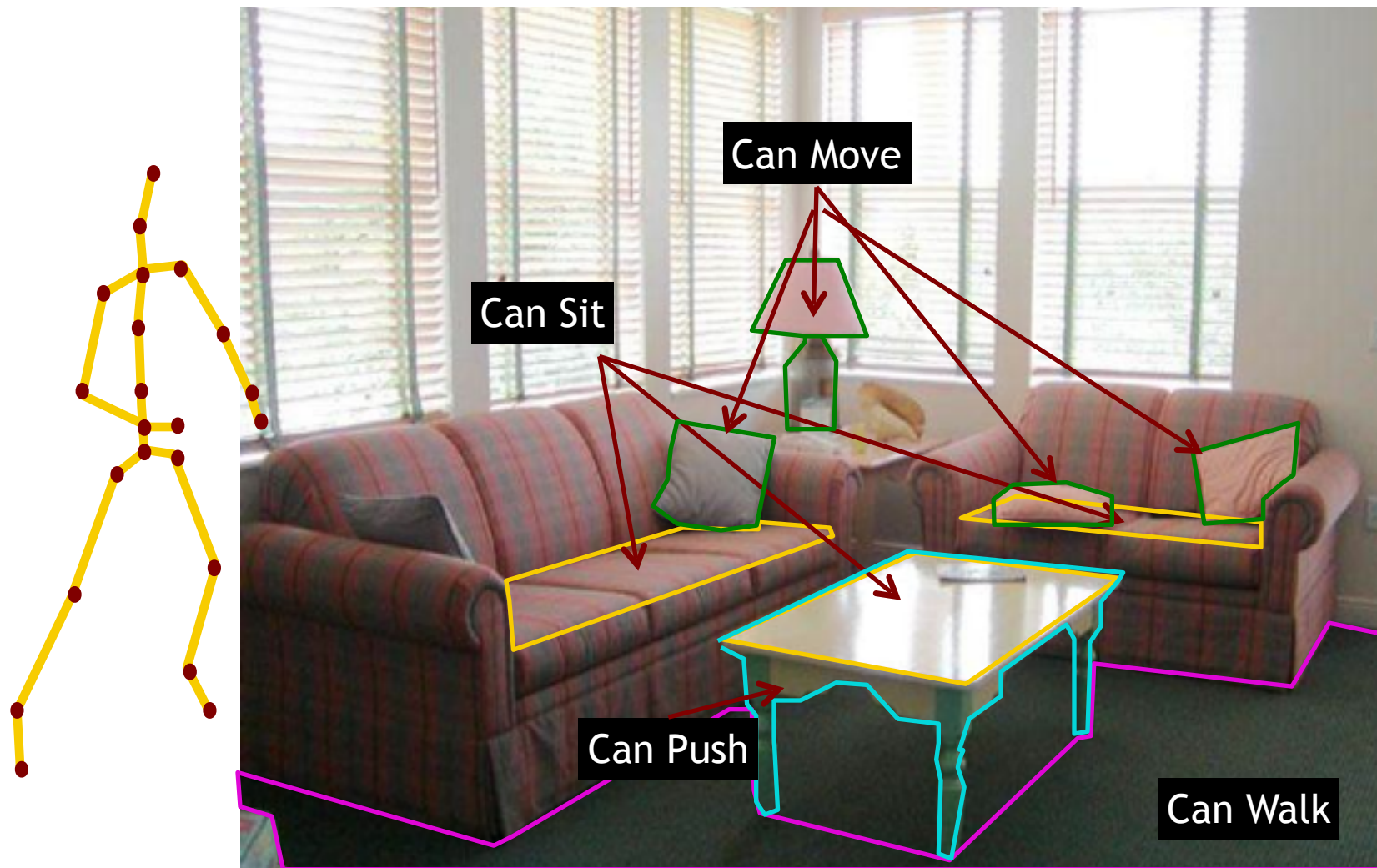
semantics



3D understanding



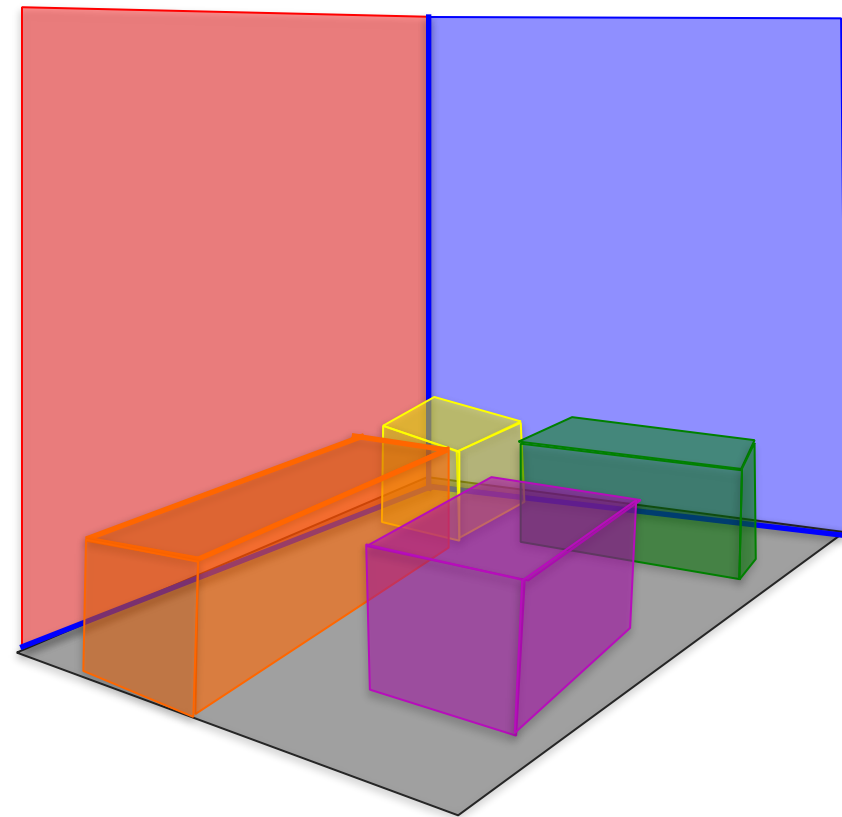
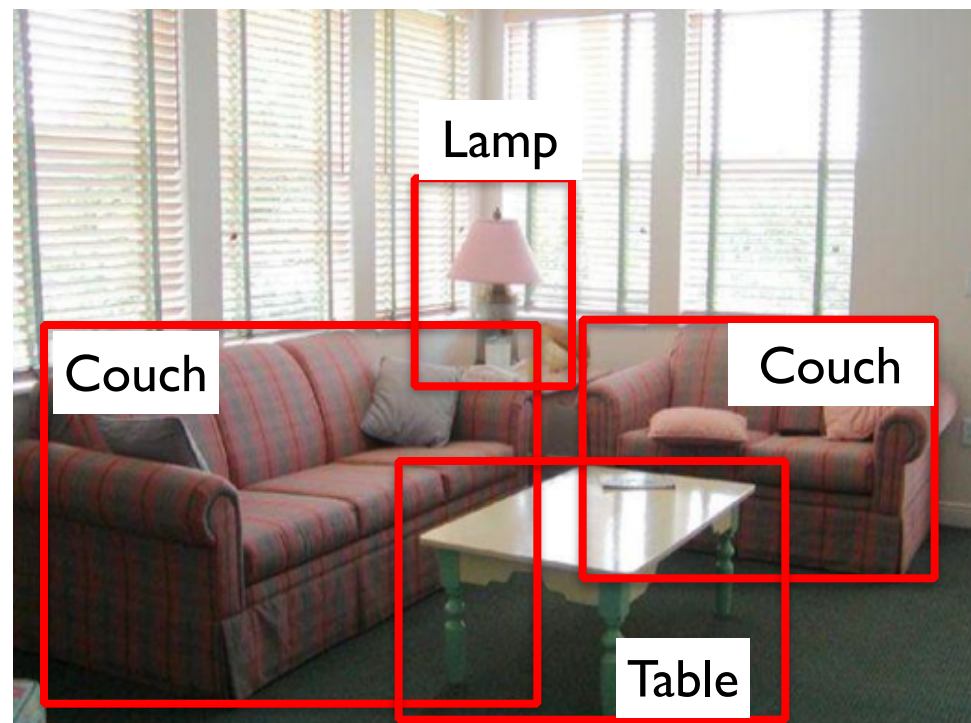
Functions — what can we do?



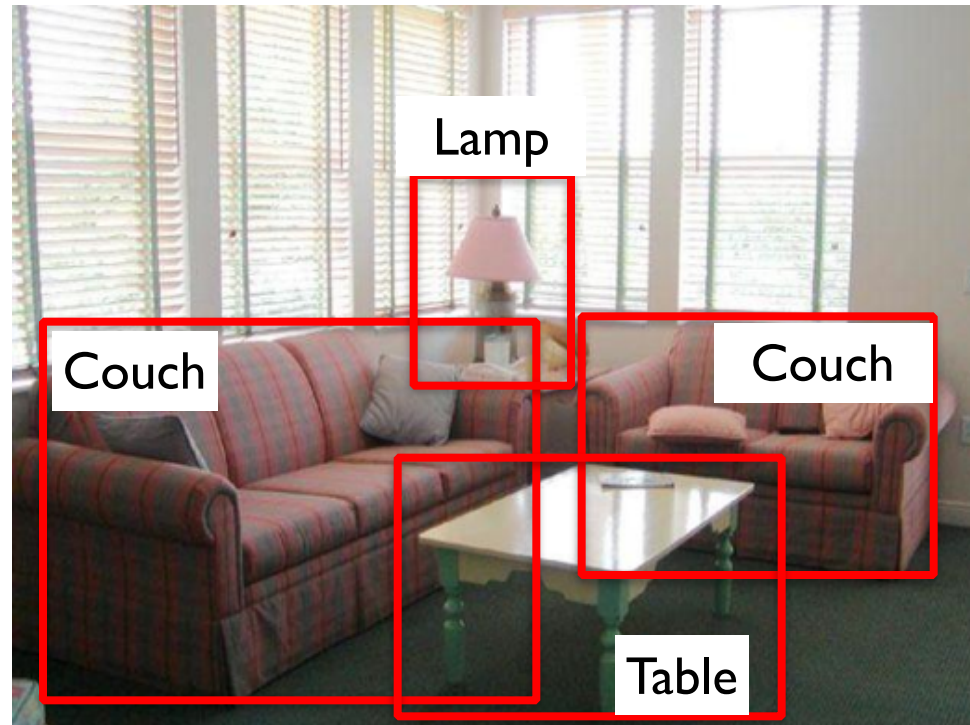
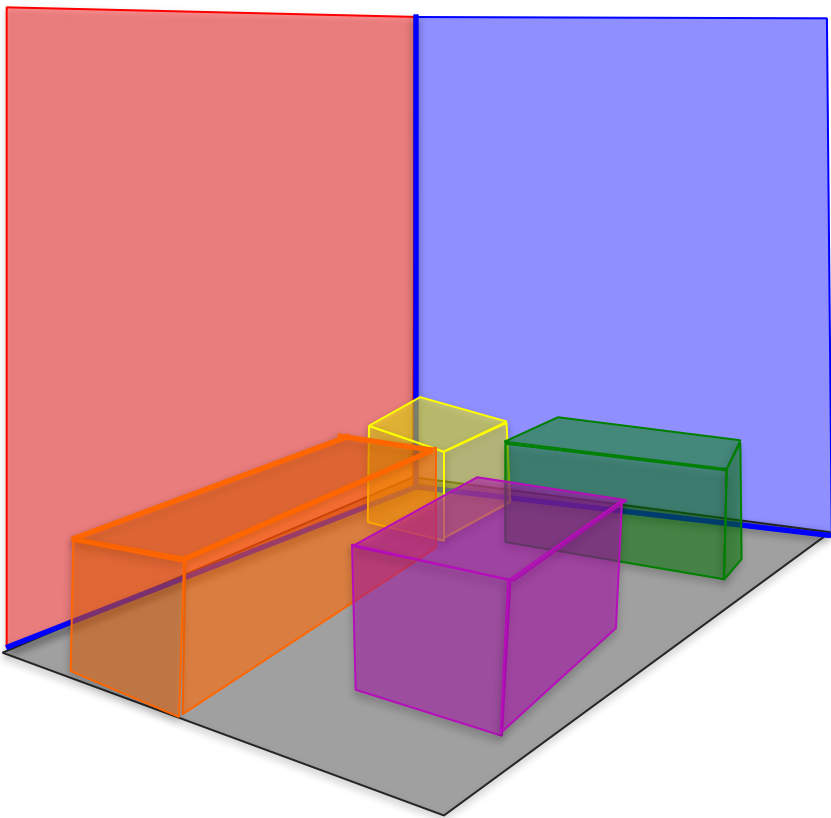
**fundamental
questions**

how are these tasks related?

semantics -> 3D



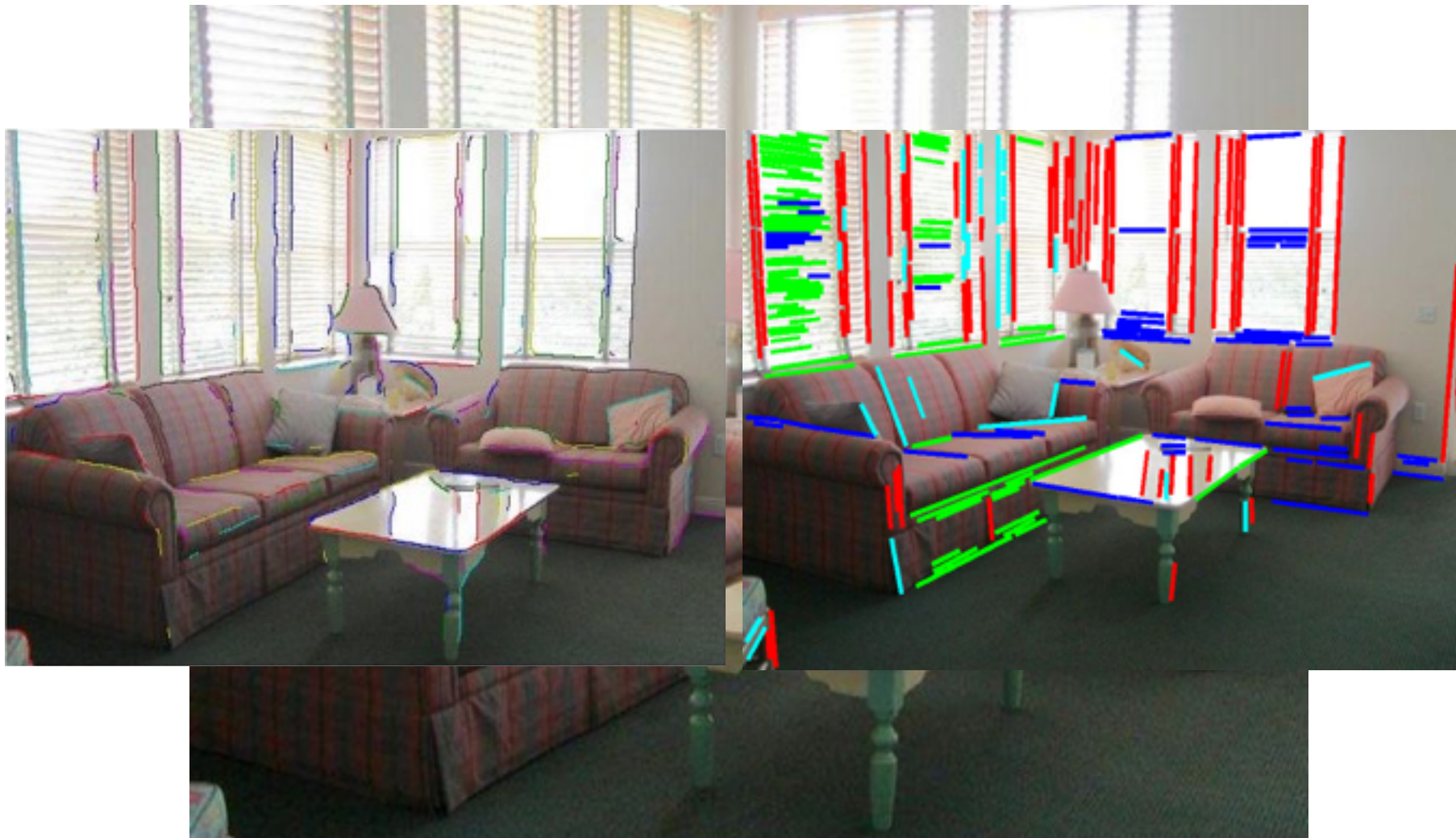
3D -> semantics



fundamental questions

What are the processes involved?

Lines, contours, segments



object detection



matching scenes



how do we answer all these questions?



how do we answer all these questions?

Understanding Human Vision

Goal: Understand Perception

Approach 1: Discover all the facts related to perception !

Is that even possible?

Understanding a process as complex as perception requires more than listing facts

Theories

What is a Theory?

Theory

An integrated set of related hypotheses about underlying mechanisms or principles that organize, explain, and predict facts about the phenomena of a given domain.

What is a Theory?

Three important functions of a theory:

- Integrate “old” facts

- Predict “new” facts

- Lead to “understanding”

Three criteria for evaluating a theory:

- Logical consistency

- Empirical adequacy

- Parsimony (“Occam’s razor”)

Theories of Vision

Categorization of Theories of Vision

- “*Why do things look as they do?*”
-- Kurt Koffka (1935)
- Nativism vs. Empiricism
 - “Because we were born (*evolved*) to see them that way” vs.
 - “Because we have learned to see them that way”
- Atomism vs. Holism
 - “because of the way each pixel appears” vs.
 - “because of the way the entire scene appears”
- Organism vs. Environment
 - “Because we are the way we are” vs.
 - “Because the world is the way it is”

4th Axis

- How do you derive the theory?
 - Introspection (Conscious Inference)
 - Behavior (Measurement of Human Performance)
 - -Koffka: why do the things look the way they do?
 - -What does vision enable us to do?

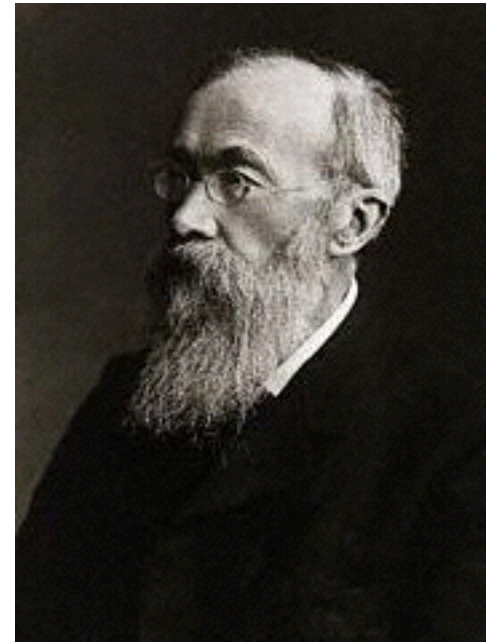
Structuralism

THEORY	NATIVISM vs. EMPIRICISM	ATOMISM vs. HOLISM	ORGANISM vs. ENVIRONMENT	PRINCIPAL ANALOGY	METHOD
Structuralism					
Gestaltism					
Ecological Optics					
Constructivism					

Structuralism

Derived from philosophy of British Empiricists (e.g., Locke, Berkeley, Hume, and Mills).

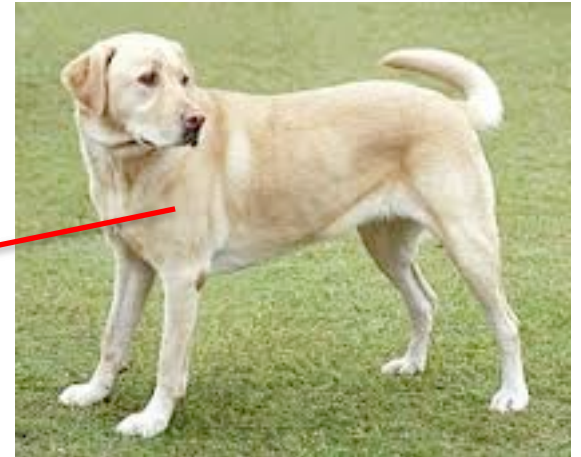
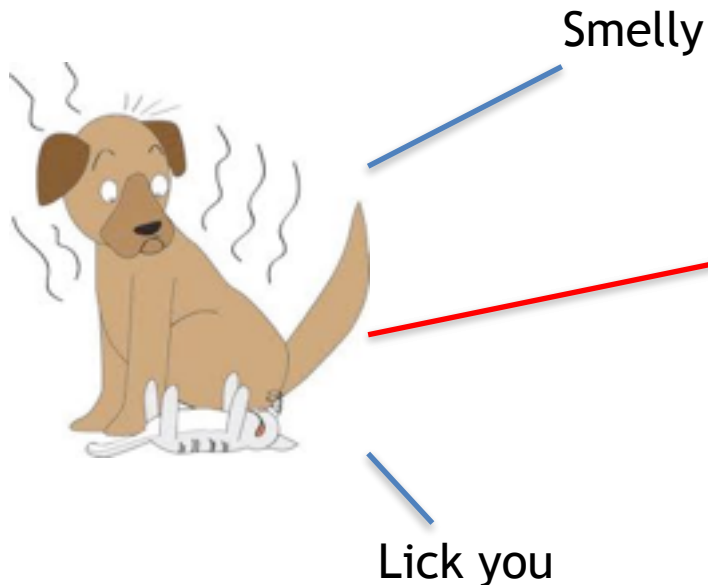
Proposed by Wilhelm Wundt, the father of modern Psychology.



Structuralism

Structuralism:

Perception results from the association of basic sensory atoms in memory via repeated, prior joint occurrences.

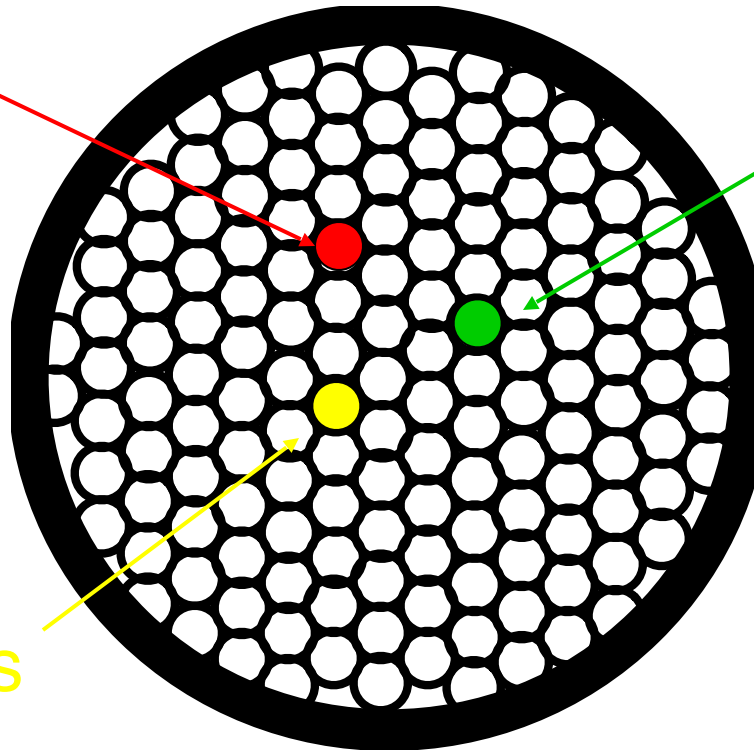


Structuralism

Sensory Atoms

Redness
at (x_1, y_1)

Greenness
at (x_3, y_3)

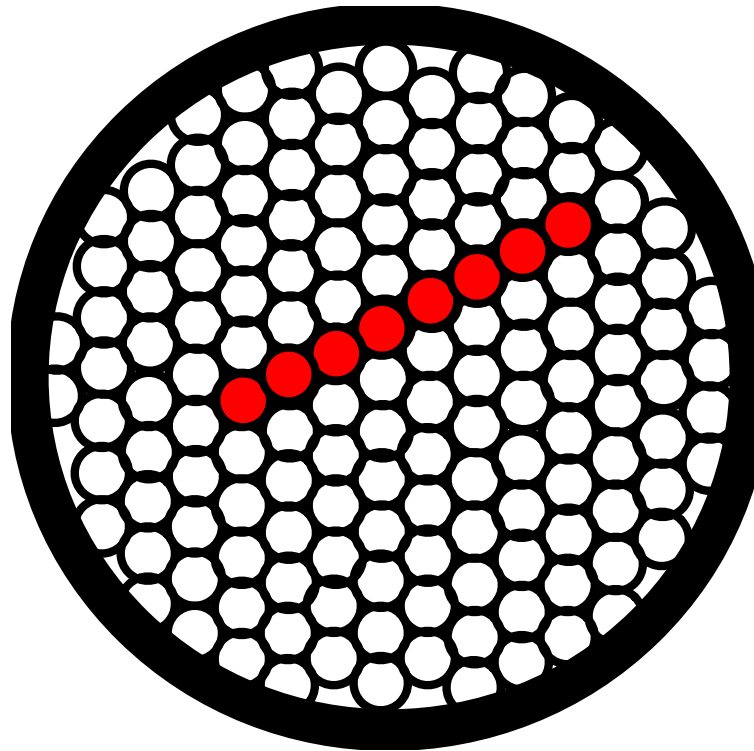


Yellowness
at (x_2, y_2)

Retinal mosaic

Structuralism

Perceptual Complexes

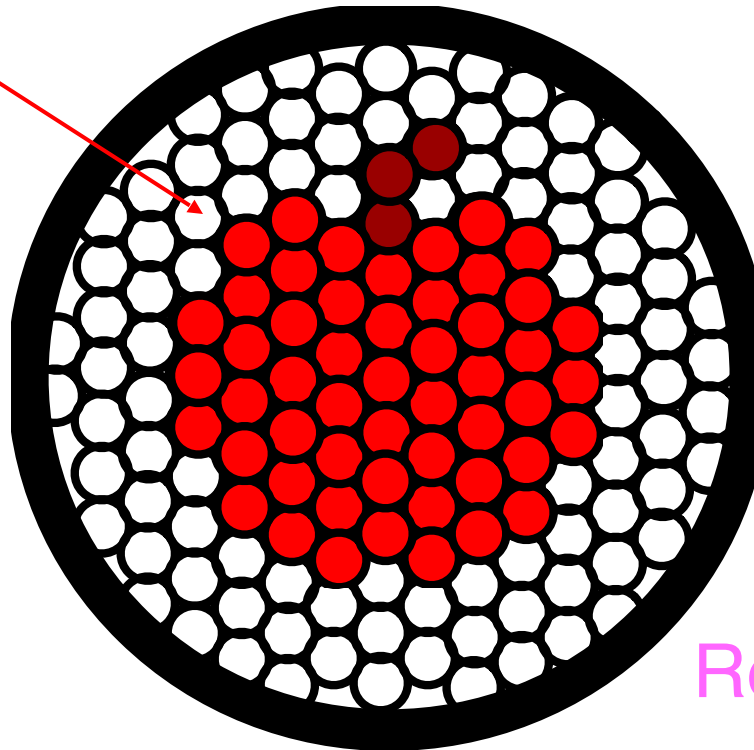


Retinal mosaic

Structuralism

Perceptual Complexes

Red apple
at (x_0, y_0)

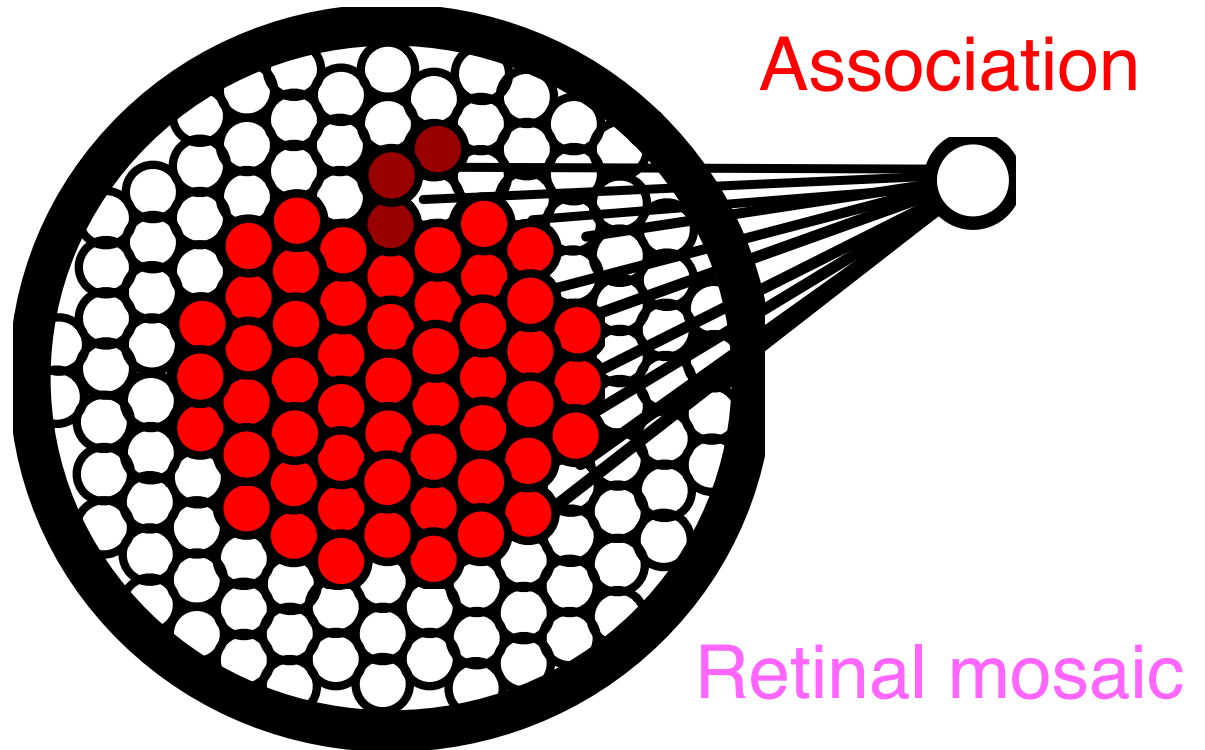


Retinal mosaic

Structuralism

Perceptual Complexes

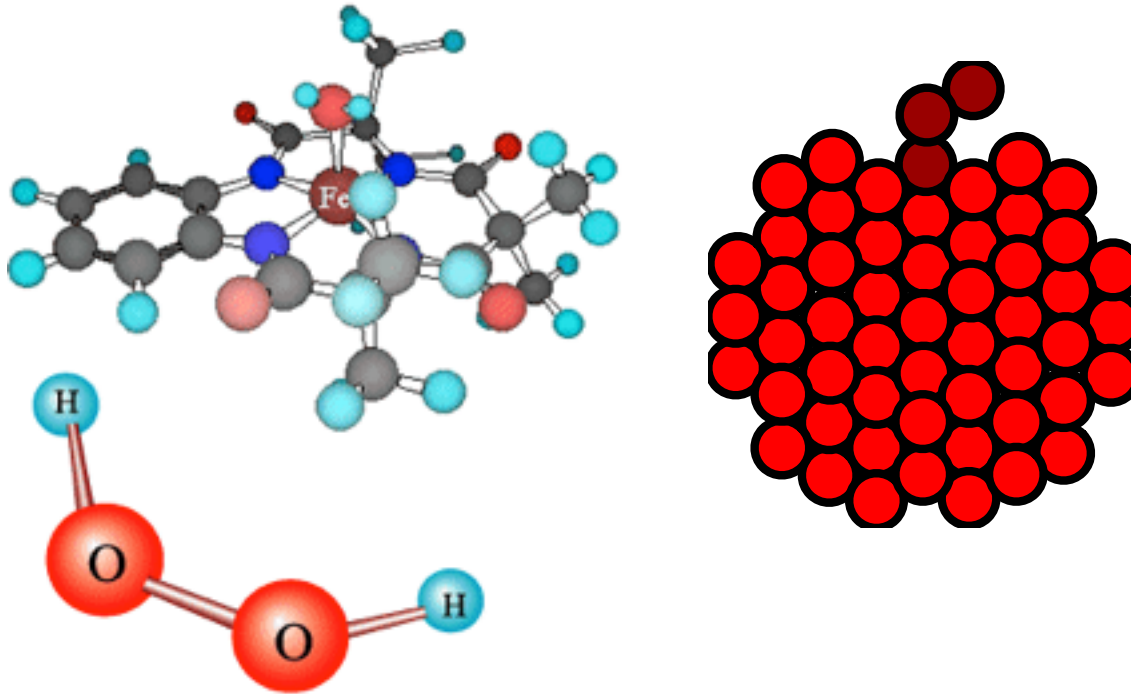
Red apple
at (x_0, y_0)



Structuralism

Chemical Analogy

Perceptions are made of basic sensory experiences
just as molecules are made of basic atoms.



Structuralism

Training Experts to discover the elementary units
of perception

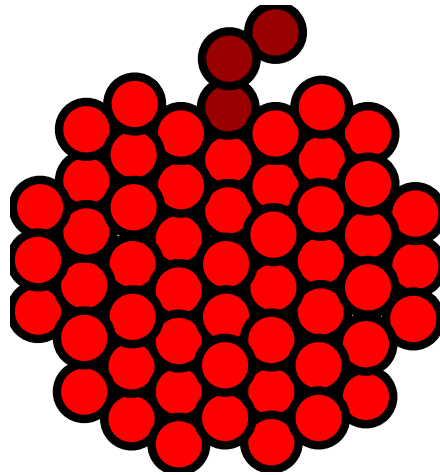
Nature of training led to different conclusions –
One of the reasons for the downfall

Structuralism

THEORY	NATIVISM vs. EMPIRICISM	ATOMISM vs. HOLISM	ORGANISM vs. ENVIRONMENT	PRINCIPAL ANALOGY	METHOD
Structuralism	Empiricism	Atomism	Organism	Chemistry	Trained Introspection
Gestaltism					
Ecological Optics					
Constructivism					

Gestaltism

*Whole is different from sum
of its parts*



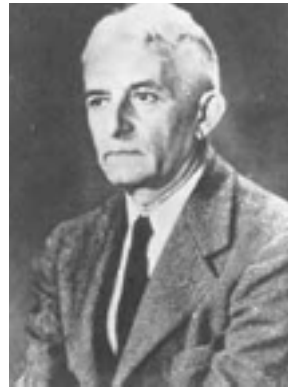
Gestaltism

Gestaltism:

Perception results from the interaction between the intrinsic structure of the stimulus and the intrinsic structure of the brain.



Max
Wertheimer



Wolfgang
Köhler



Kurt
Koffka

Gestaltism

The Gestalt movement in perceptual theory was primarily a reaction against Structuralism:

Rejected atomism

Rejected empiricism

Rejected associationism

Successful in arguing against Structuralism, but less successful in promoting its own theoretical agenda.

Gestaltism

Principles of Gestalt Theory

Holism:

The whole is different from the sum of its parts.

Prägnanz:

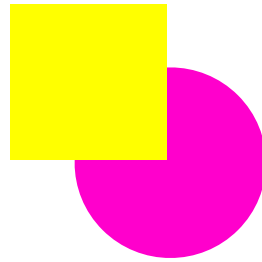
The percept will be as “good” as the prevailing conditions allow, I.e. simplest explanation

Nativism:

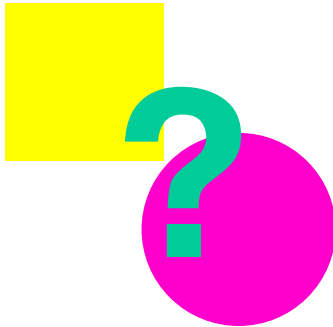
Not a total rejection of learning, but rejection of its primacy.

Gestaltism

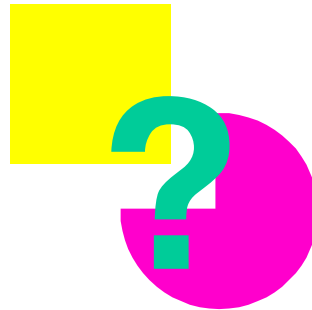
Prägnanz: the percept will be as
“good” as the prevailing conditions allow



What is this?



square &
circle?



square &
pacman?



squigit &
pacman?

etc....

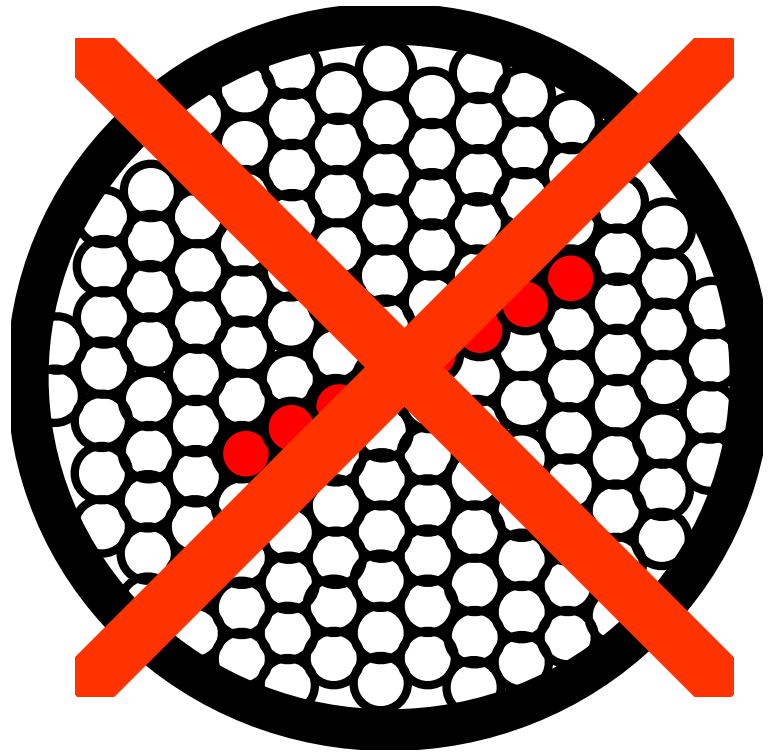
Gestaltism

Holism: The whole is different from the sum of its parts.

Emergent properties:

Features of a configuration that are not features of its components, e.g.:

- length
- orientation
- curvature
- closure
- connectedness



Emergent Properties



The Problem of Organization

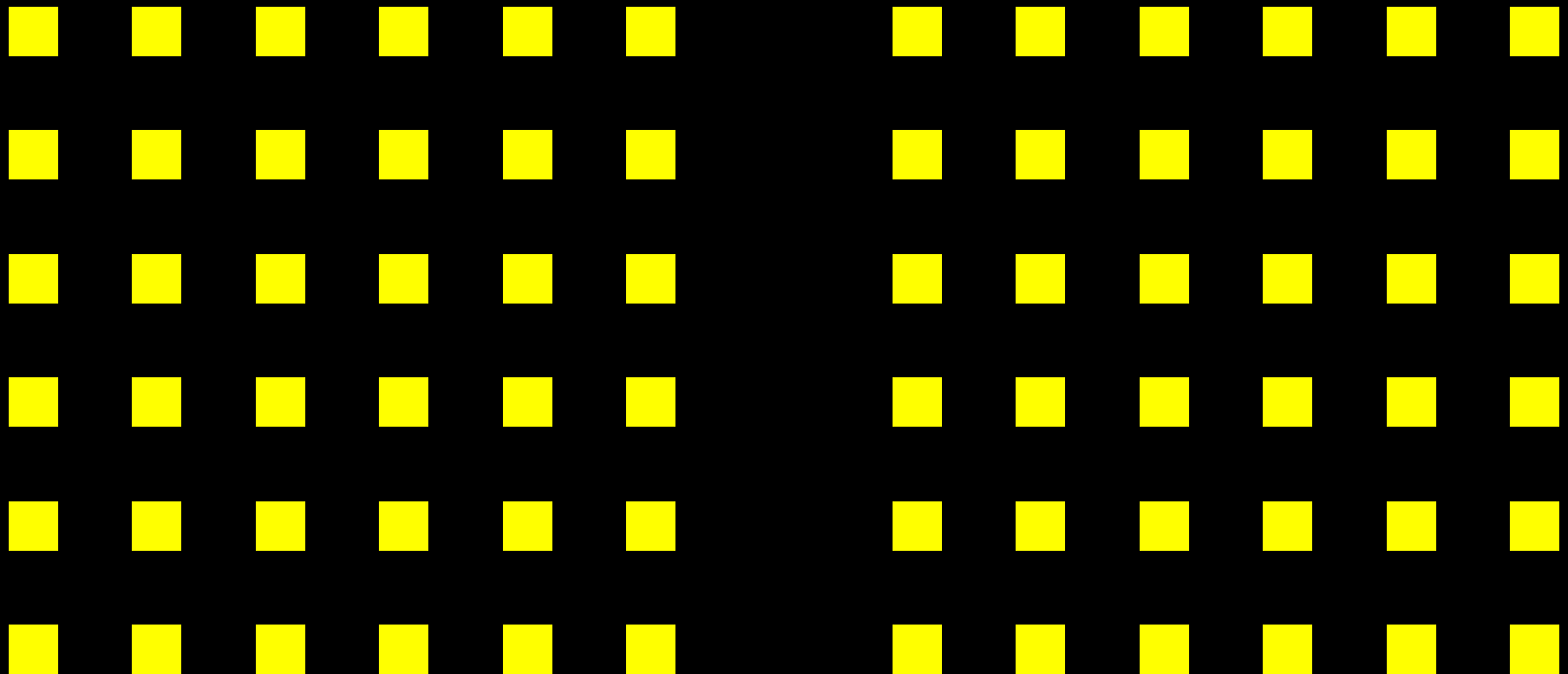
The problem of grouping:

“I stand at the window and see a house, trees, sky.”
Theoretically, I might say that there were 327 brightnesses and nuances of color. Do I have “327”? No. I have sky, house, and trees. (Wertheimer, 1923)



Perceptual Grouping

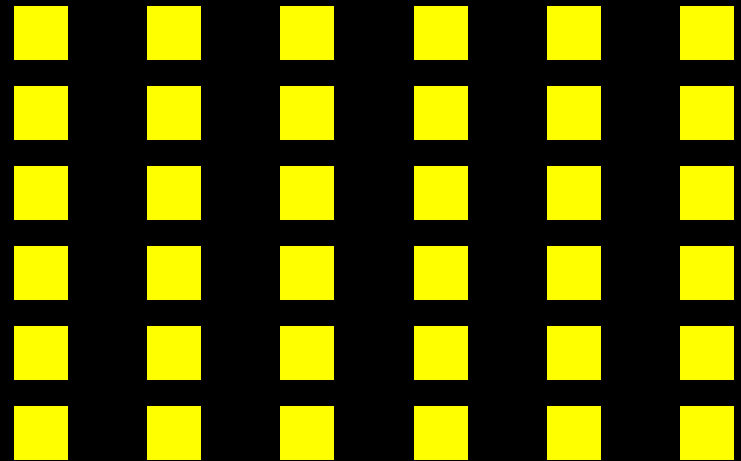
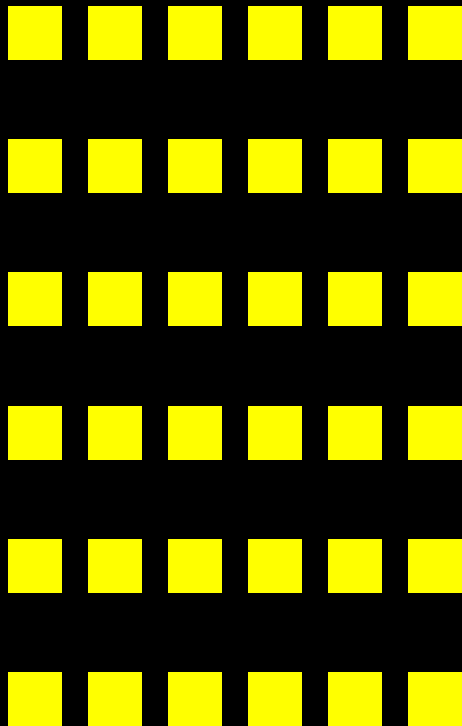
Wertheimer's 'laws' of grouping



Perceptual Grouping

14.20

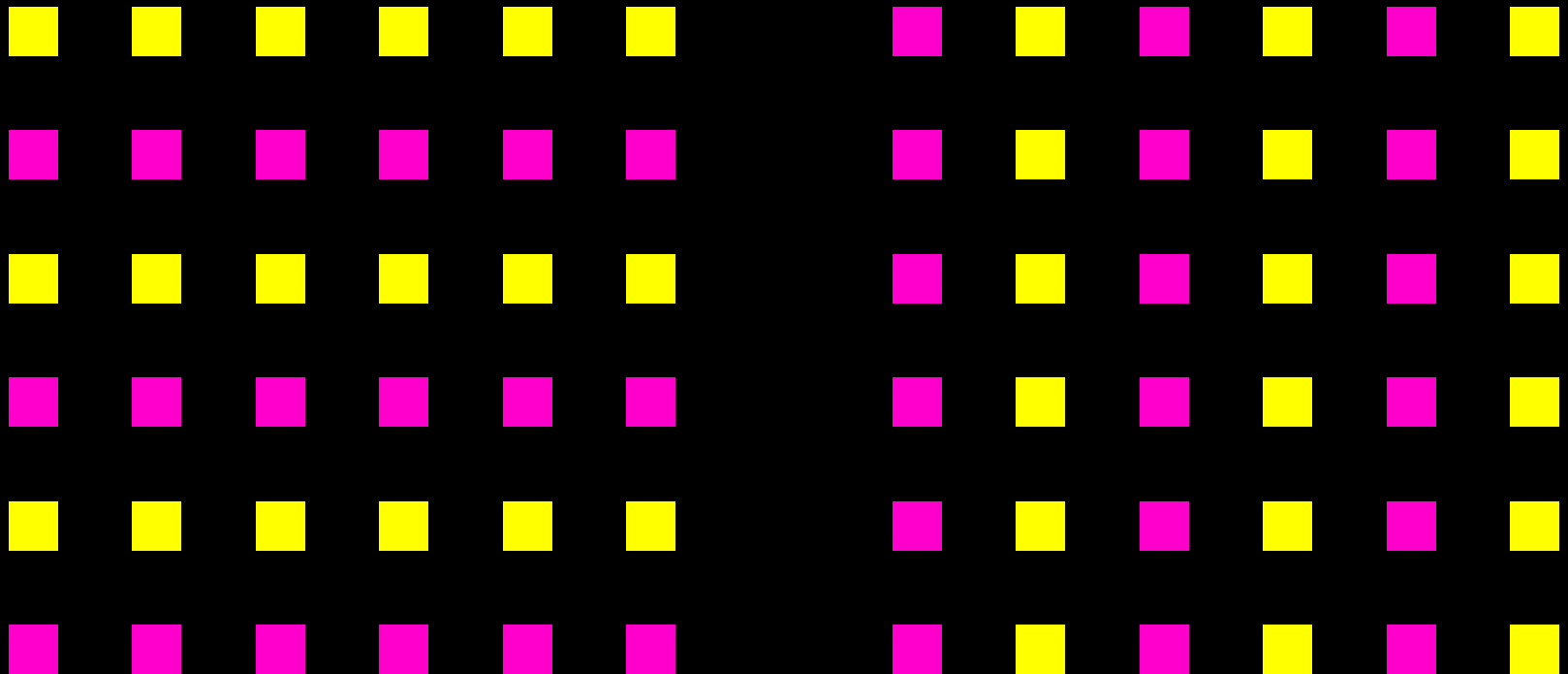
Proximity



Perceptual Grouping

14.21

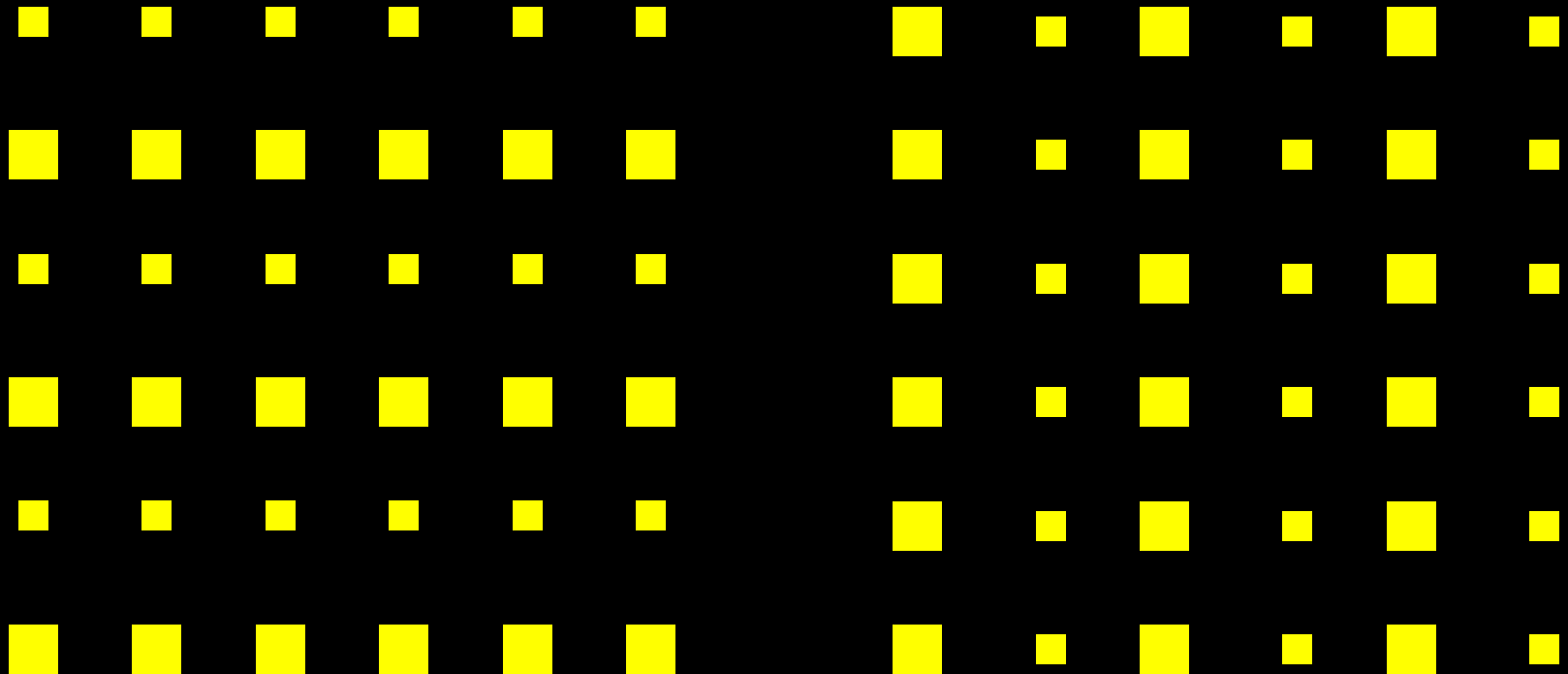
Color Similarity



Perceptual Grouping

14.22

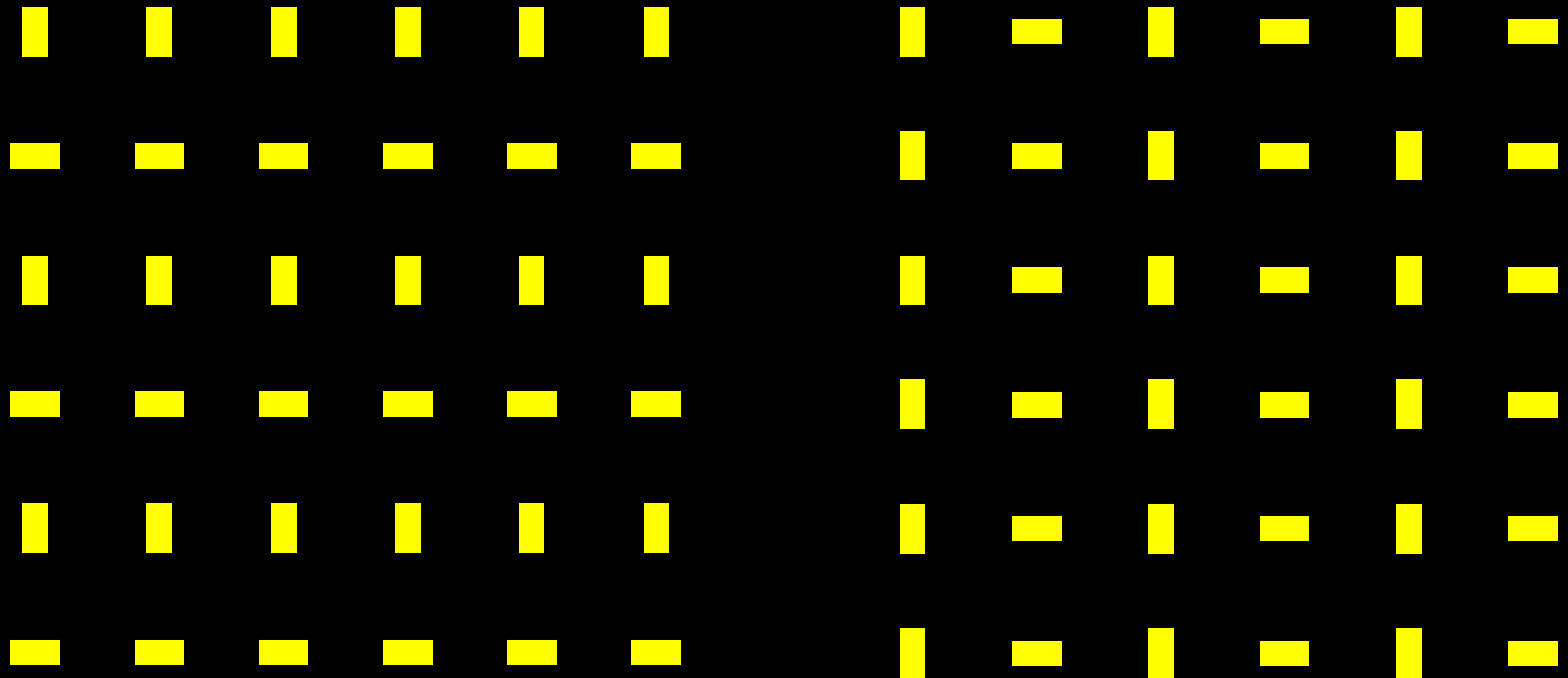
Size Similarity



Perceptual Grouping

14.23

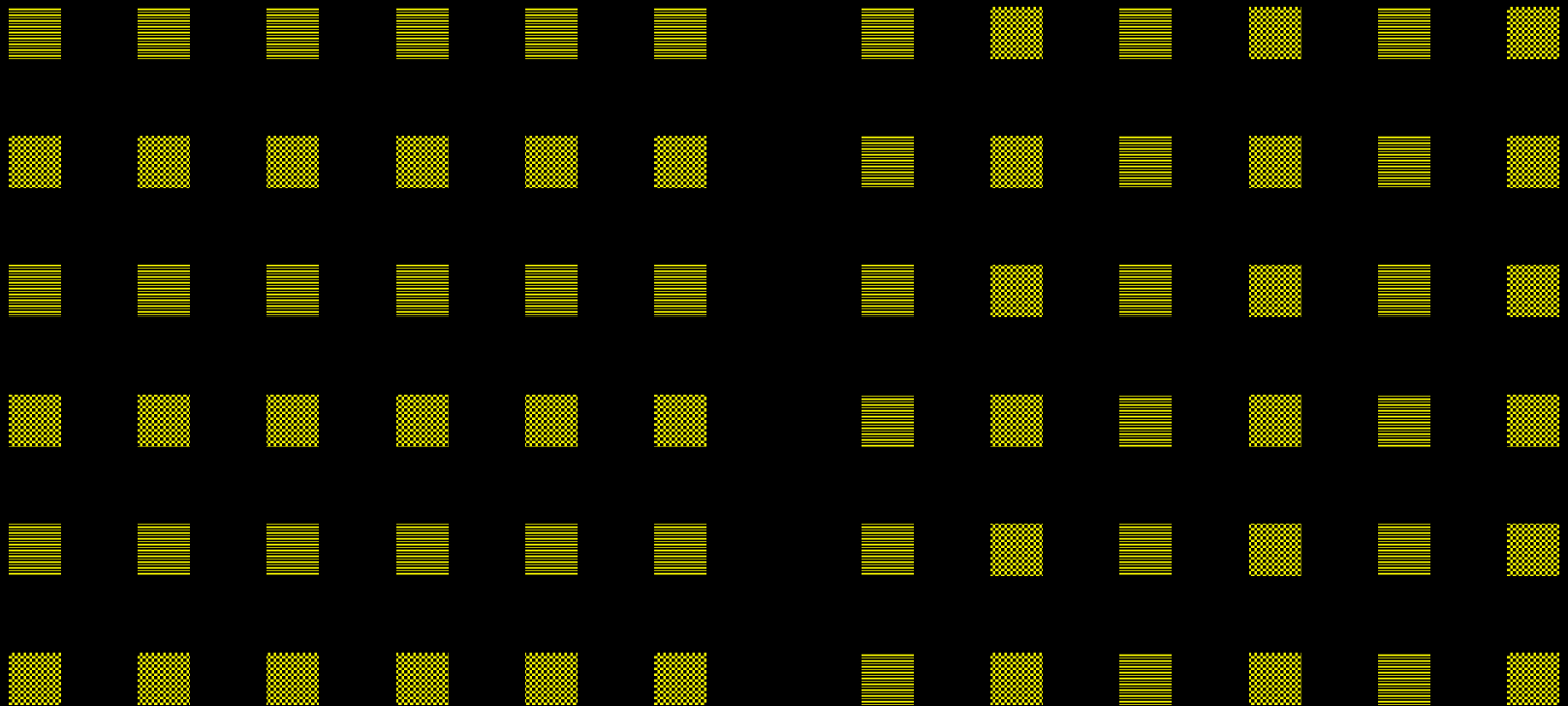
Orientation Similarity



Perceptual Grouping

14.24

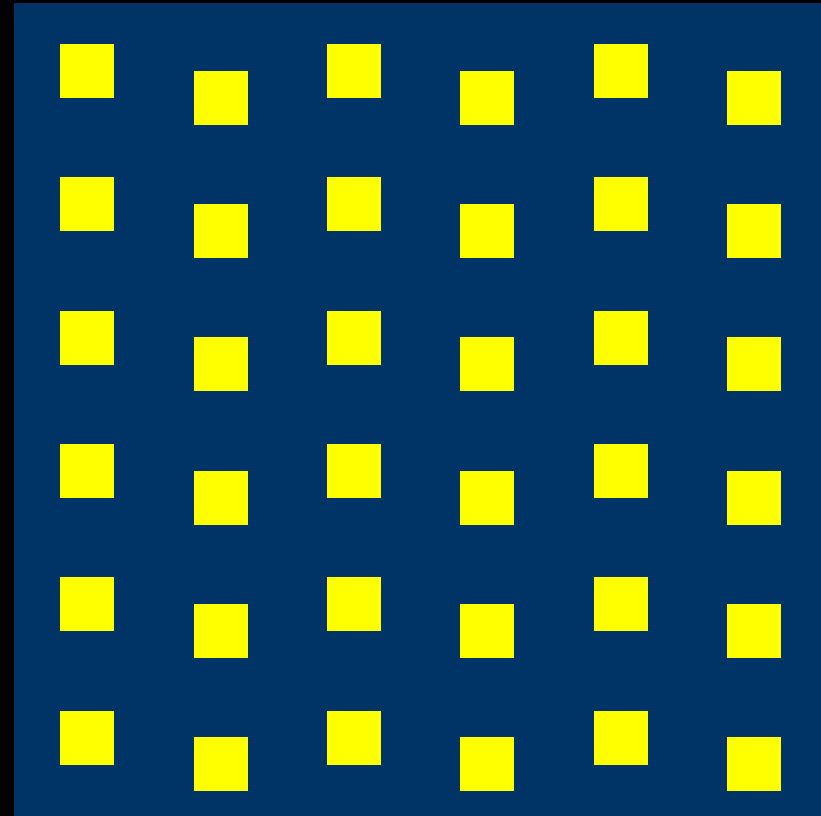
Similarity of texture



Perceptual Grouping

14.26

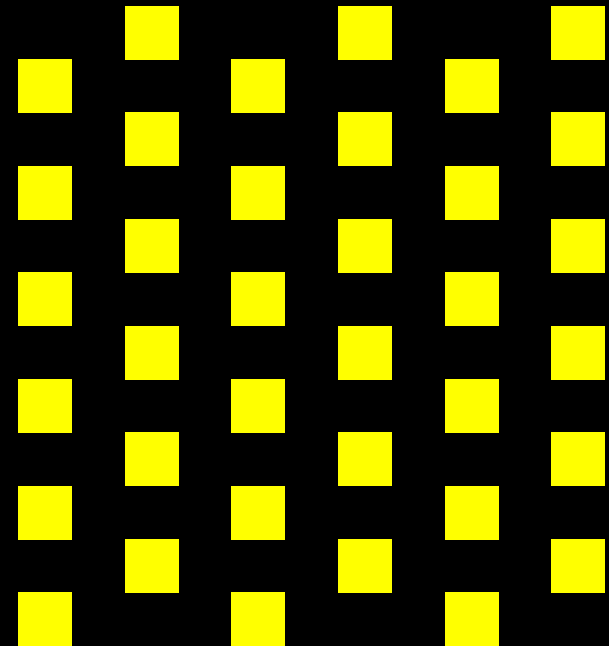
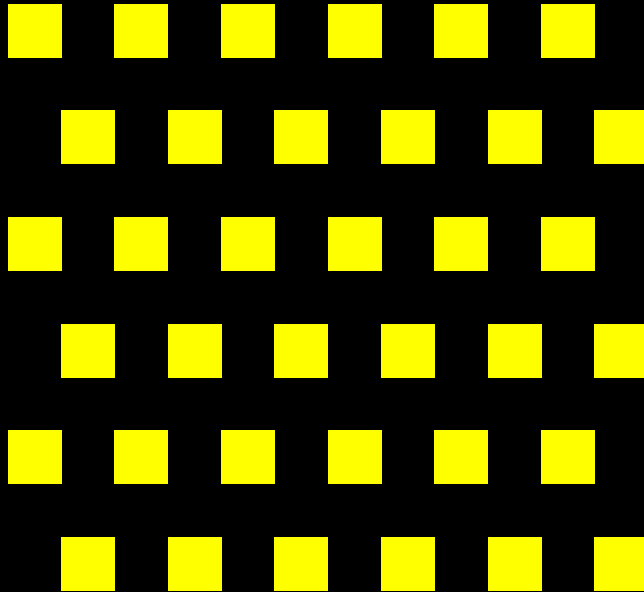
Common Fate



Perceptual Grouping

14.27

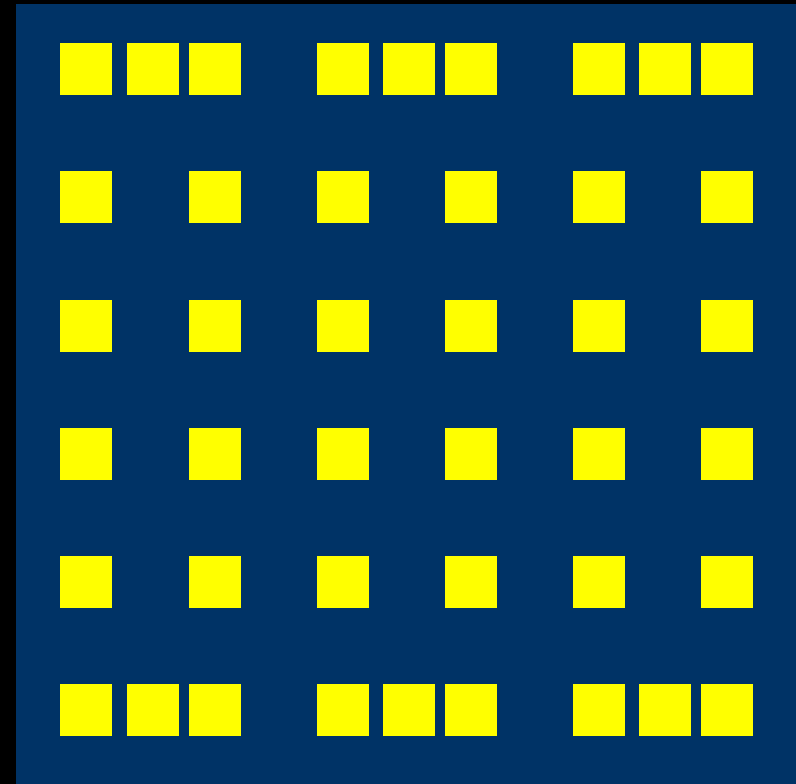
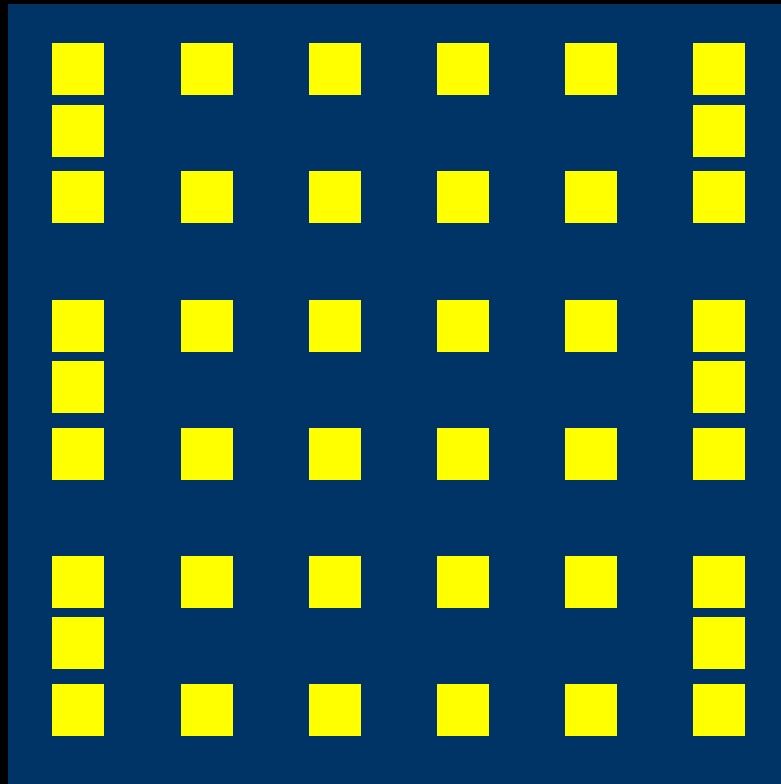
Good Continuation



Perceptual Grouping

14.28

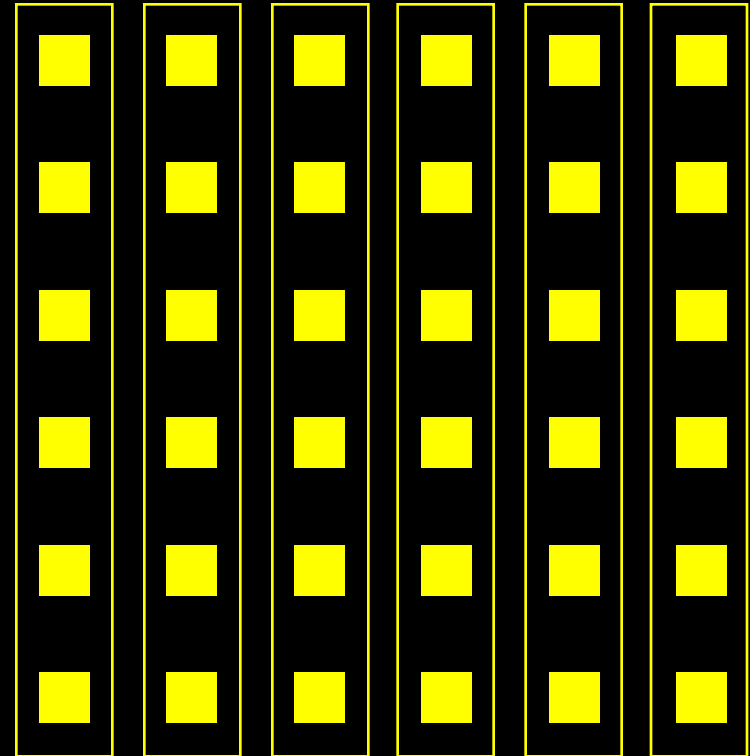
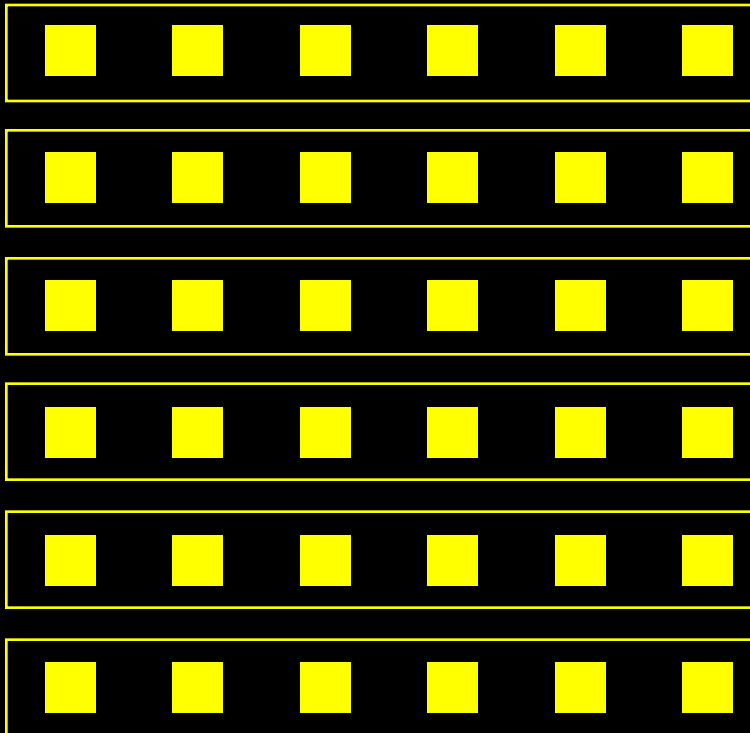
Closure



Perceptual Grouping

14.29

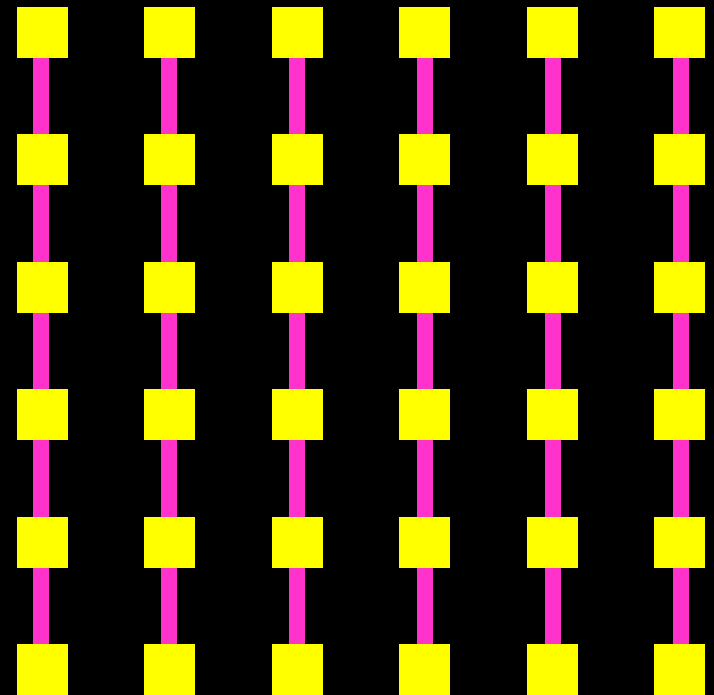
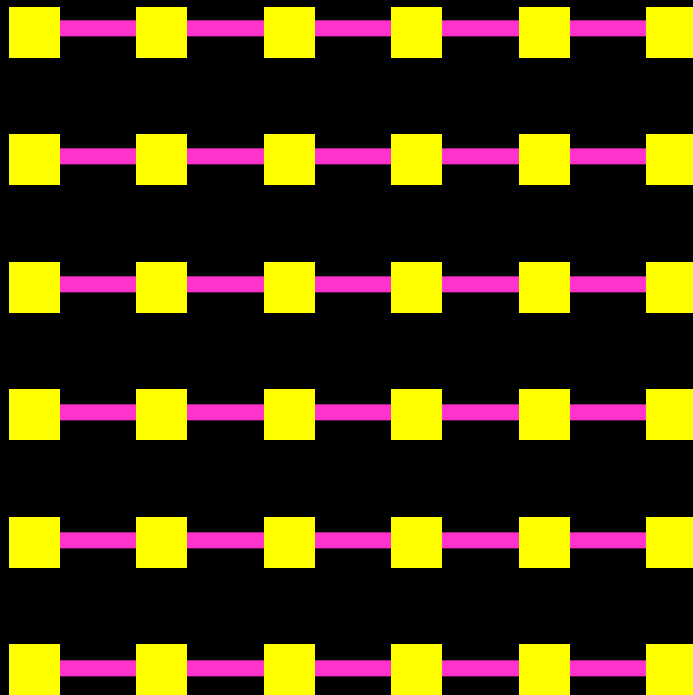
Common Region



Perceptual Grouping

14.30

Element Connectedness



Perceptual Grouping

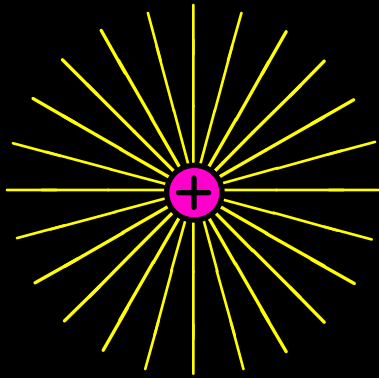
14.32

Past Experience

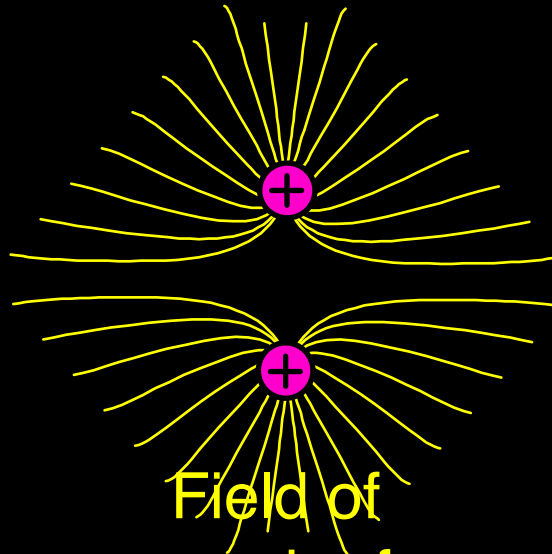


Gestaltism

Field Theoretic Analogy



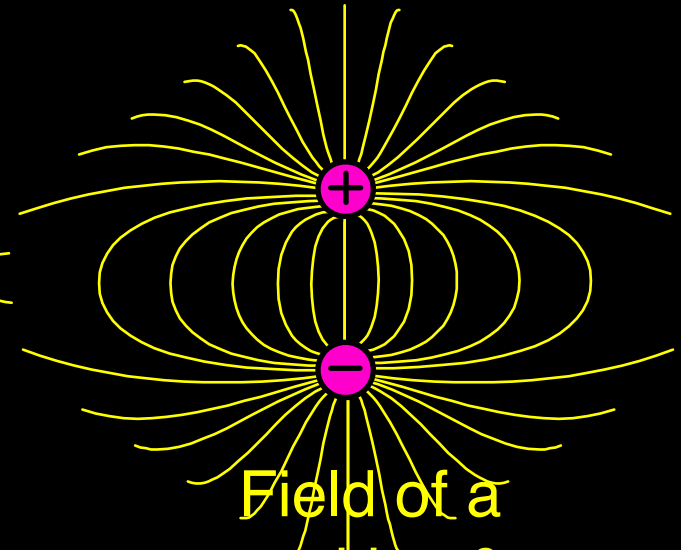
Field of
a single
positive
charge



Field of
a pair of
positive
charges



Repulsion



Field of a
positive &
a negative
charge



Attraction

Gestaltism

THEORY	NATIVISM vs. EMPIRICISM	ATOMISM vs. HOLISM	ORGANISM vs. ENVIRONMENT	PRINCIPAL ANALOGY	METHOD
Structuralism	Empiricism	Atomism	Organism	Chemistry	Trained Introspection
Gestaltism	Nativism	Holism	Organism	EM Fields	Naïve Introspection
Ecological Optics					
Constructivism					

Ecological Optics

Ecological Optics: Perception is the direct apprehension of the visible environment by extracting invariants in the dynamic ambient optic array and the affordances of objects.

James J. Gibson: Ask not what's inside your head, but what your head's inside of!
(Ecological optics is a theory of stimulus structure.)



How does the world structure the light stimulus received by the brain?

Ecological Optics

Texture Gradients are gradual changes in the size & shape of texture elements in an image when a plane recedes in depth.



Tile
floor



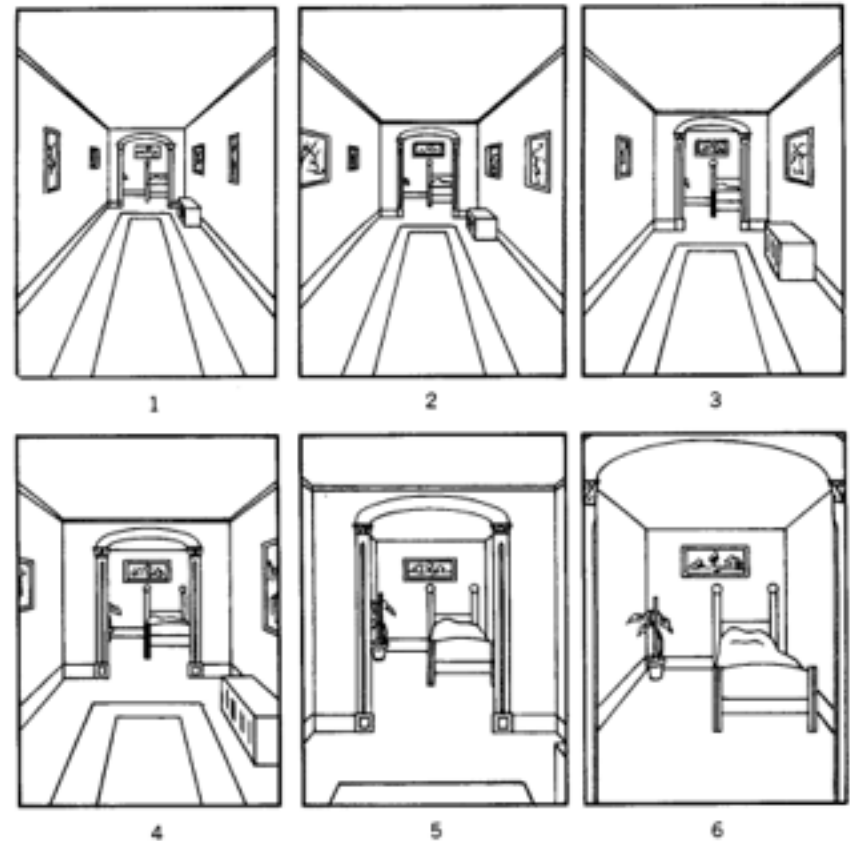
Wheat
field



Geometric
surface

Ecological Optics

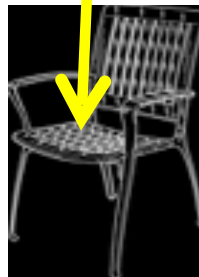
Perception of the 3-D world is NOT ambiguous for an actively exploring organism.



Direct Perception: Perception is not mediated by processes like unconscious thinking

- **Affordances:** Opportunities of interaction in the scene.
- Infer functions rather than identities.

Sittable



Ecological

THEORY	NATIVISM vs. EMPIRICISM	ATOMISM vs. HOLISM	ORGANISM vs. ENVIRONMENT	PRINCIPAL ANALOGY	METHOD
Structuralism	Empiricism	Atomism	Organism	Chemistry	Trained Introspection
Gestaltism	Nativism	Holism	Organism	EM Fields	Naïve Introspection
Ecological Optics	Nativism	Holism	Environment	Resonance	Ecological Analysis
Constructivism					

Constructivism

Constructivism: Perception is the result of unconscious inferences about the scene most likely to have caused the retinal image or event.

Hermann von Helmholtz
originated the idea of
unconscious inference and
the likelihood principle.

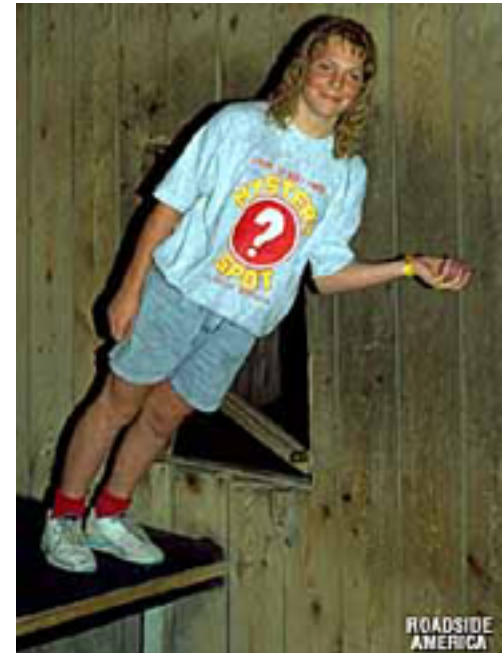


Constructivism

Unconscious Inference: the process of recovering environmental information by logically combining retinal information with heuristic assumptions.

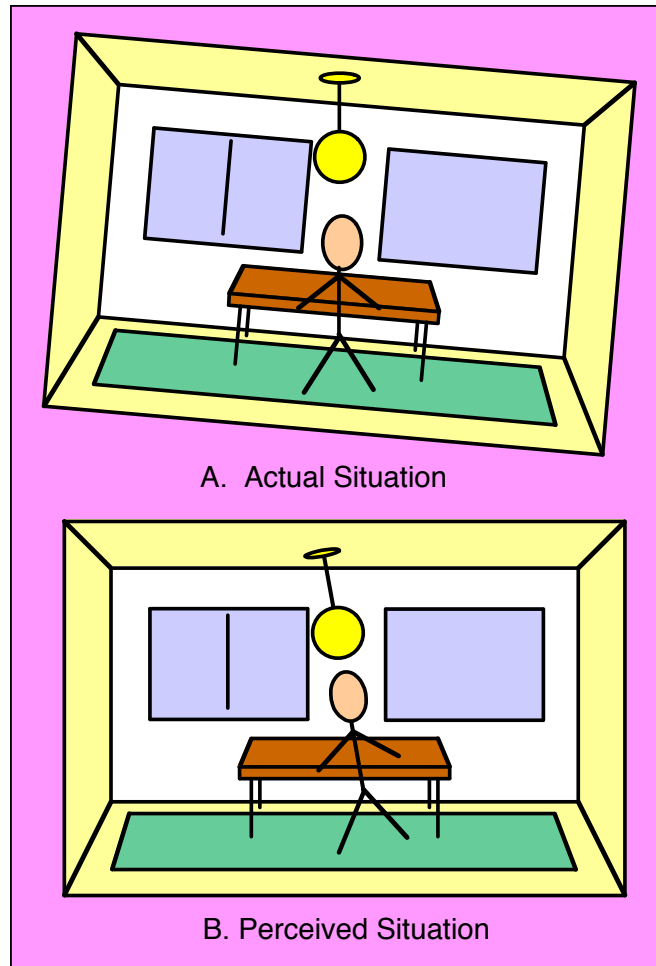
Tilted room illusion:

If you assume that the walls and floor of the room are vertical and horizontal, then you must be tilted —and you feel that way!



Constructivism

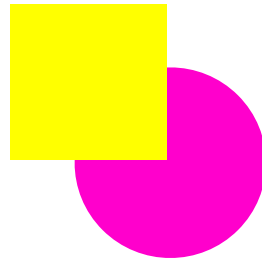
Tilted room illusion



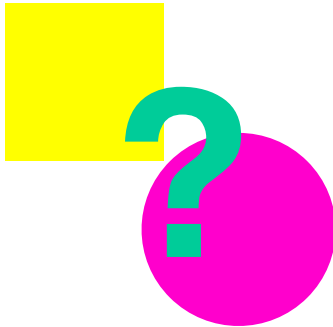
Constructivism

Prägnanz vs. Likelihood

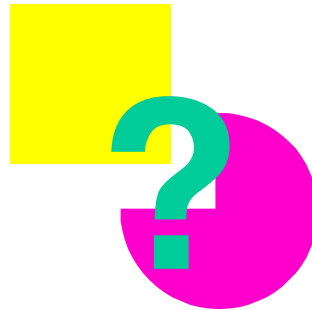
What governs what we see: goodness or probability?



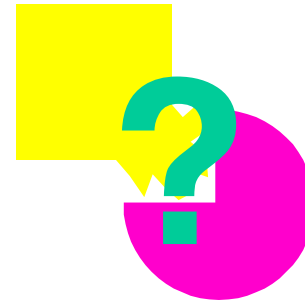
What is this?



square &
circle?



square &
pacman?



squigit &
pacman?

etc....

Constructivism

THEORY	NATIVISM vs. EMPIRICISM	ATOMISM vs. HOLISM	ORGANISM vs. ENVIRONMENT	PRINCIPAL ANALOGY	METHOD
Structuralism	Empiricism	Atomism	Organism	Chemistry	Trained Introspection
Gestaltism	Nativism	Holism	Organism	EM Fields	Naïve Introspection
Ecological Optics	Nativism	Holism	Environment	Resonance	Ecological Analysis
Constructivism	both	both	both	Logical Inference	Behavior

So which theory is correct?

Probably none of them!

Or maybe all of them, to some degree!

It's good to keep them in mind, when
designing your algorithms!

**“You can’t play 20-questions
with nature and win!”**
(Allen Newell, 1973)



Recap-Theories of Human Vision

- Importance of theories
- Issues to answer in creating a theory for vision
 - Empiricism vs. Nativism
 - Organism vs. Environment
 - Atomism vs. Holism
- Theories
 - Structuralism
 - Gestaltism
 - Ecology
 - Constructivism

Enter the
computers

MASSACHUSETTS INSTITUTE OF TECHNOLOGY
PROJECT MAC

Artificial Intelligence Group
Vision Memo. No. 100.

July 7, 1966

THE SUMMER VISION PROJECT

Seymour Papert.

The summer vision project is an attempt to use our summer workers effectively in the construction of a significant part of a visual system. The particular task was chosen partly because it can be segmented into sub-problems which will allow individuals to work independently and yet participate in the construction of a system complex enough to be a real landmark in the development of "pattern recognition".

Computational Thinking

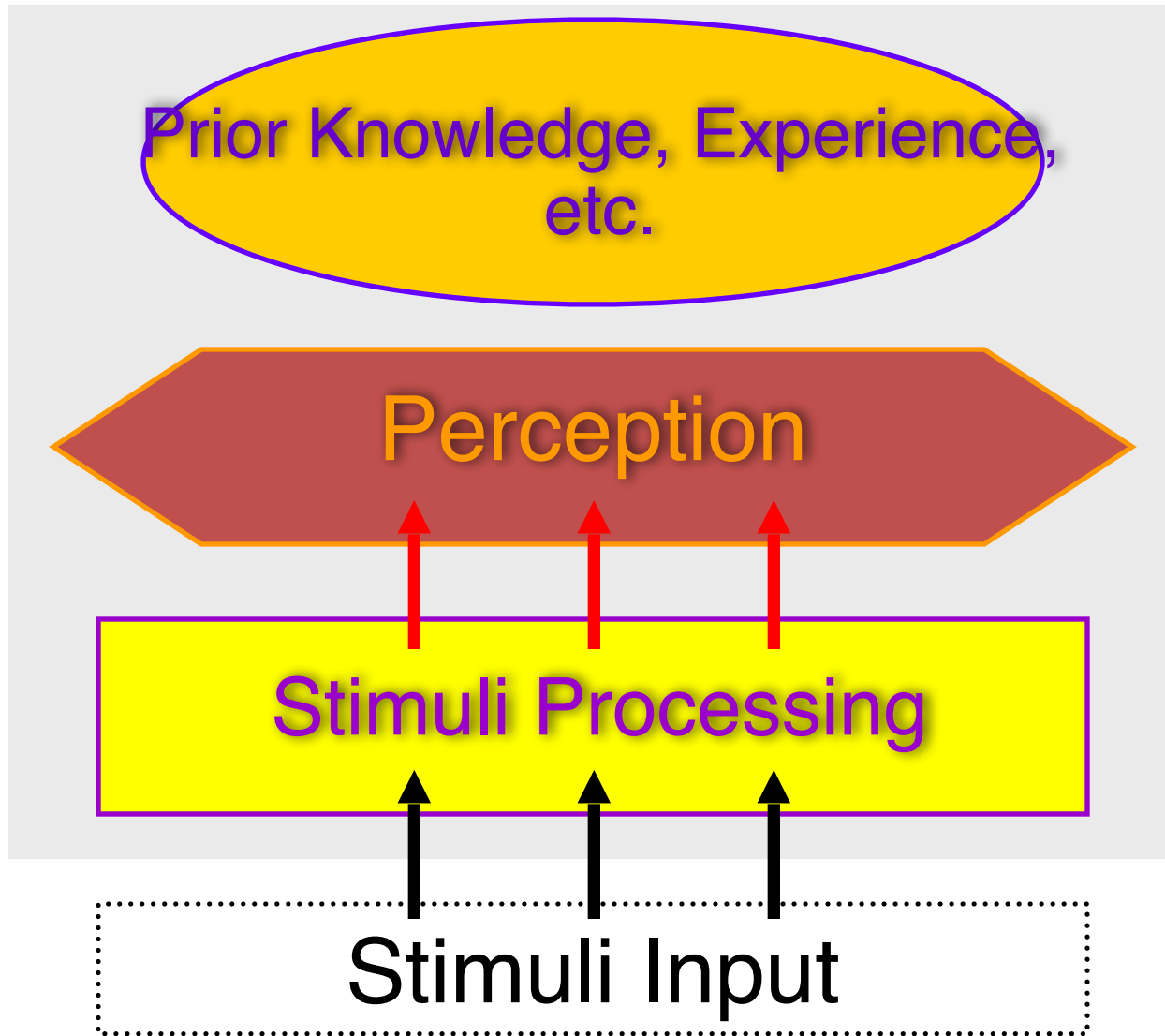
- + Real Images rather than made up Stimuli
- + No more vague theories. Needed more details.
- + Principles of Information Processing

Top-down vs. Bottom-up

Top-Down versus Bottom-Up Perception

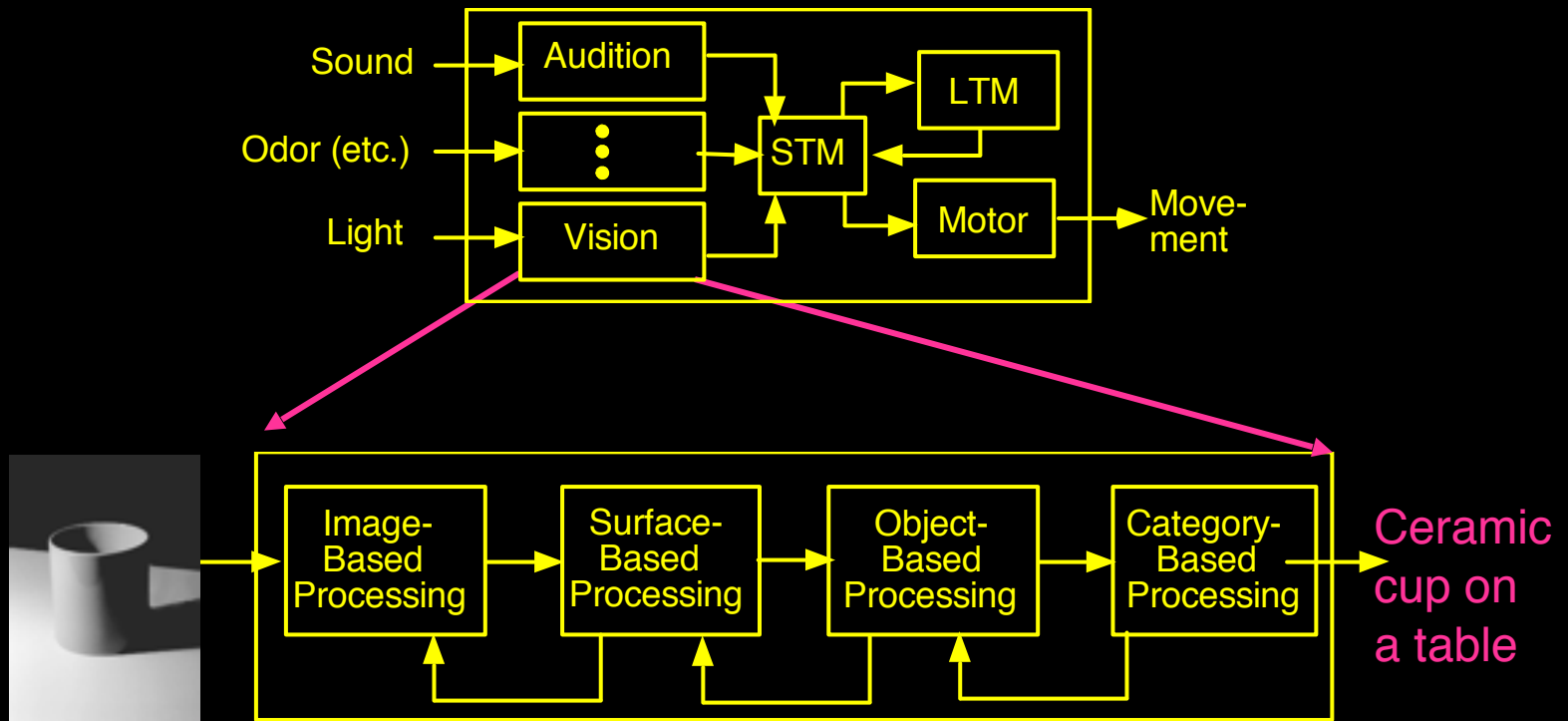
- Top-Down
 - Perceive the whole and then individual parts as needed.
 - Experience-driven as opposed to stimulus or input-data driven.
 - Quick and highly inferential but also a source of misperception.
- Bottom-up
 - Perceive the individual parts and organize them into a whole, if possible.
 - Information available in the stimulus itself.

Bottom-Up Processing



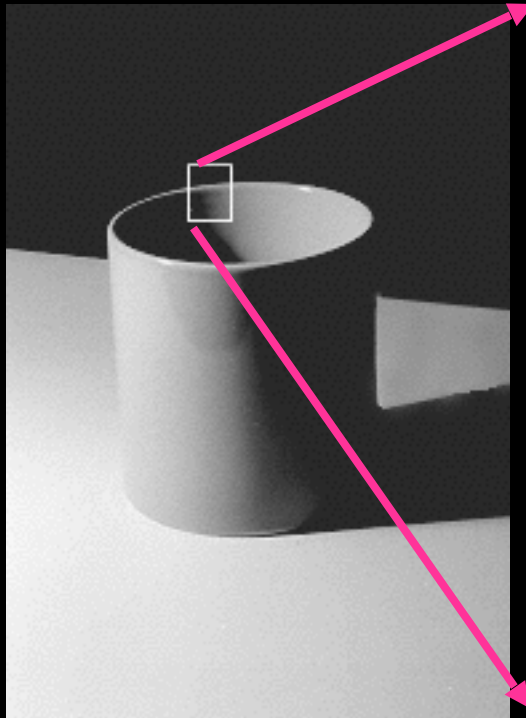
David Marr's Theory

Four Stages of Visual Perception

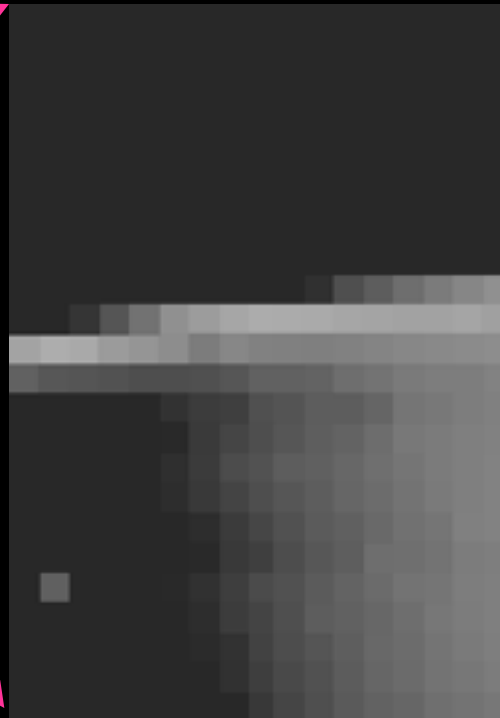


Four Stages of Visual Perception

The Retinal Image



An Image



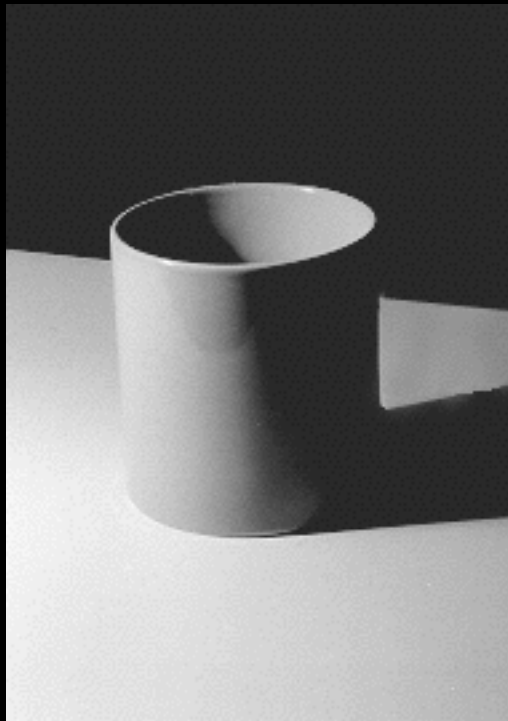
(blowup)



Receptor Output

Four Stages of Visual Perception

Retinal Image



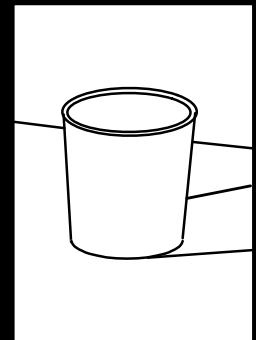
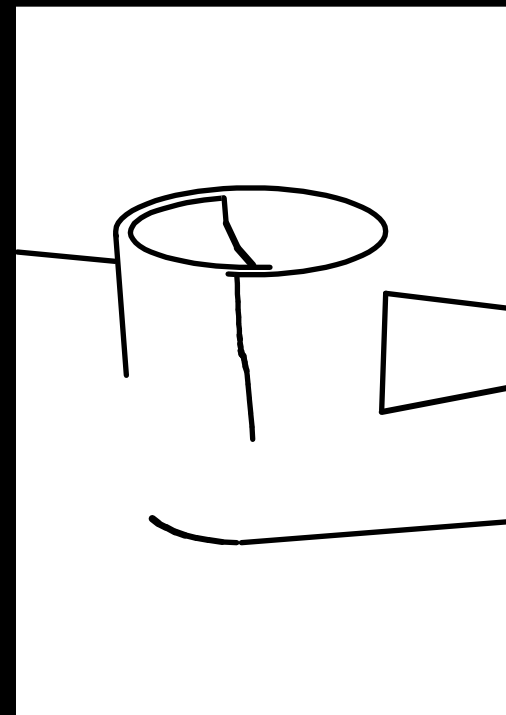
An Image

Image-based processes



Edges
Lines
Blobs
etc.

Image-based Representation

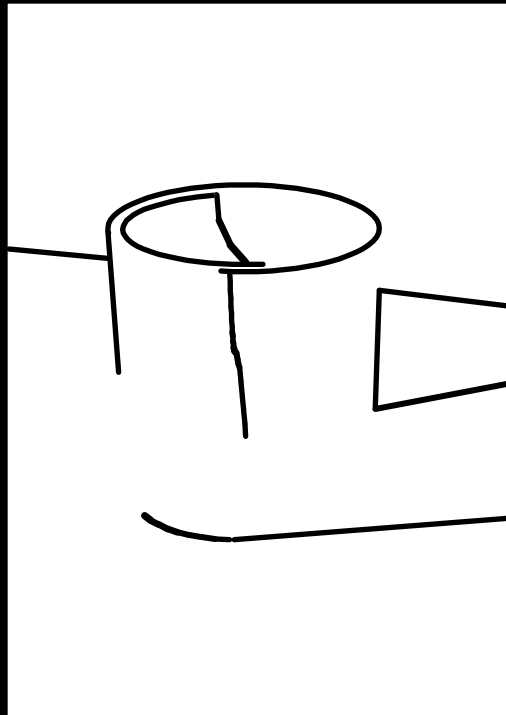


(Line Drawing)

Primal Sketch
(Marr)

Four Stages of Visual Perception

Image-based
Representation



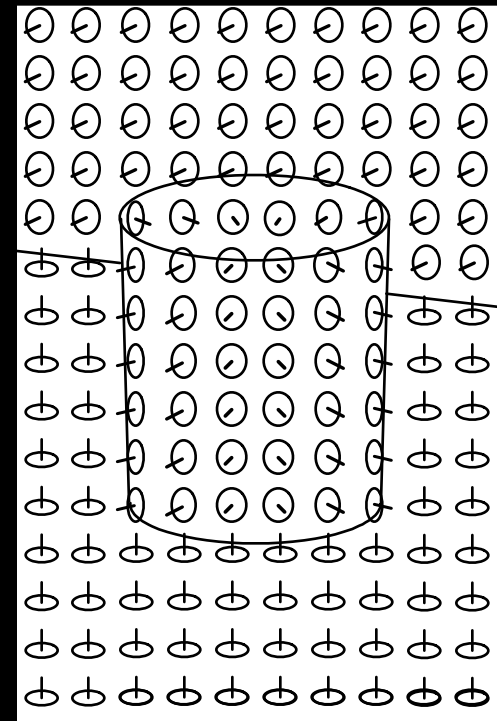
Primal Sketch

Surface-
based
processes



Stereo
Shading
Motion
etc.

Surface-based
Representation



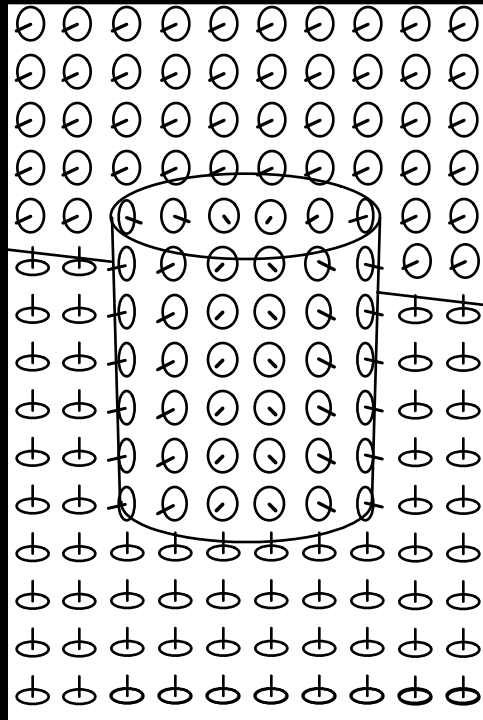
2.5-D Sketch

Surfaces etc.

- **Initially proposed by Gibson (1950).**
- **Departure from Object-based 3-D representations.**
- **But what is this 2.5D and how it is inferred (especially in 2D images).**

Four Stages of Visual Perception

Surface-based
Representation



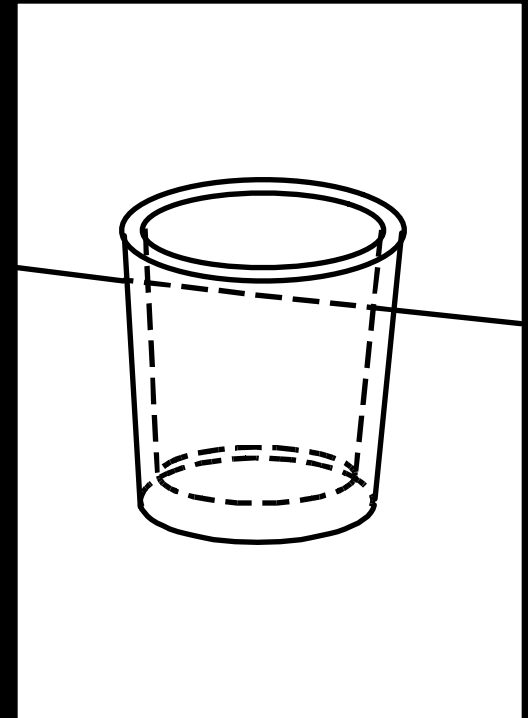
2.5-D Sketch

Object-
based
processes



Grouping
Parsing
Completion
etc.

Object-based
Representation



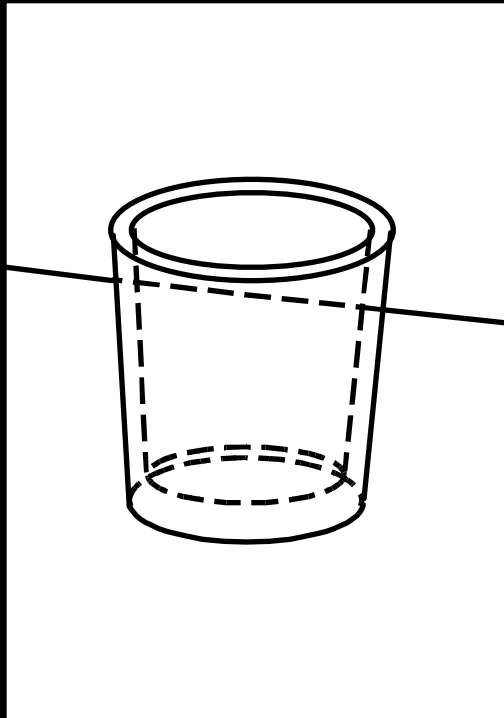
Volumetric Sketch

Object-based

- **Move from surfaces to volumes.**
- **Seeing the hidden layer.**
- **But how do you move from surface to volumes? Bottom-up or Top-down?**

Four Stages of Visual Perception

Object-based
Representation



Volumetric Sketch

Category-
based
processes



Pattern-
Recognition

Spatial-
description

Category-based
Representation

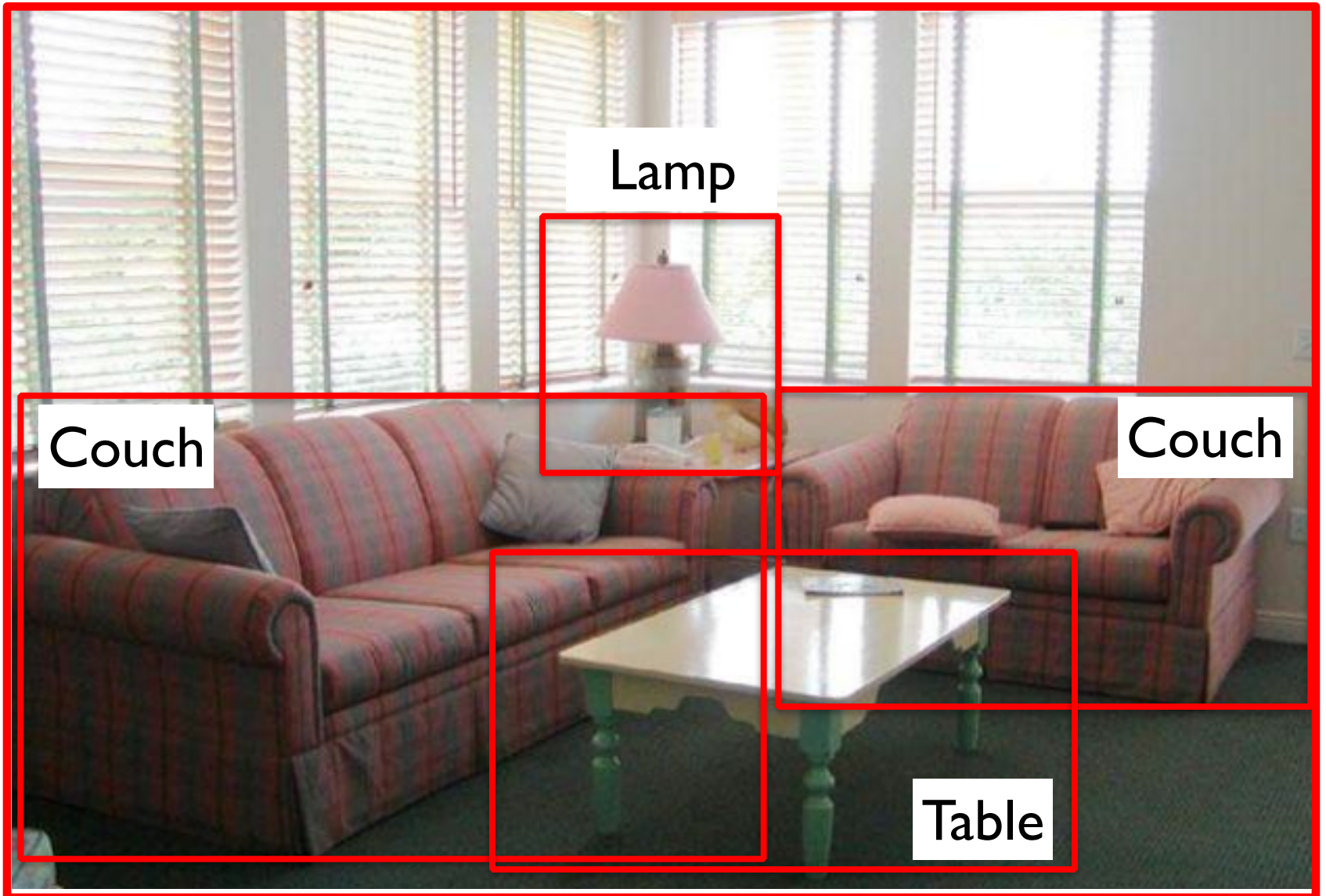
Category: cup
Color: light-gray
Size: 6"
Location: table

Basic-level Category

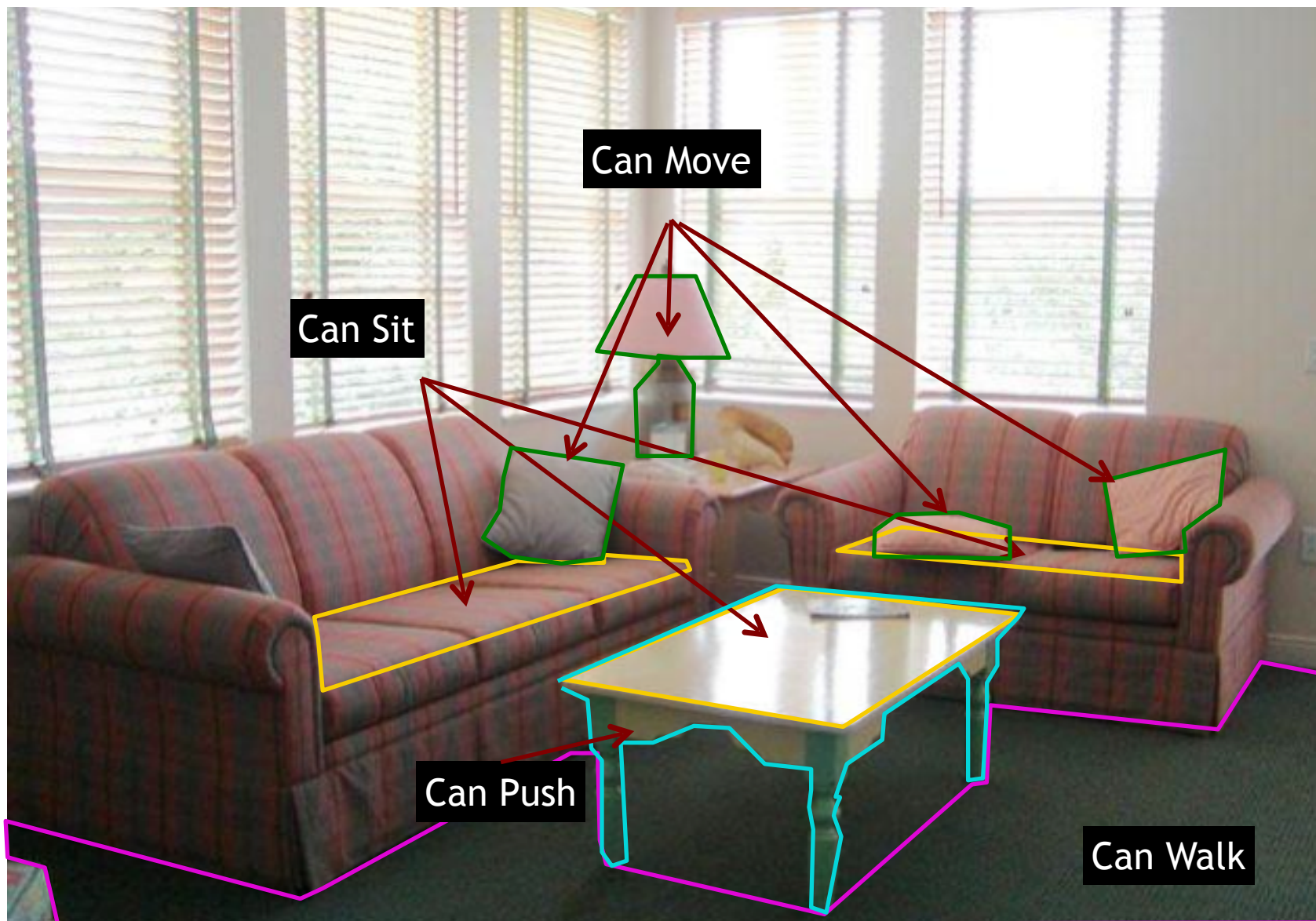
Category-based

- **Move from volumes to categories.**
- **Essentially this step relates to association of object with memory elements.**
- **But what are the right categories?**

Semantic?

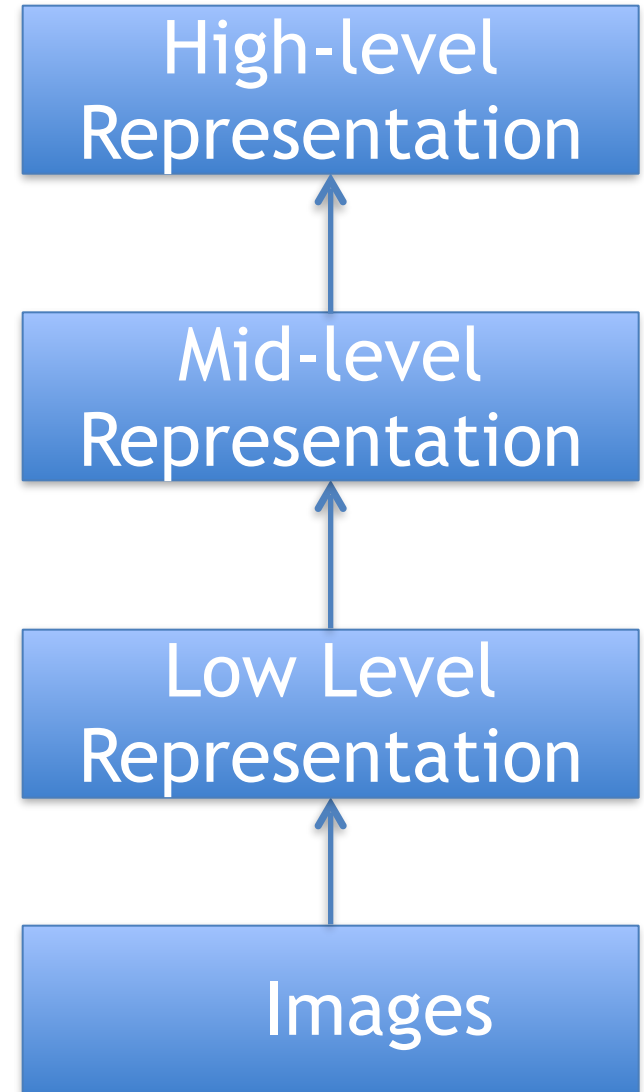
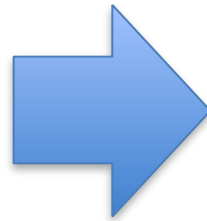
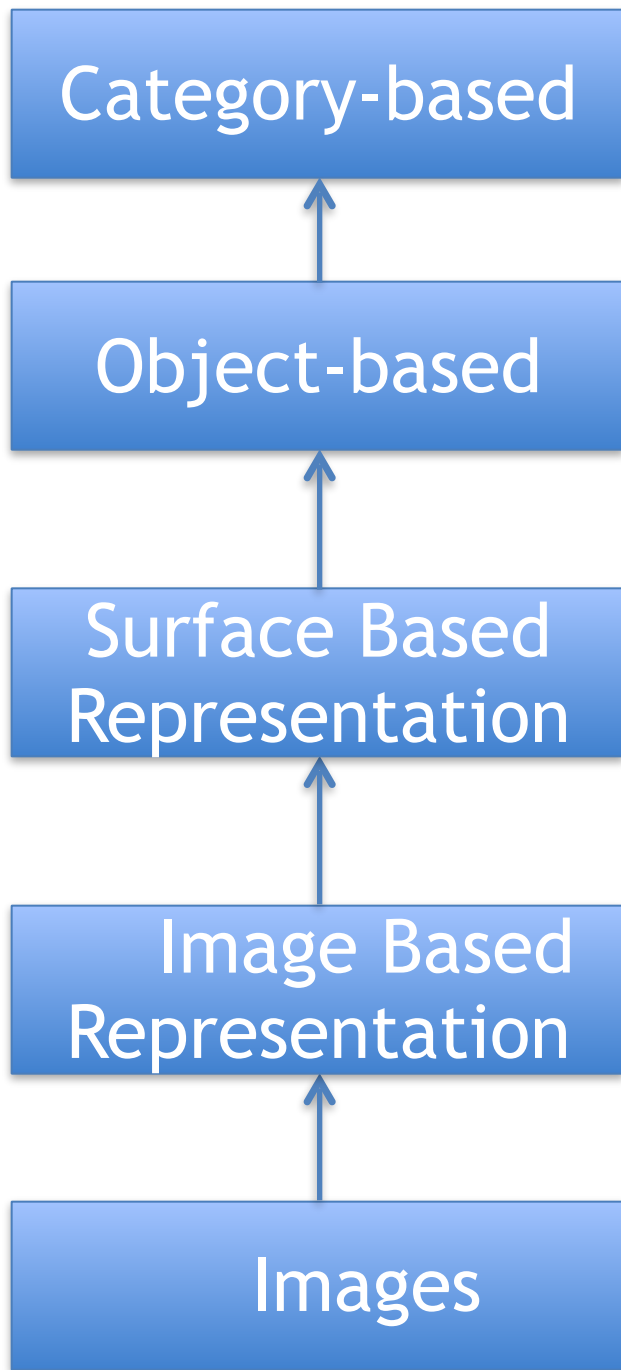


Functional?

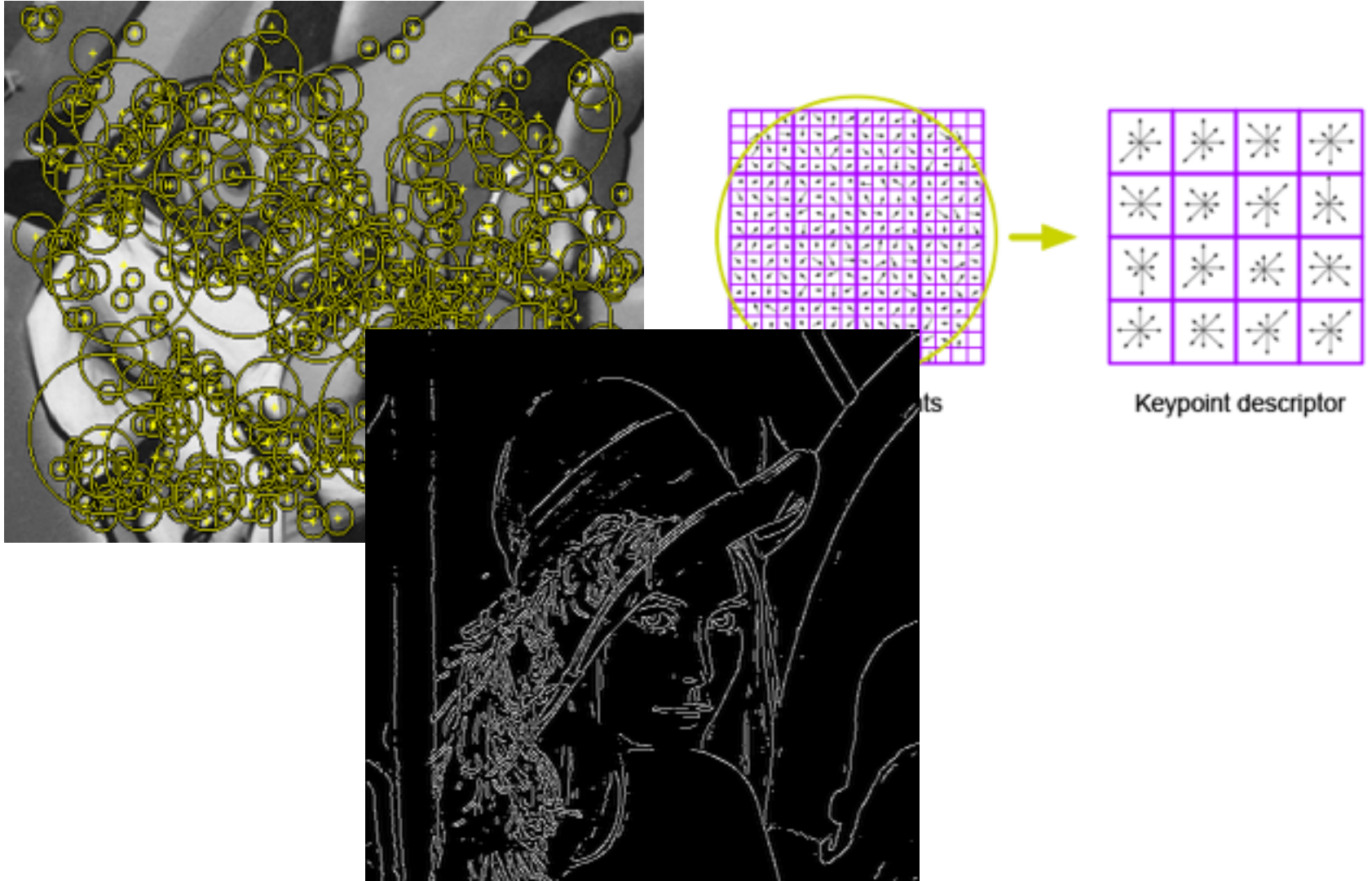


Exemplars maybe?



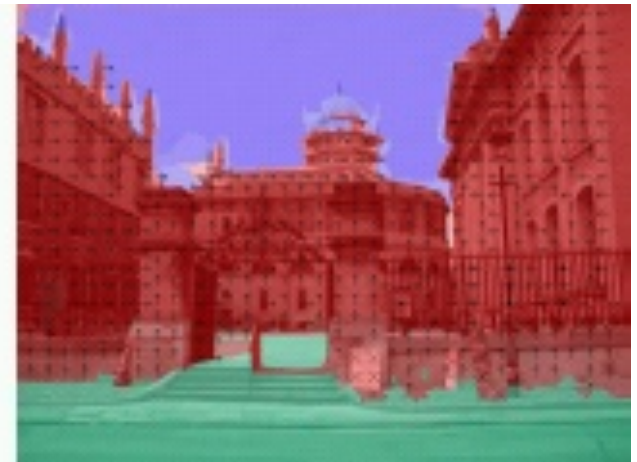
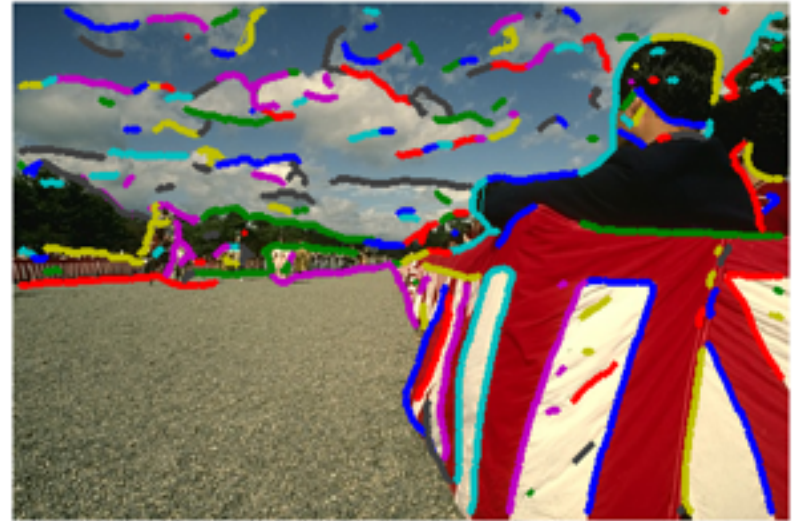


Low-Level Representation

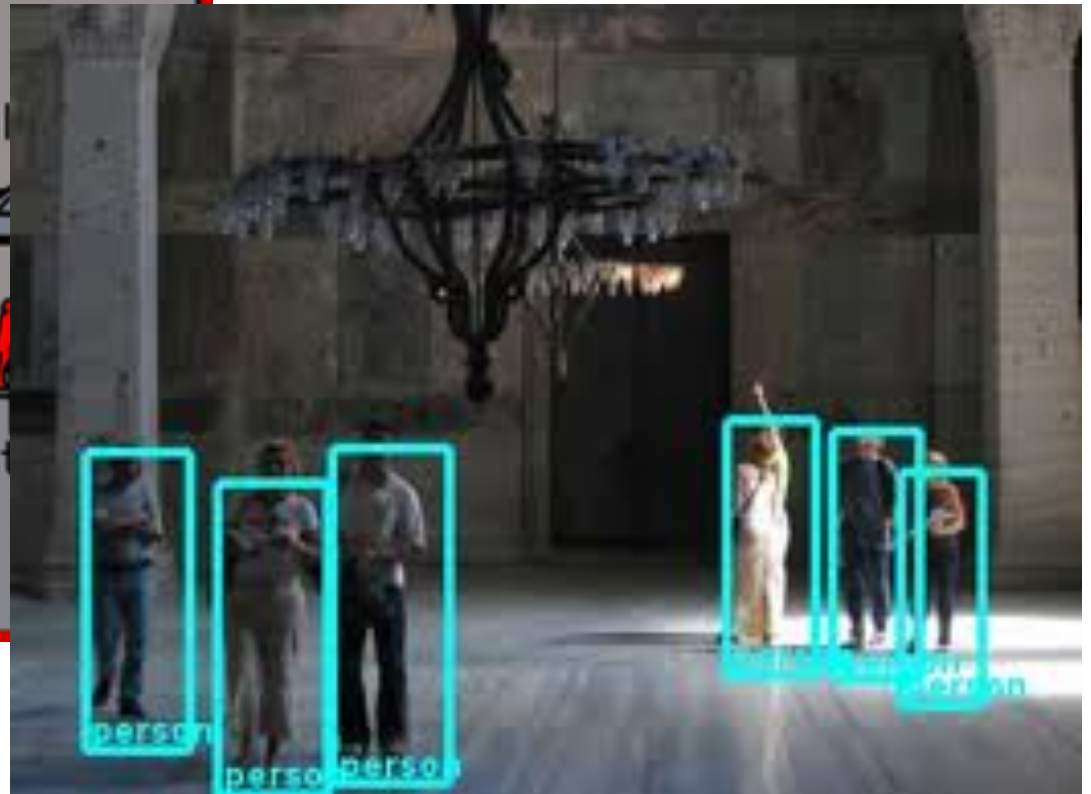


Mid-level Representations

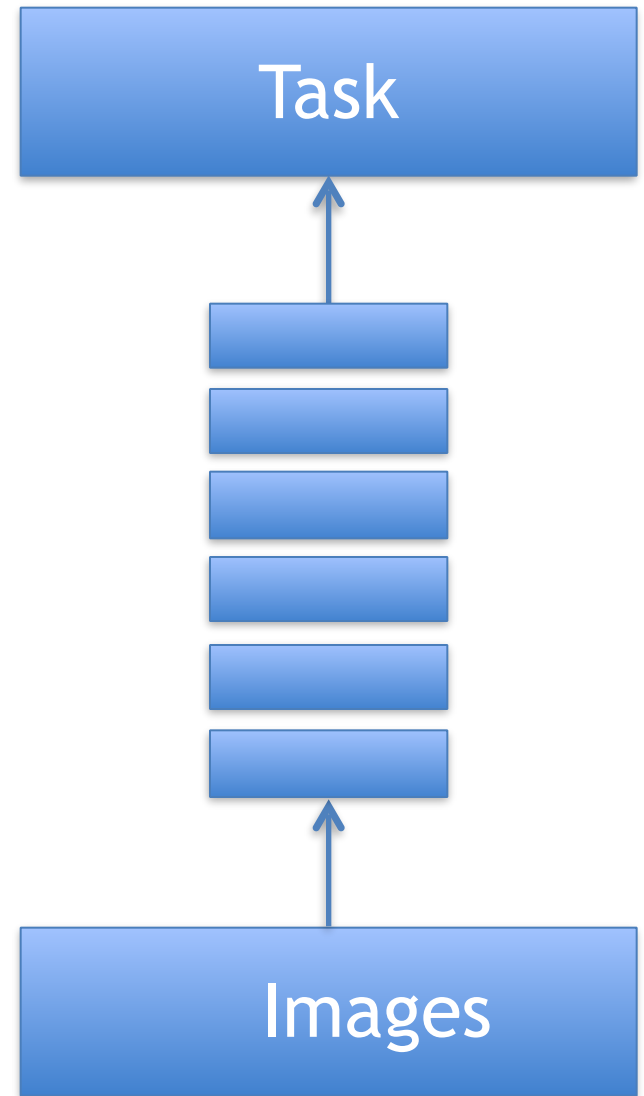
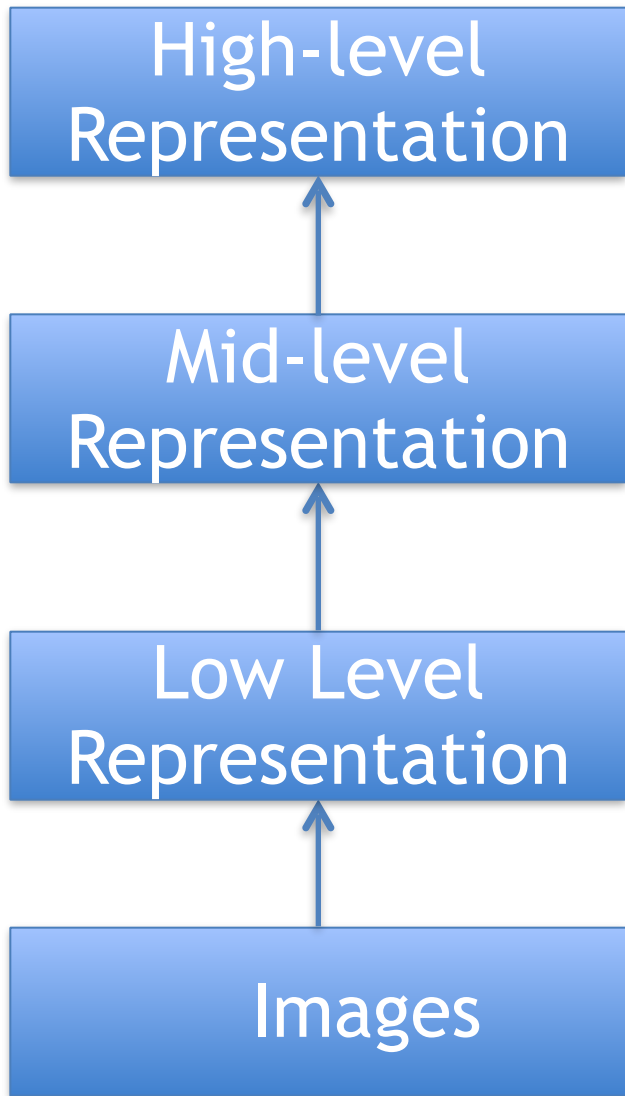
- Grouping, 3D Properties



High-level Representation



- But what are the boundaries?
- How does the information flow?



In Search of Objects: 50 years of wondering



Geometric Era

- Influenced by Ecological theory: Computer Vision started by geometric modeling the Visual World (Objects).

Geometric Era

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- Invariance to Viewpoint (Geometric Modeling allowed prediction under perspective).

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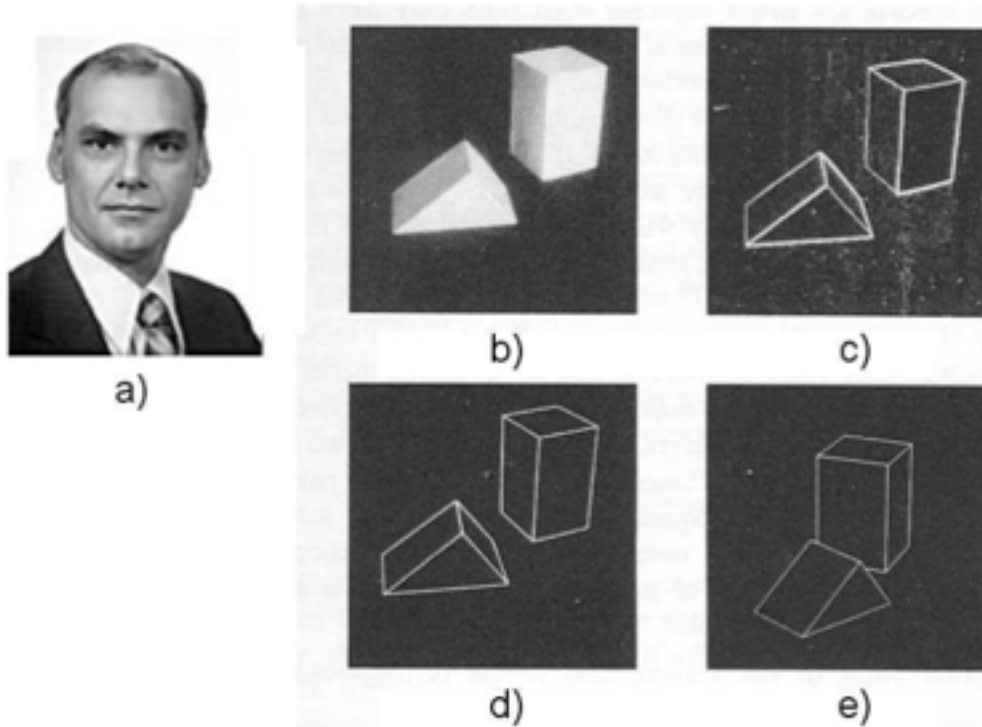
Geometric Era

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- Man-made World: Most objects are designed by humans using CAD..therefore rightly modeled using geometric primitives

Geometric Era

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- Invariance to Illumination (Matching was based on edges which were robust to illumination variations).
- Man-made World: Most objects are designed by humans using CAD..therefore rightly modeled using geometric primitives
- Well formed theories in geometry.

Blocks World at MIT



L. G. Roberts, [*Machine Perception of Three Dimensional Solids*](#), Ph.D. thesis, MIT Department of Electrical Engineering, 1963.

Fig. 1. A system for recognizing 3-d polyhedral scenes. a) L.G. Roberts. b) A blocks world scene. c) Detected edges using a 2x2 gradient operator. d) A 3-d polyhedral description of the scene, formed automatically from the single image. e) The 3-d scene displayed with a viewpoint different from the original image to demonstrate its accuracy and completeness. (b) - e) are taken from [64] with permission MIT Press.)

Why Blocks World?

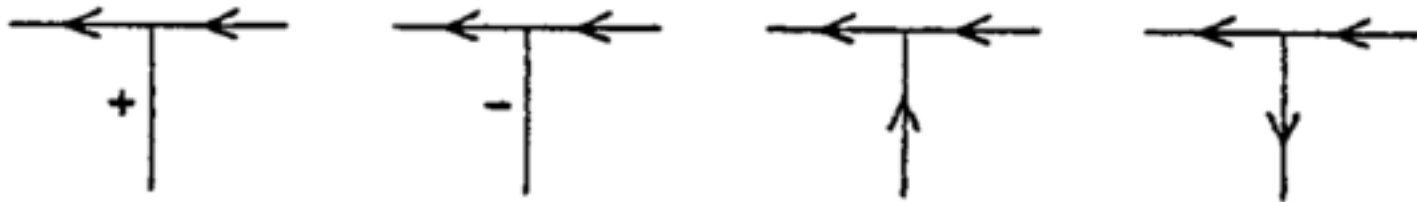
start with a simplification of the world so that the mathematical models can apply rigorously

Line Labeling and Reasoning

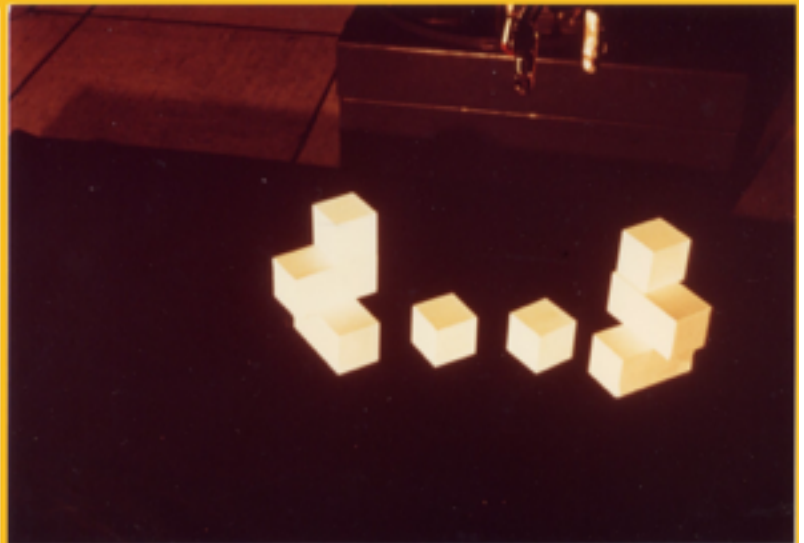
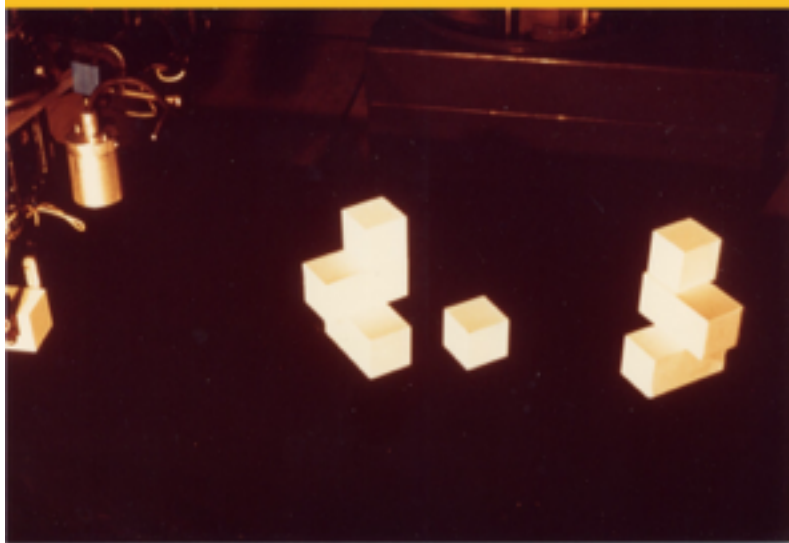
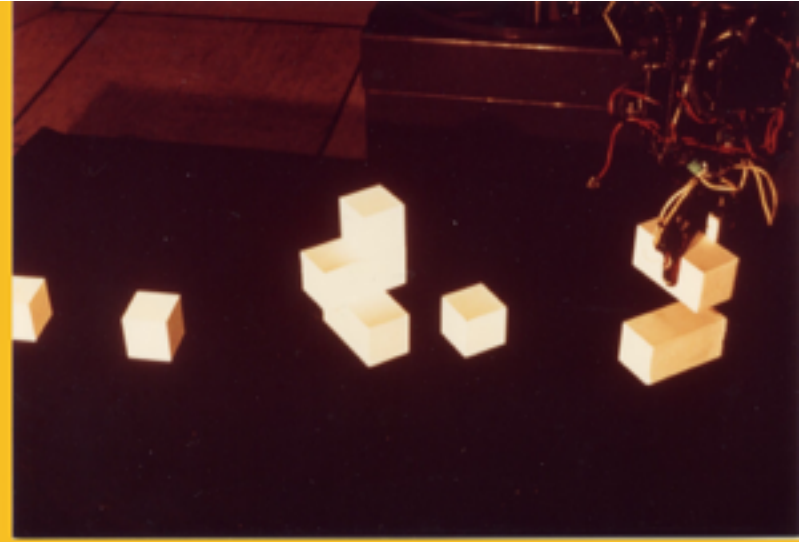
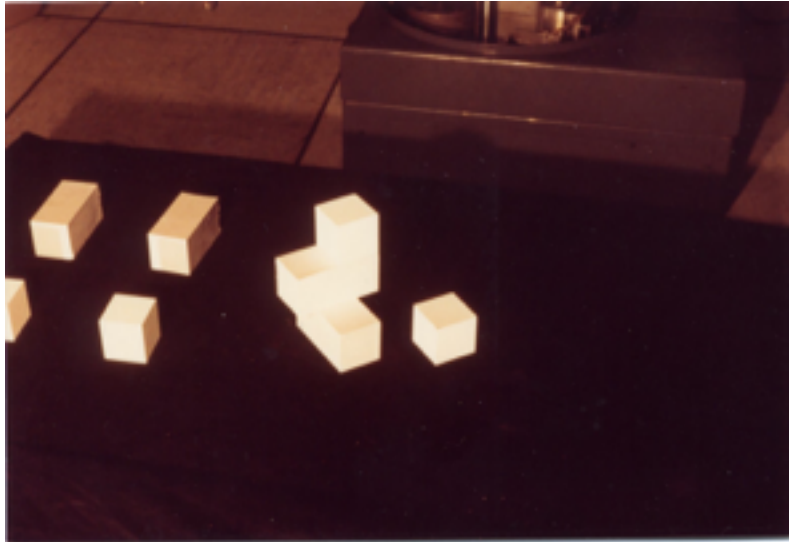
- Essentially there was no learning because there was not enough data.
- Edges were labeled convex and concave and the way these edges combine are using rules.



Figure 8 Junction types treated in this paper.



MIT Copy Demo



Why Blocks World?

Decade of Research but culminated soon...assumptions did not translate to real world

- Modeling Assumptions
 - Curved surfaces and boundaries;
 - articulated and moving objects;
- Edge Detection and Segmentation
 - complex background and 3-d texture such as foliage; - specular or mutually illuminating surfaces;
 - multiple light sources and remote shadowing; - transparent or translucent surfaces.
- Occlusion by unknown shapes;

Modeling Generalizations

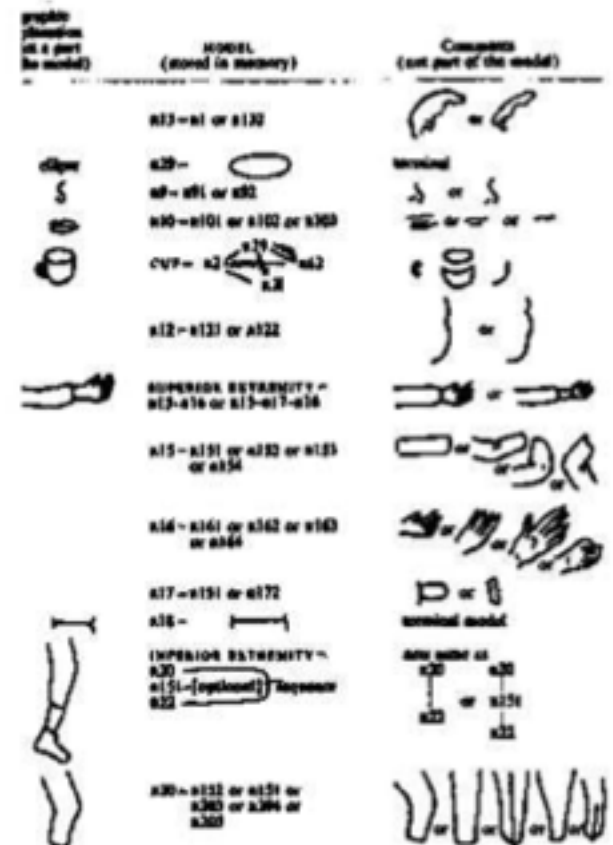
From Blocks World



a)

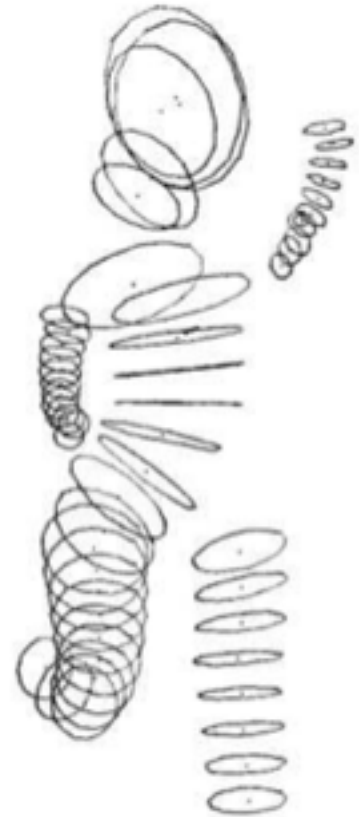


b)



c)

Generalized Cylinders

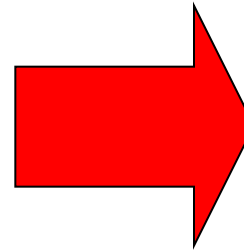
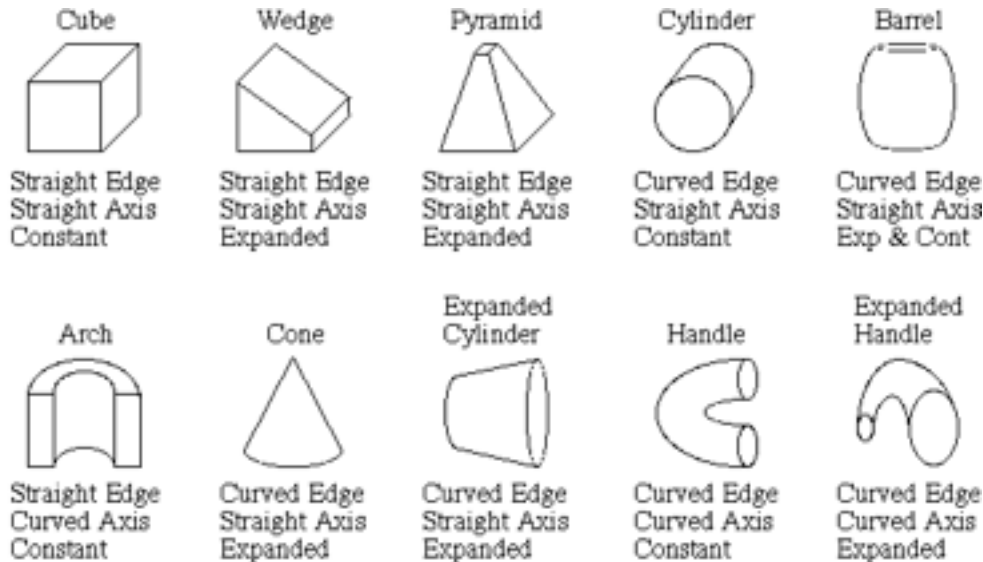


curved shapes can be expressed as a sweep of a variable cross section along a curved axis.

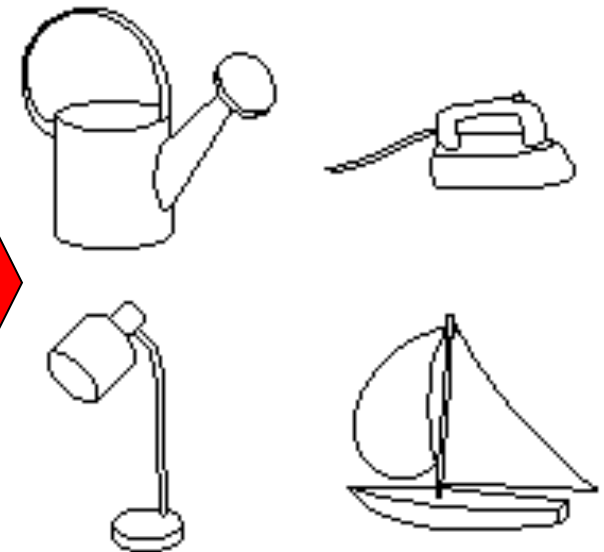
Recognition by components - Geons

Biederman (1987)

Primitives (geons)



Objects



[http://en.wikipedia.org/wiki/
Recognition_by_Components_Theory](http://en.wikipedia.org/wiki/Recognition_by_Components_Theory)

Edge/Grouping Generalization

Aspect Graph

- Until 1970-80's, research focused on object-centered representation.
- No focus on modeling image intensity variations (model every-thing from edges)
- Alternative to 3D representation of objects
- How would appearance change under different viewpoint?

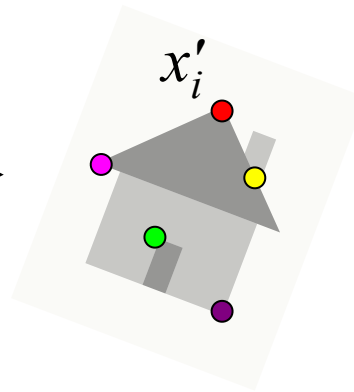
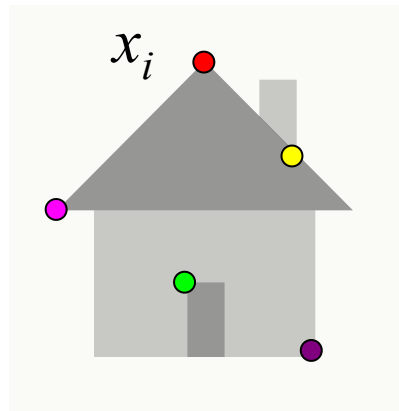
Aspect Graph

- Network of object appearances under different viewpoint.
- The nodes of the graph represent object views that are adjacent to each other on the unit sphere of viewing directions but differ in some significant way.
- the aspect graph arise from transitions in the graph structure relating vertices, edges and faces of the projected object.
- pre-compiling 2-d views into an efficient recognition plan.
- Did not carry forward due to high complexity!!

Alignment

Alignment

- Alignment: fitting a model to a transformation between pairs of features (*matches*) in two images

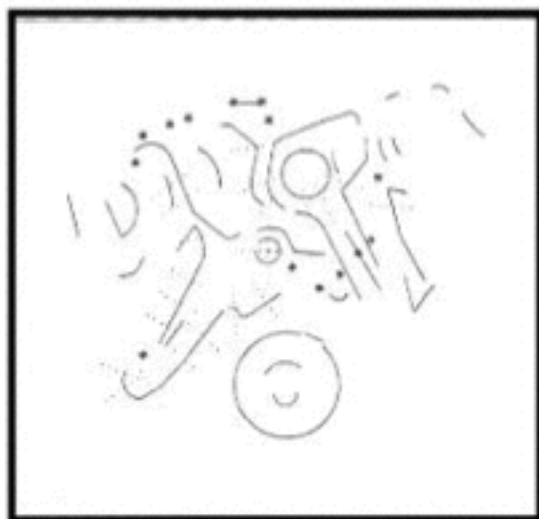


Find transformation T
that minimizes

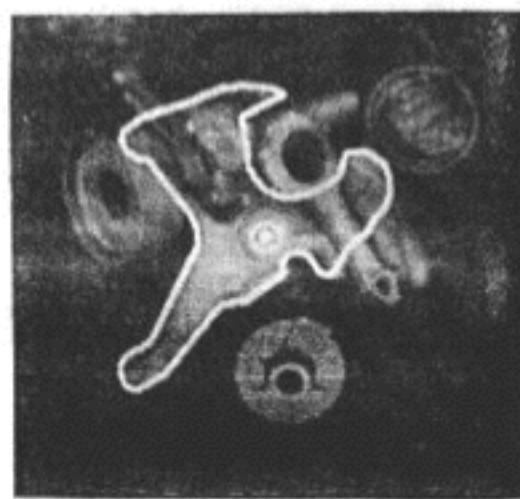
$$\sum_i \text{residual}(T(x_i), x'_i)$$



a)

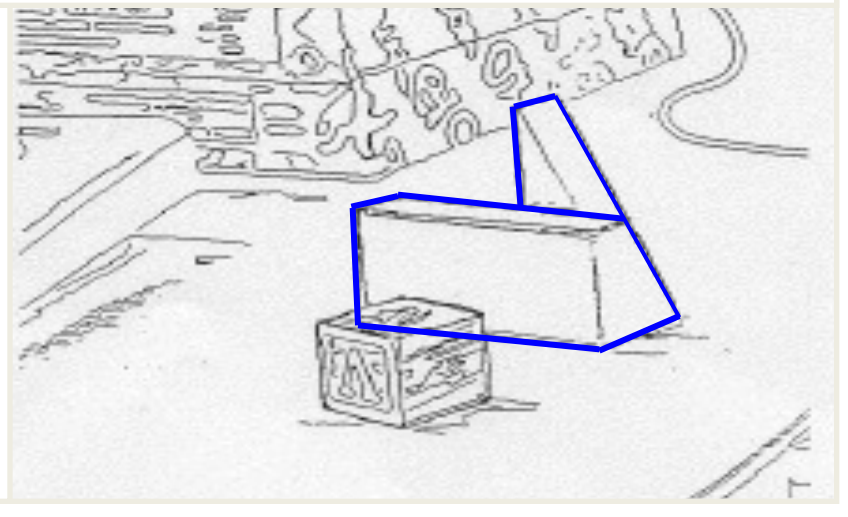
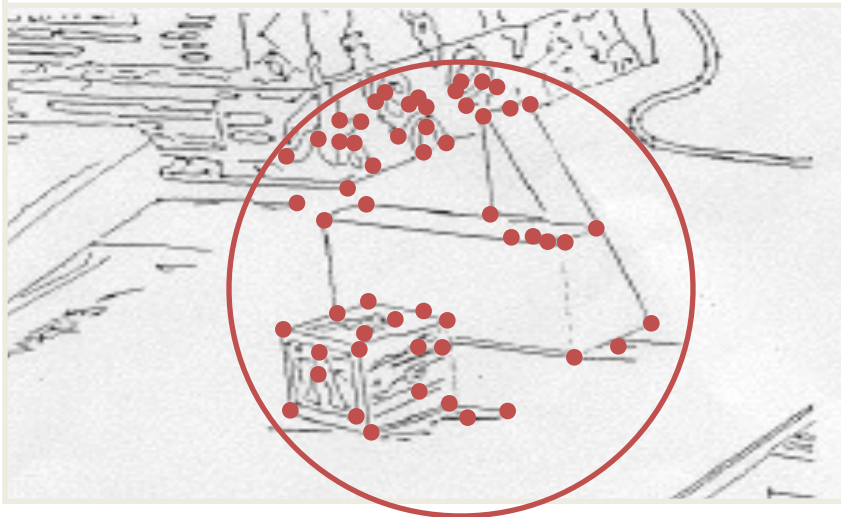
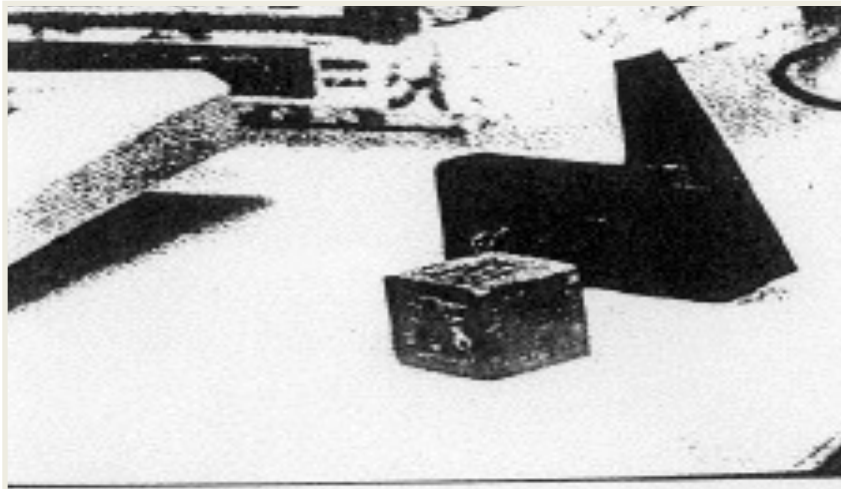


b)



c)

Alignment: Huttenlocher & Ullman (1987)



Geometric Invariants such as cross-ratios and ratio of area ratios.

Projective invariants (Rothwell et al., 1992):



General 3D objects do not admit monocular viewpoint invariants (Burns et al., 1993)

History of ideas in recognition

- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models

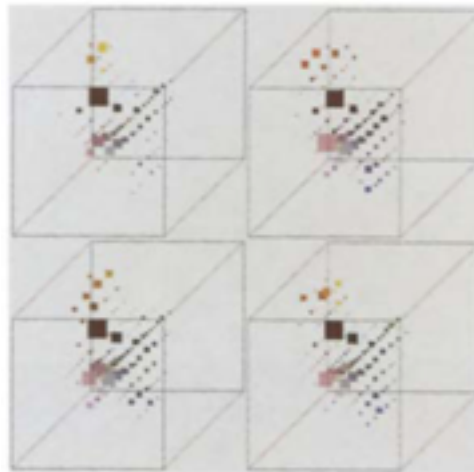
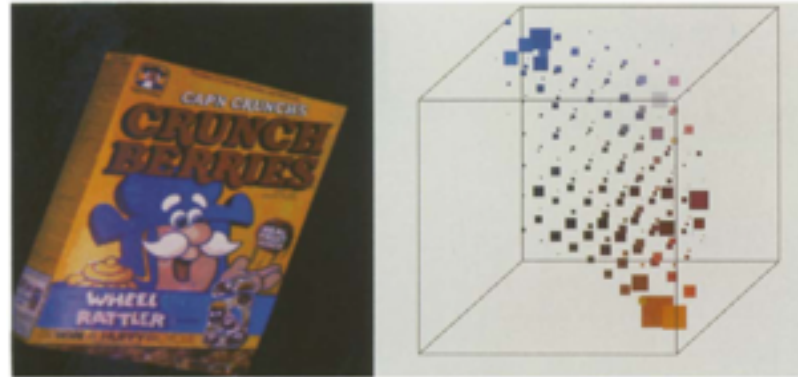
Appearance makes an Appearance

Eigenfaces (Turk & Pentland, 1991)



Experimental Condition	Correct/Unknown Recognition Percentage		
	Lighting	Orientation	Scale
Forced classification	96/0	85/0	64/0
Forced 100% accuracy	100/19	100/39	100/60
Forced 20% unknown rate	100/20	94/20	74/20

Color Histograms



Swain and Ballard, [Color Indexing](#), IJCV 1991.

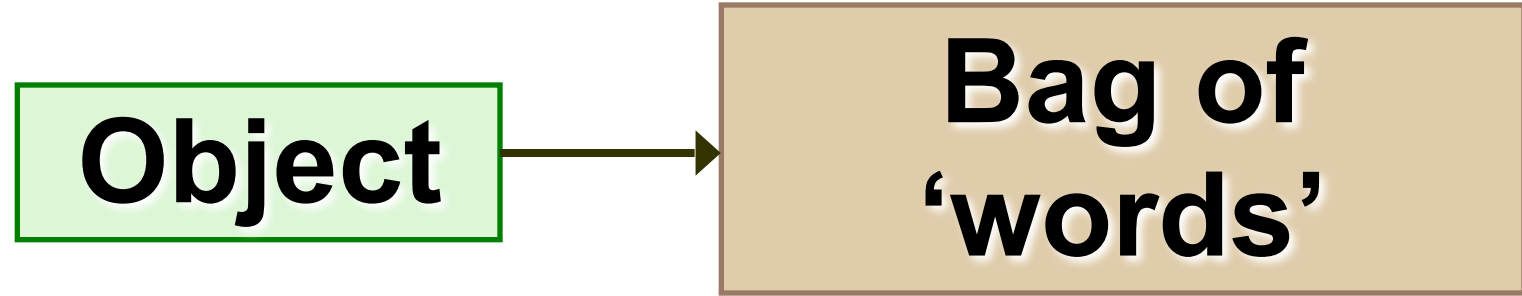
Limitations of global appearance models

- Requires global registration of patterns
- Not robust to clutter, occlusion, geometric transformations



Modeling Local Appearances

Bag-of-words models



Objects as texture

- All of these are treated as being the same



- No distinction between foreground and background: scene recognition?

Sliding-Window Paradigm

History of ideas in recognition

- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models
- 1990s – 2013: sliding window approaches

Sliding window approaches



Sliding window approaches



- Turk and Pentland, 1991
- Belhumeur, Hespanha, & Kriegman, 1997
- Schneiderman & Kanade 2004
- Viola and Jones, 2000



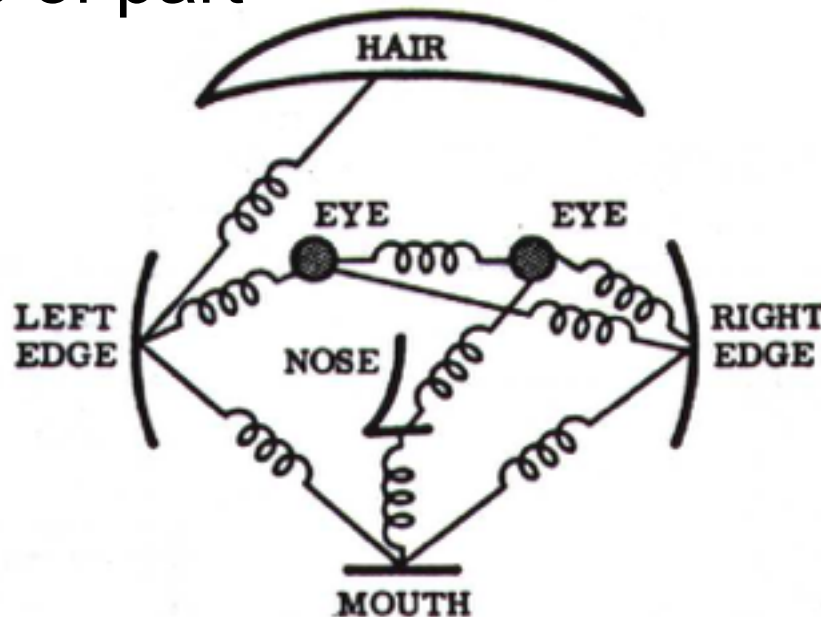
- Schneiderman & Kanade, 2004
- Argawal and Roth, 2002
- Poggio et al. 1993

History of ideas in recognition

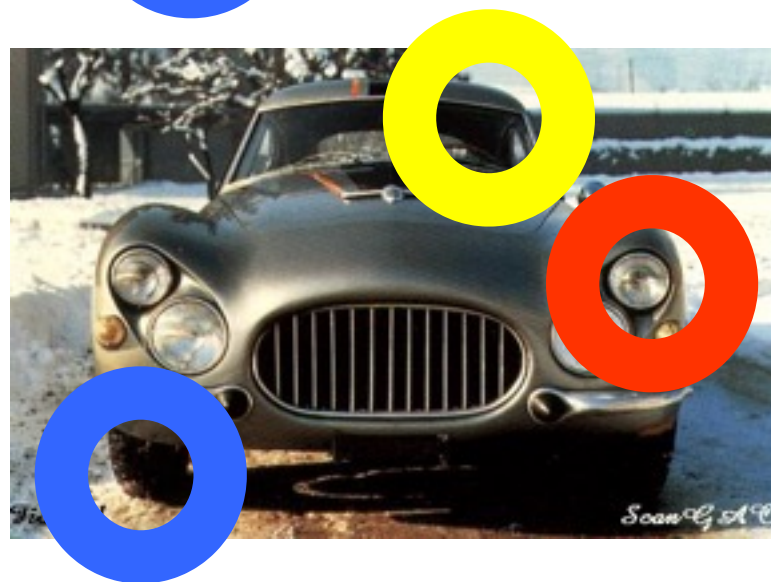
- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models
- Mid-1990s: sliding window approaches
- Early 2000s: parts-and-shape models

Parts-and-shape models

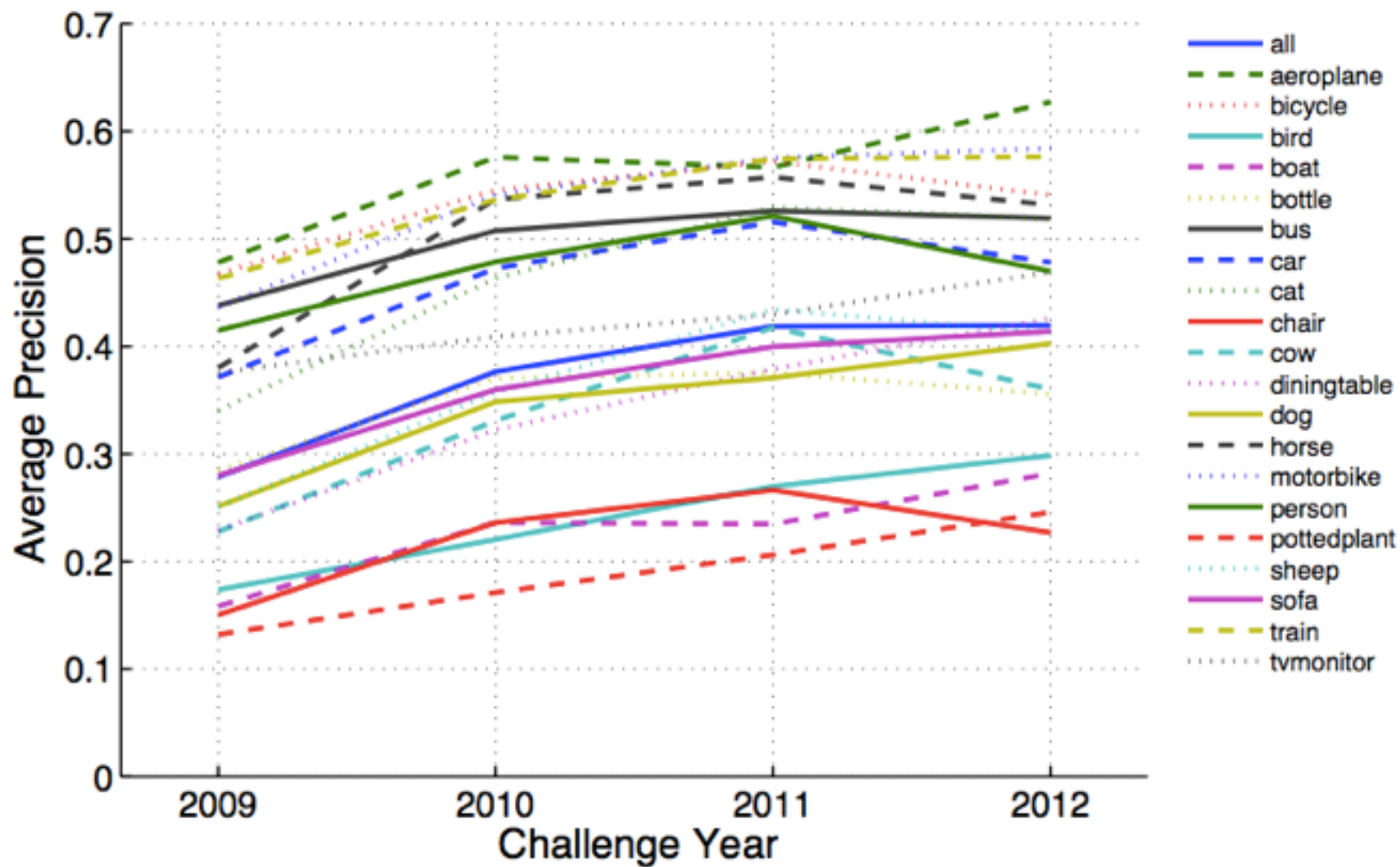
- Model:
 - Object as a set of parts
 - Relative locations between parts
 - Appearance of part



Constellation models



Weber, Welling & Perona (2000), Fergus, Perona & Zisserman (2003)



VOC Benchmark Dataset

History of ideas in recognition

- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models
- 1990s-Present: sliding window approaches
- Early 2000s-Present: parts-and-shape models
- Return of the Convolutional neural networks.

Convolutional Neural Networks

- Designed in 80's inspired by human neuron network in brain.
- Quite successful in the domain of digit recognition - Used in Postal Sorting Center
- But never competitive for generic recognition tasks.

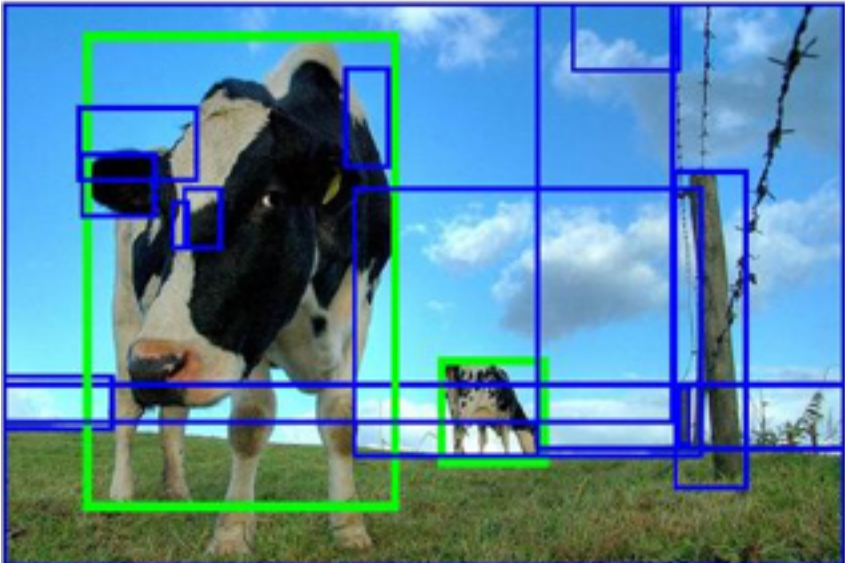
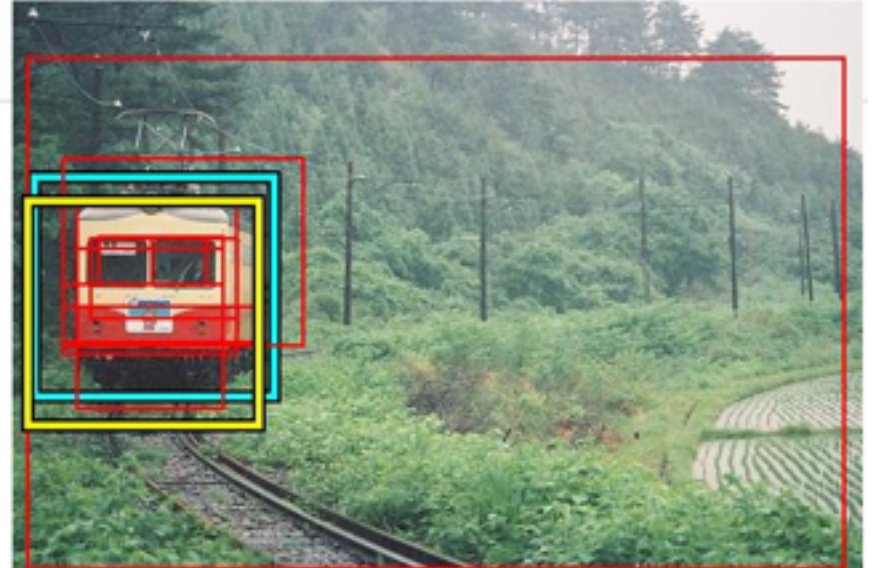
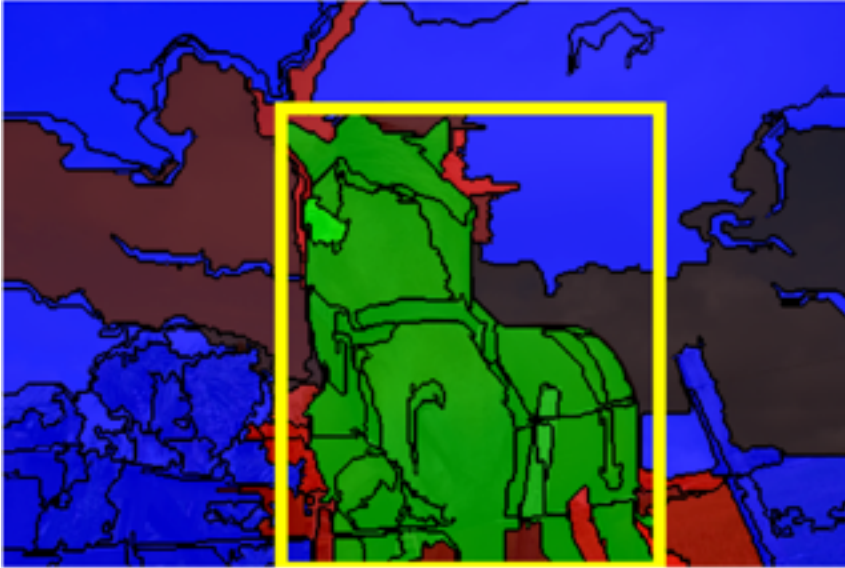
2012

- Success with the advent of bigger (10s of millions of image datasets) and better computing capacity.
- Completely supervised domain.
- Broken state of art in Classification, Detection, 3D Scene Understanding, Human Pose Estimation.
 - Videos still remain a challenge!

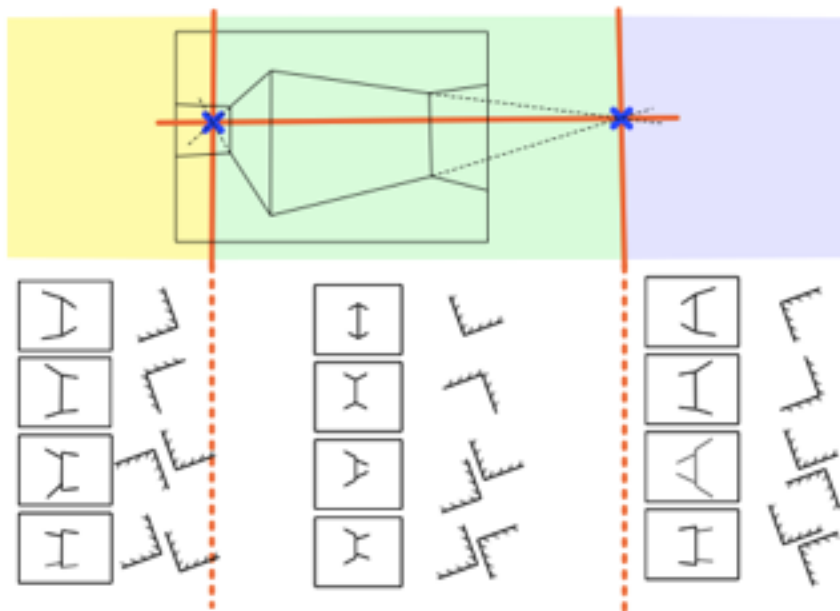
History of ideas in recognition

- 1960s – early 1990s: the geometric era
- 1990s: appearance-based models
- 1990s-Present: sliding window approaches
- Early 2000s-Present: parts-and-shape models
- Return of the Convolutional neural networks.
- 2010s - Object Proposals + CNNs

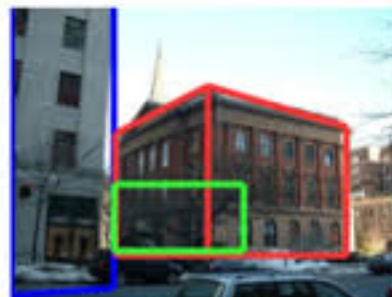
Object Proposals



What's next?
Return of Geometric Models?

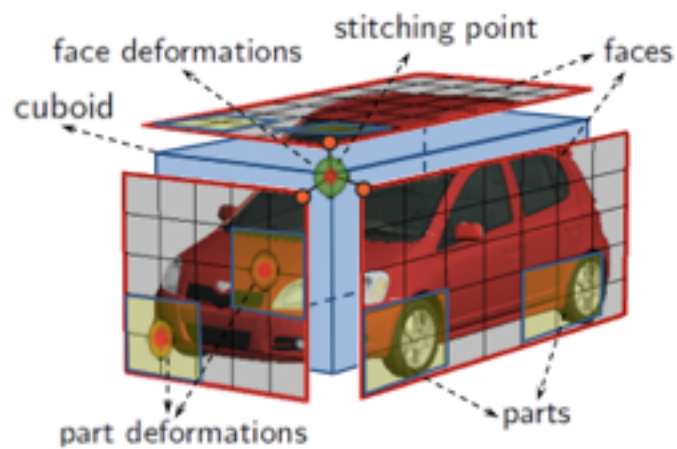


Geometric Layout (CVPR'09)

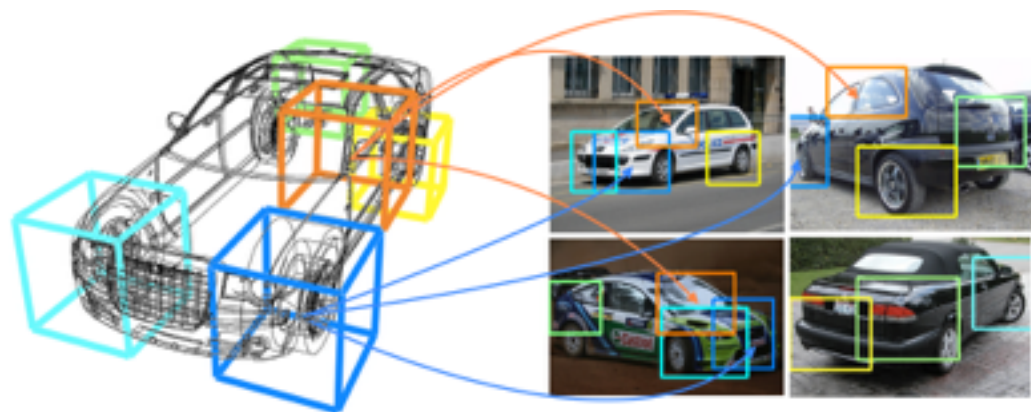


Blocks World Revisited

ECCV 2010



NIPS 2012



CVPR 2012

What “works” today

- Reading license plates, zip codes, checks



3	6	8	1	7	9	6	6	9	1
6	7	5	7	8	6	3	4	8	5
2	1	7	9	7	1	2	8	4	5
4	8	1	9	0	1	8	8	9	4
7	6	1	8	6	4	1	5	6	0
7	5	9	2	6	5	8	1	9	7
2	2	2	2	2	3	4	4	8	0
0	2	3	8	0	7	3	8	5	7
0	1	4	6	4	6	0	2	4	3
7	1	2	8	7	6	9	8	6	1

Biometrics



Fingerprint scanners on many new laptops, other devices



Face recognition systems now beginning to appear more widely
<http://www.sensiblevision.com/>



Source: S. Seitz

Mobile visual search: Google Goggles

Google Goggles in Action

Click the icons below to see the different ways Google Goggles can be used.



Landmark



Book



Contact Info.



Artwork



Places



Wine



Logo



Face detection

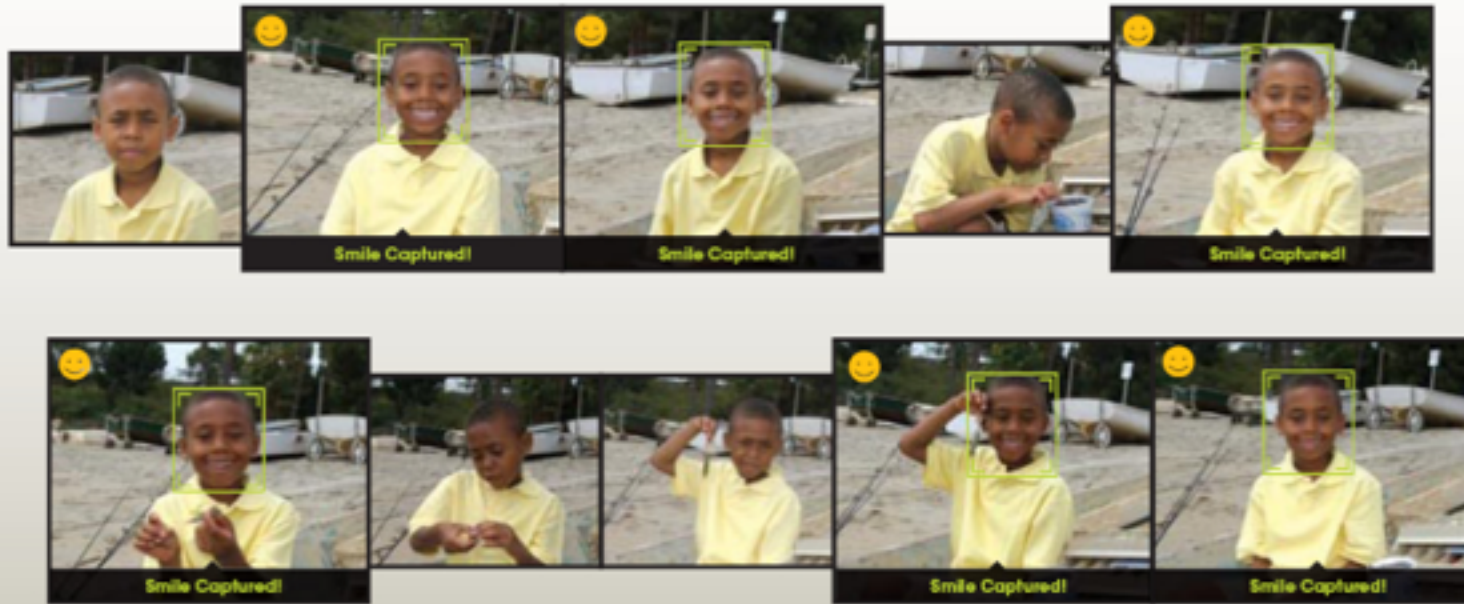


- Many new digital cameras now detect faces
 - Canon, Sony, Fuji, ...

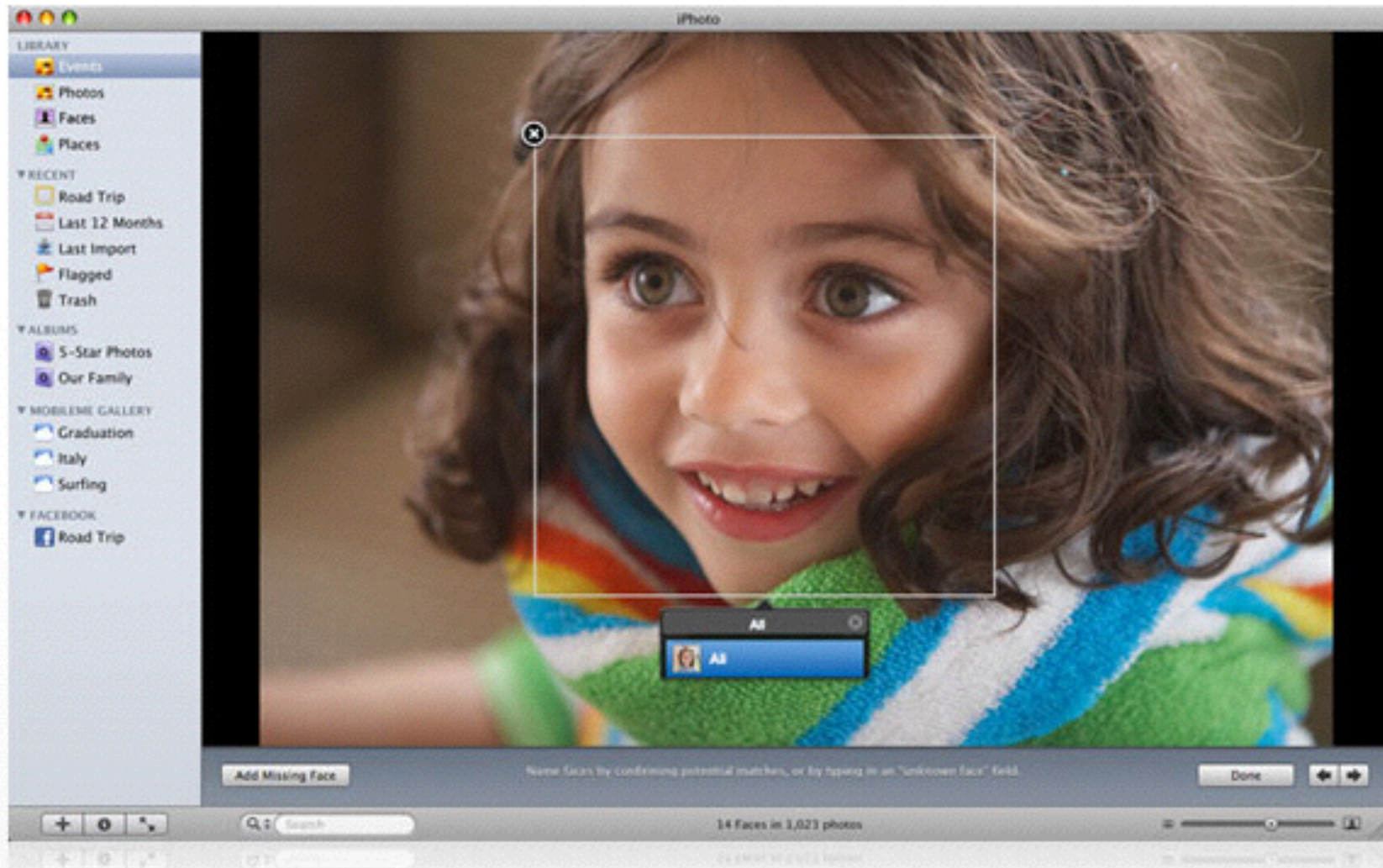
Smile detection

The Smile Shutter flow

Imagine a camera smart enough to catch every smile! In Smile Shutter Mode, your Cyber-shot® camera can automatically trip the shutter at just the right instant to catch the perfect expression.



Face recognition: Apple iPhoto, Facebook, Google, etc



Object recognition (in supermarkets)



[LaneHawk by EvolutionRobotics](#)

“A smart camera is flush-mounted in the checkout lane, continuously watching for items. When an item is detected and recognized, the cashier verifies the quantity of items that were found under the basket, and continues to close the transaction. The item can remain under the basket, and with LaneHawk, you are assured to get paid for it... “



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Last Updated: Wednesday, 31 August 2005, 05:44 GMT 06:44 UK

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Computer alert for drowning girl

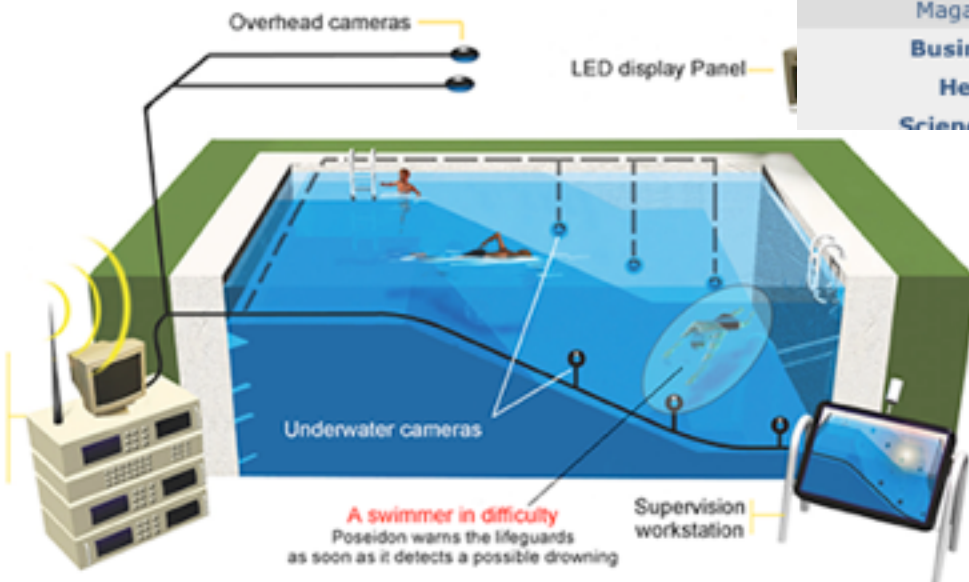
A 10-year-old girl has been saved from drowning by a computer system designed to raise the alarm when swimmers get into difficulties.



[▶ VIDEO](#) [Watch the rescue](#)

The girl, from Rochdale, was at the deep end of the pool in Bangor, north Wales, when she sank to the bottom.

The £65,000 system, called Poseidon, detected her on the pool floor and sounded the alarm. A lifeguard pulled her out and she recovered in hospital.



Security

Local 

Cameras help confirm Scott suicide ruling

Friday, December 04, 2009



TAGS: [local](#), [paul meincke](#)

 Comment Now [Email](#) [Print](#) [Report a typo](#)      



Paul Meincke

More: [Bio](#), [News Team](#)


December 4, 2009 (CHICAGO) (WLS) -- Chicago police have closed the case in the death of Chicago School Board President Michael Scott.

Police Supt. Jody Weis says investigators used police cameras in the city to trace Scott's last steps in the hours before his body was found in November.

Scott's death has been ruled a suicide. The medical examiner's office concluded --not long after Scott's body was found -- that he had committed suicide. Police did not dispute the finding but wanted to pursue all the investigative leads they could. They say they have done that and have now reached the same conclusion.

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abcNEWS

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- BP Gas Recall: BP finds, fixes source of bad gas
- Teachers union, board resume negotiating
- Back to School
- 5 injured in South Side shooting 49 min ago
- Pastor: Stacy Peterson said she lied for Drew



Automotive safety



The image shows a screenshot of the Mobileye website. The main banner features a top navigation bar with 'manufacturer products' and 'consumer products' tabs. Below this is the slogan 'Our Vision. Your Safety.' and a central image of a car with four camera fields of view: rear, forward, side, and forward. The sidebar on the right contains a 'News' section with headlines about Volvo's first collision warning system and a new collision warning system. Below the news is an 'Events' section listing Mobileye's presence at Equip Auto in Paris and SEMA in Las Vegas. The main content area below the banner is divided into three sections: 'EyeQ Vision on a Chip' with an image of a chip, 'Vision Applications' with an image of a pedestrian and text about road, vehicle, and pedestrian protection, and 'AWS Advance Warning System' with an image of a car icon and a 'CB' label.

manufacturer products consumer products

Our Vision. Your Safety.

rear looking camera forward looking camera side looking camera

> EyeQ Vision on a Chip

> Vision Applications Road, Vehicle, Pedestrian Protection and more

> AWS Advance Warning System

News

> Mobileye Advanced Technologies Power Volvo Cars World First Collision Warning With Auto Brake System

> Volvo: New Collision Warning with Auto Brake Helps Prevent Rear-end

> all news

Events

> Mobileye at Equip Auto, Paris, France

> Mobileye at SEMA, Las Vegas, NV

> read more

- [Mobileye](#): Vision systems in high-end BMW, GM, Volvo models
 - Pedestrian collision warning
 - Forward collision warning
 - Lane departure warning
 - Headway monitoring and warning

Google cars



Oct 9, 2010. ["Google Cars Drive Themselves, in Traffic"](#). [The New York Times](#). John Markoff

June 24, 2011. ["Nevada state law paves the way for driverless cars"](#). [Financial Post](#). Christine Dobby

Aug 9, 2011, ["Human error blamed after Google's driverless car sparks five-vehicle crash"](#). *The Star* (Toronto)

Vision-based interaction: Xbox Kinect

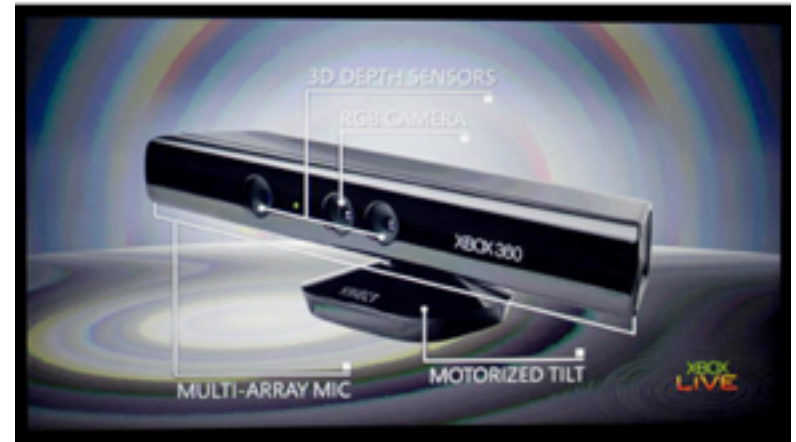


Image Search and Annotation



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- cocker spaniel

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dogs:breed=cocker spaniel