

# Chapter 9. Computer Vision

The Quest for Artificial Intelligence, Nilsson, N. J., 2009.

## Lecture Notes on Artificial Intelligence

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# Contents

## Overview of early computer vision

### 9.1 Hints from Biology

### 9.2 Recognizing Faces

### 9.3 Computer Vision of Three-Dimensional Solid Objects

Early vision approaches

Early computer vision for 3D solid objects

## Appendix

Computer Vision- Shepard table

9.3.1 An Early Vision System

9.3.2 The “Summer Vision Project”

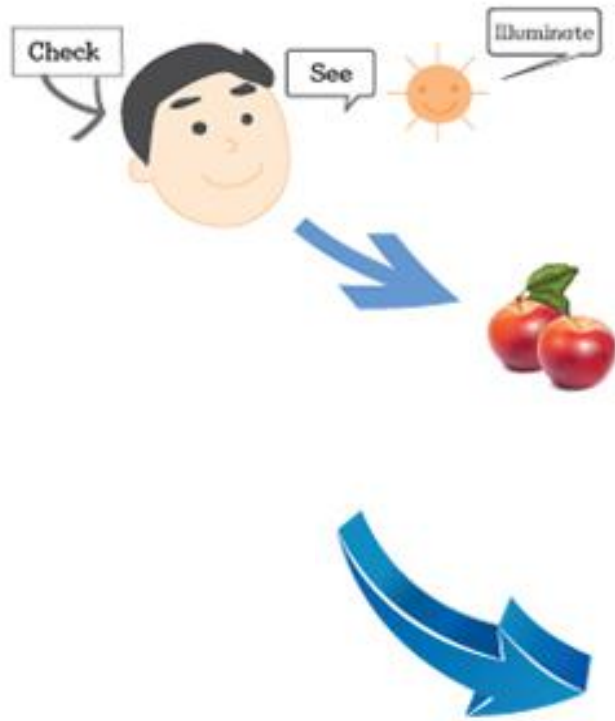
9.3.3 Image Filtering

9.3.4 Processing Line Drawing

# Overview of Chapter 9

- Computer Vision - Getting information through vision by computer
- Why difficult for computer?
  - The image is a two-dimensional project of a three-dimensional scene
  - the imaging process loss information.
- Human has stereo vision
  - Providing depth information
  - Sometimes cause errors (Shepherd's diagram)

# Human Vision vs Computer Vision



## Human VS Computer Vision



Chapter 9. Computer Vision

# 9.1 Hints from Biology

# 9.1 Hints from Biology

- “What the Frog’s Eye Tells the Frog’s Brain” (Lettvin et al., 1968.)
  - Frogs only has some detectors for objects which can help for survive. (e.g. food, predator)
- Neurophysiologic research about ‘Simple cells’ (Hubel et al, 1962, 1968)
  - Certain neurons respond selectively to images and parts of images of specific shapes.

Chapter 9. Computer Vision

## 9.2 Recognizing Faces

# 9.2 Recognizing Faces

- Early works in 1960s and 1970s
  - Trials to automatically recognize faces
  - Panoramic Research
    - Extracting set of features by human (the coordinates of center of pupils)
    - Computer calculate distance of feature like width of mouth
    - Build the database with value of features
  - Michael Kelly
    - Automatically detect facial features in pictures and use them to identify people
    - Find the location of features such as eyes, nose, or shoulders
    - match by nearest-neighbor method
  - Takeo Kanade
    - Use Program coded by assembly language
    - Run on a machine with 10ms cycle CPU and 20kB main memory
- Limitation
  - Images should be of faces of standard scale, pose, expression, and illumination



Chapter 9. Computer Vision

## **9.3 Computer Vision of Three-Dimensional Solid Objects**

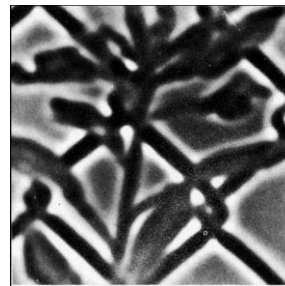
# Early Vision Approaches

- Summer Vision Project in MIT
  - Trial to develop a visual system
- Two areas on early computer vision for 3D solid objects
  - Low-level vision: constructing a representation of the image (image filtering)
  - Analyzing the line drawing as an assemblage of separate objects

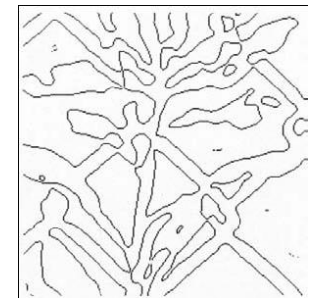
# Early computer vision for 3D solid objects

## ■ Image filtering

- Extracting features by image filtering using averaging windows
- Examples
  - Sobel operator: Finding large gradient points on vertical and horizontal directions
  - Marr-Hildreth edge detector
- Marr argued that “brain use edge detector as operator in produce “primal sketch”



Passing Sobel operator




Original image

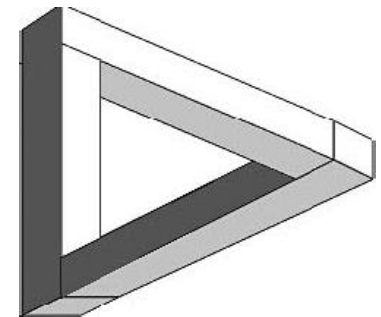
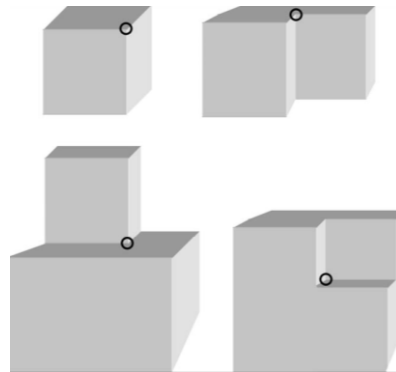
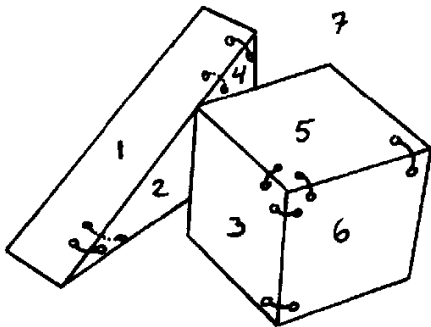
LoG-processed version

Final result

# Early computer vision for 3D solid objects

## ■ Processing line drawing

- Methods for analyzing line drawings to find objects in images
- Huffman-Clowes labeling 
  - Assigning labels to the lines by the ways in which planes could come together of the object
  - Impossible object case
    - Labels might be locally consistent, but globally inconsistent



An impossible object

Chapter 9. Computer Vision

# Appendix

# Computer Vision- Shepard table

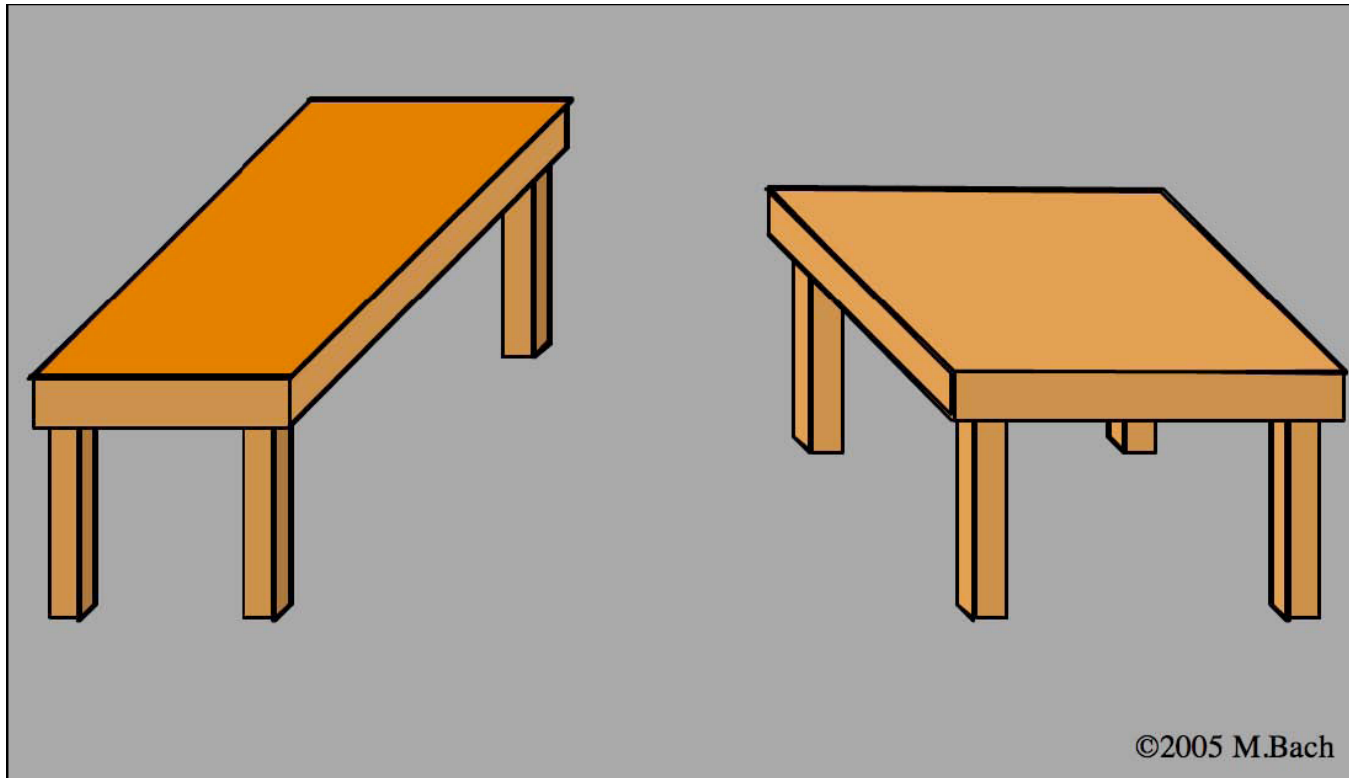


Figure 9.1: Two tables (Illustration courtesy of Michael Bach.)

Visit Bach's Web site, [http://www.michaelbach.de/ot/sze\\_shepardTables/](http://www.michaelbach.de/ot/sze_shepardTables/)

# 9.3.1 An Early Vision System

## ■ Roberts's System

- Identify objects in black-and-white photographs and determine their orientation and position in space.
- Produce a representation of a line drawing.
- Roberts Cross
  - Detecting abrupt brightness changes in an image

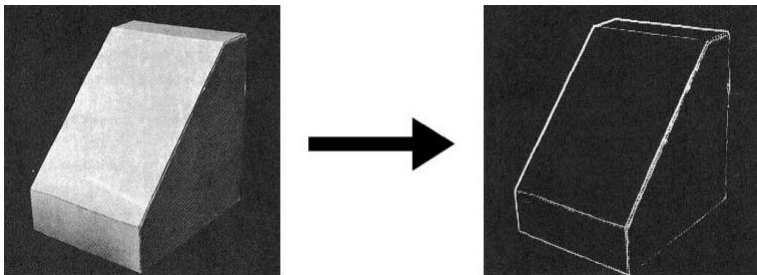


Figure 9.2 :  
Detecting changes in intensity.

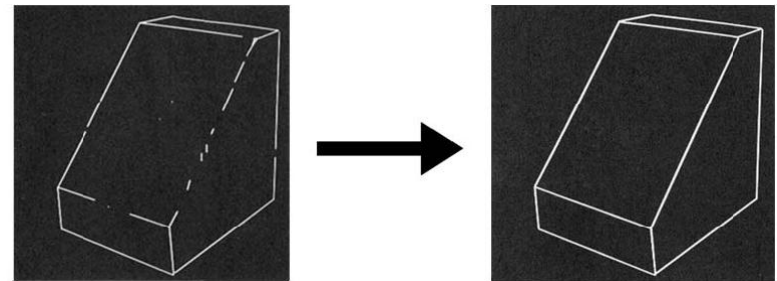


Figure 9.3 :  
Producing the final line drawing.

## 9.3.2 The “Summer Vision Project”

- Trial to develop a visual system in MIT
  - Goal
    - Building a vocabulary of known objects
    - Analyzing a picture from a “videsector” (scanner)
    - Naming objects by matching
- Two areas on early computer vision
  - Low-level vision: constructing a representation of the image (image filtering)
  - Analyzing the line drawing as an assemblage of separate objects



# 9.3.3 Image Filtering

- Why filtering images?
  - correct for noise
  - enhance image features.
- Averaging Window
  - Image averaging takes into account adjacent values and combine them
  - These intensity values correspond to an image whose right side is bright and whose left is dark with a sharp edge.

0	0	0	0	0	10	10	10	10	10
0	0	0	0	0	10	10	10	10	10
0	0	0	0	0	10	10	10	10	10
0	0	0	0	0	10	10	10	10	10
0	0	0	0	0	10	10	10	10	10
0	0	0	0	0	10	10	10	10	10
0	0	0	0	0	10	10	10	10	10
0	0	0	0	0	10	10	10	10	10
0	0	0	0	0	10	10	10	10	10
0	0	0	0	0	10	10	10	10	10

Figure 9.4: An array of image intensity values and an averaging window.



# 9.3.3 Image Filtering

## ■ Sobel Operator

- Operator use two filtering windows
  - sensitive to large gradients in the vertical direction
  - sensitive to large gradients in the horizontal direction
- The image intensity at each point is multiplied by the number in the corresponding cell of the filtering window before adding all of the numbers.

-1	0	+1	+1	+2	+1
-2	0	+2	0	0	0
-1	0	+1	-1	-2	-1

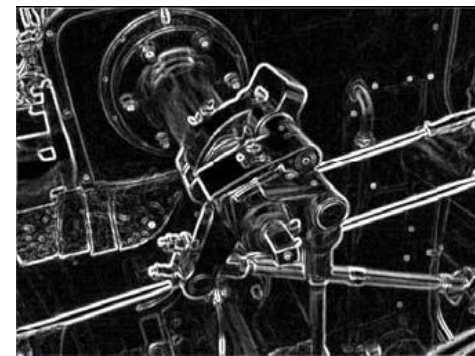


Figure 9.5: Sobel's vertical (left) and horizontal (right) filters.

Figure 9.6: Finding abrupt changes in image brightness with the Sobel Operator.

Visit [http://en.wikipedia.org/wiki/Sobel\\_operator](http://en.wikipedia.org/wiki/Sobel_operator).

# 9.3.3 Image Filtering

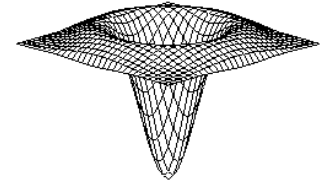


## ■ Marr-Hildreth edge detector

- Laplacian of Gaussian (LoG)
  - Mexican hat
  - Sombrero function
- LoG-processing
  - Giving whitish bands surrounding darker parts
- Finding the transitions from light to dark (and vice versa)
- Marr argued that “brain use edge detector as operator in produce “primal sketch”

0	0	3	2	2	2	3	0	0
0	2	3	5	5	5	3	2	0
3	3	5	3	0	3	5	3	3
2	5	3	-12	-23	-12	3	5	2
2	5	0	-23	-40	-23	0	5	2
2	5	3	-12	-23	-12	3	5	2
3	3	5	3	0	3	5	3	3
0	2	3	5	5	5	3	2	0
0	0	3	2	2	2	3	0	0

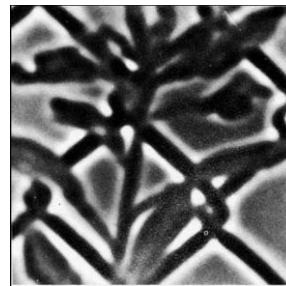
LoG filtering window



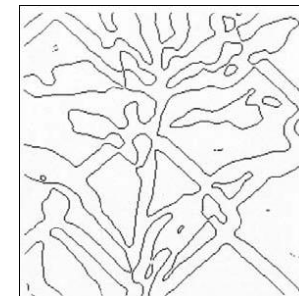
LoG surface



Original image



LoG-processed version



Final result

# 9.3.4 Processing Line Drawing

- Methods for analyzing line drawings to find objects in images
- Guzman-Arenas' SEE
  - Developed LISP program called SEE
  - Procedure
    - Sorting its vertices into a number of different types.
    - For each vertex, depending on its type, SEE connected adjacent planar surfaces with "links."
  - Capable of finding bodies in rather complex scenes.
  - Could not identify blocks that had holes in them.

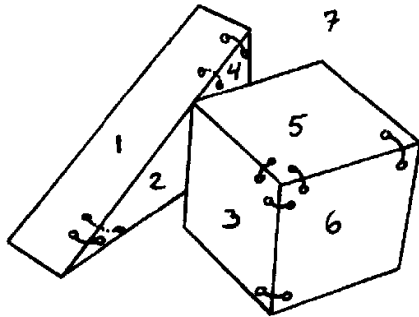


Figure 9.11: Links established by SEE for a sample scene.

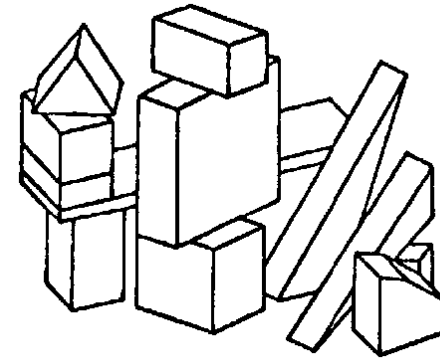


Figure 9.12: A scene analyzed by SEE.

# 9.3.4 Processing Line Drawing

- **Huffman's Work (and Max Clowes)**
  - Huffman-Clowes labeling
    - Assigning labels to the lines by the ways in which planes could come together of the object
    - T-nodes
      - Four types in which three plane surface can come together at a vertex
      - Line intersection types caused by one object in a scene occluding another
    - Impossible object case
      - Labels might be locally consistent, but globally inconsistent

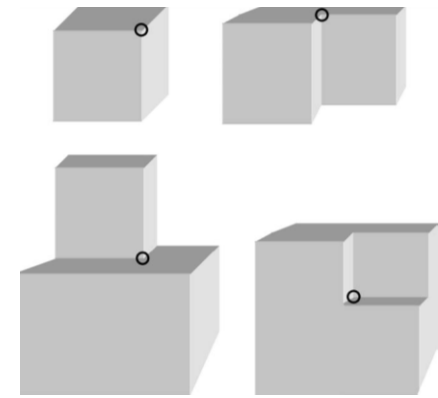


Figure 9.13: The four different kinds of vertices that can occur in trihedral solids.

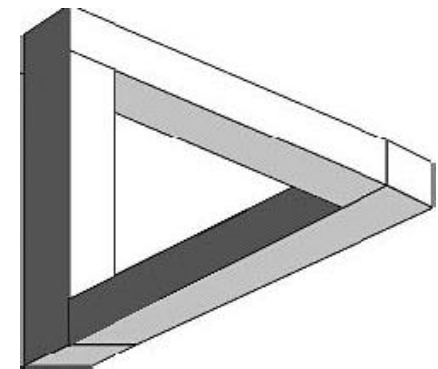


Figure 9.14: An impossible object.

# 9.3.4 Processing Line Drawing

## ■ Waltz's Work

- Extend Huffman's scheme to allow for line drawings of scenes with shadows and possible "cracks" between two adjoining object.
- Propose and implement an efficient computational method for satisfying the constraint that all of the lines must be assigned only one label

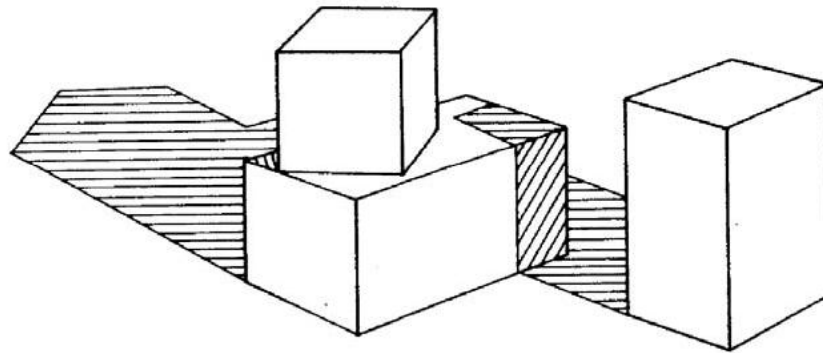


Figure 9.15: A scene with shadows analyzed by Waltz's program.