Lecture 10

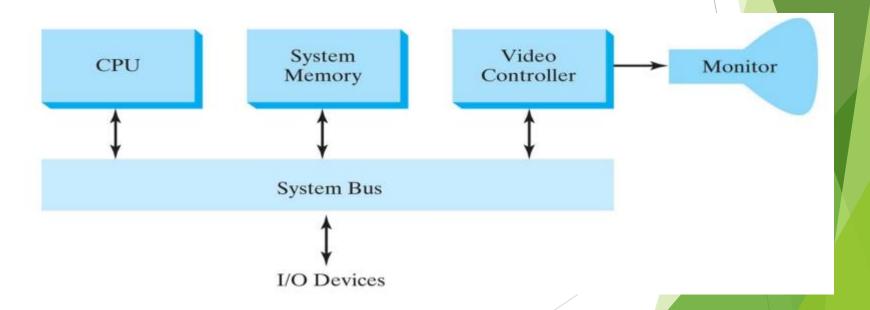
Recap

- HTML Canvas
 - Comparison to SVG browser graphics
 - Drawing shapes
 - Basic Trigonometry
 - Basic collision detection
 - Animation and User Interaction
 - Examples
 - Advanced collision handling
 - ► Linear Algebra and Transformations
- SVG
- **D**3
- ▶ 3D WebGL/Three.js

Graphics Hardware

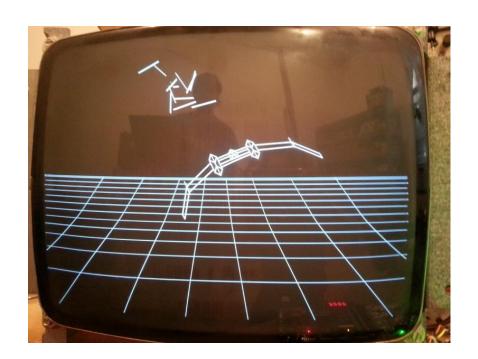
Graphics Hardware Intro

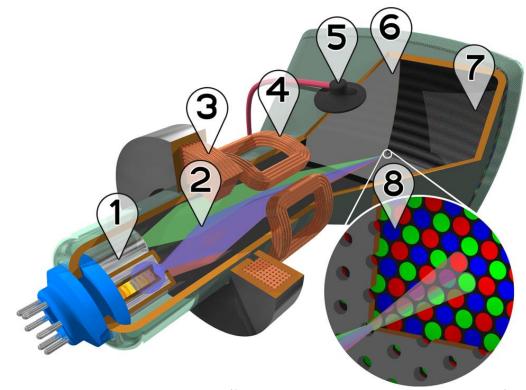
- Basic graphics hardware
 - Display devices
 - Video controller
 - Memory
 - ► CPU
 - System bus



Display Devices

- Older
 - ▶ CRT
 - Vector Display

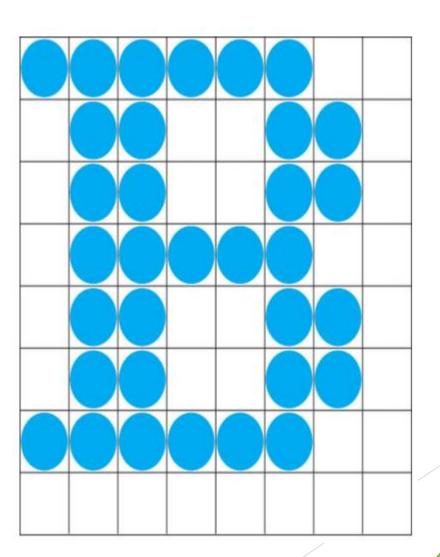




- 1. Three electron emitters (for red, green, and blue phosphor dots)
- 2. Electron beams
- 3. Focusing coils
- 4. Deflection coils
- 5. Anode (collector)
- 6. Mask for separating beams for red, green, and blue part of displayed image
- 7. Phosphor layer (screen) with red, green, and blue zones
- 8. Close-up of the phosphor-coated inner side of the screen

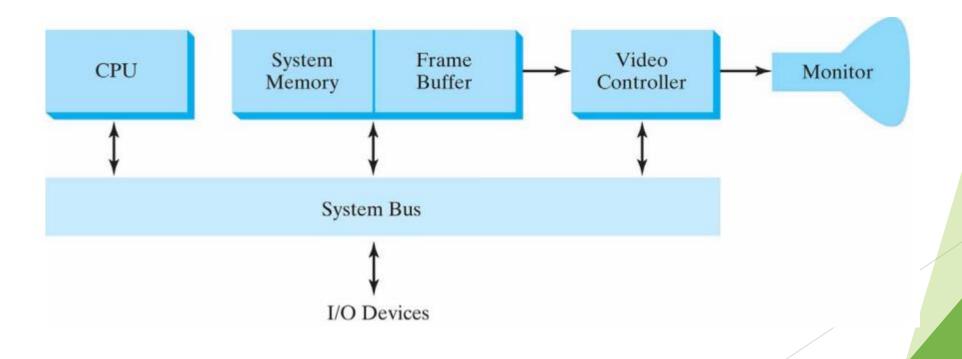
Raster Displays

Example - how to draw a character:



Integrated Graphics Hardware for Raster Display System

Architecture of a raster system with a fixed portion of the system memory reserved for the frame buffer.

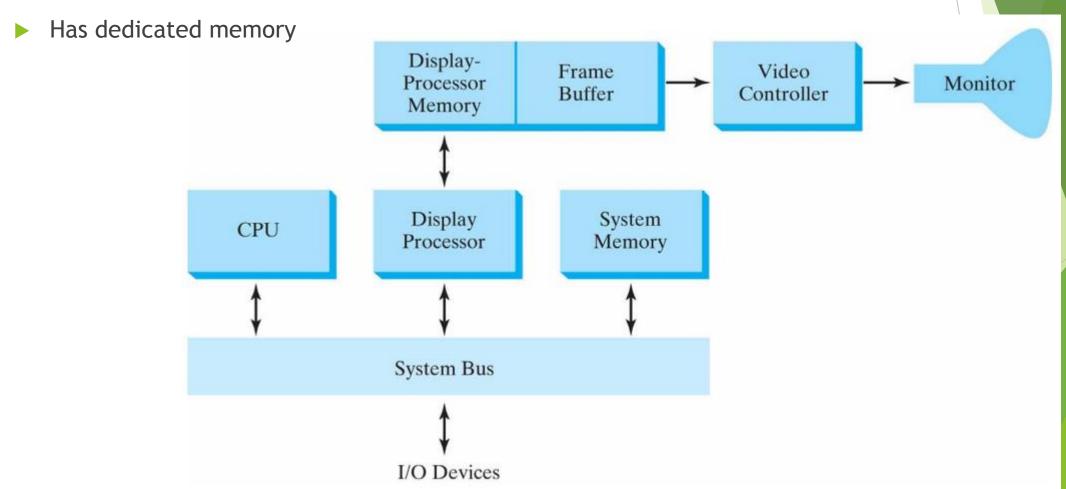


Display Processor-based Raster-Graphics Computer System.

- As the demand for better graphics increased, hardware manufacturers created a way to decrease the amount of CPU time required to fill the framebuffer.
 - Commonly called "graphics accelerating".
- ► Common graphics drawing commands (many of them geometric) are sent to the graphics accelerator in their raw form.
 - The accelerator then rasterizes the results of the command to the framebuffer.
 - ► This method can save thousands or millions of CPU cycles per command, as the CPU is freed to do other work.

Display Processor-based Raster-Graphics Computer System.

Architecture of a raster-graphics system with a display processor.



Graphics Card

- A graphics card, also known as a graphics accelerator card, display adapter, or video card, is a PC hardware component whose function is to generate and output images to a display. It operates on similar principles as a sound card or other peripheral devices.
- Components
 - ► Graphics processing unit (GPU): a dedicated graphics microprocessor optimized for floating point calculations which are fundamental to 3D graphics rendering.
- Video memory
 - ▶ DDR RAM, 128 MB to 12.0 GB, 400 MHz to 2.4 GHz.
 - Store program, data, and frame buffer

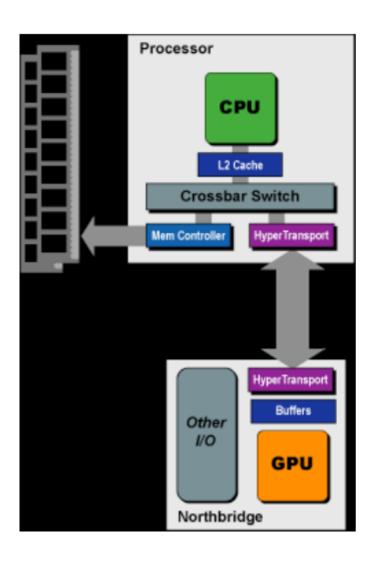
Frame Buffer

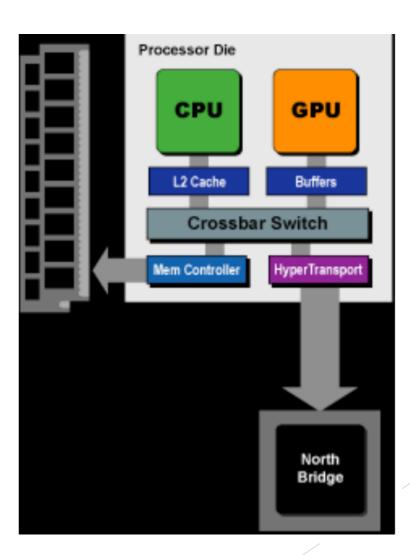
- ► The frame buffer is a part of RAM in a computer allocated to hold the graphics information for one frame of image.
- Frame buffer (FB) size determines the maximum resolution and color depth of the image
 - ► FB size = resolution × color depth,
 - ► E.g. 640X480X8 bit = 2,457,600 bits = 307,200 Bytes
 - \triangleright E.g. 1024x768x24 bit = 18,874,368 bits = 2,359,296 Bytes
 - ► 4K image: 3840x2160x32 = ~33.2 MB
- Frame buffer can be a part of main memory, or on the graphics card
- Image is generated by CPU/GPU (in vector/raster) and written (or loaded) into frame buffer
- Image in frame buffer is read out by display controller/interface to display on the screen

GPU has more computing power

- ► GPU has a massively parallel architecture
 - Many problems map well to GPU-style computing
 - GPUs have large amount of arithmetic capability
 - Increasing amount of programmability in the pipeline
- Programmable GPU
 - New features map well to GPU
 - Unified shaders
 - Direct access to computing units in new APIs
- Challenge
 - ▶ How do we make the best use of GPU hardware?
 - ► Techniques, programming models, languages

CPU/GPU Fusion (Integrated Graphics)

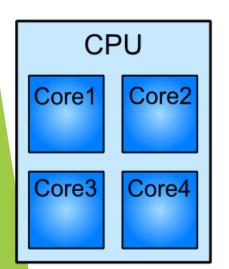


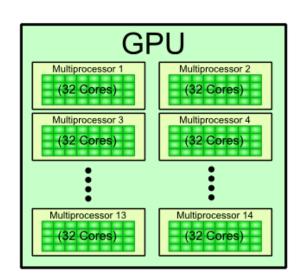


GPU vs CPU

- Highest spec:
 - http://www.nvidia.com/gtx-700-graphicscards/gtx-titan-z/
 - https://blogs.nvidia.com/blog/2014/03/25/titanz/

CPU/GPU Architecture Comparison







ULTIMATE POWER. THE NEW GEFORCE® GTX TITAN Z.

GeForce® GTX TITAN Z is a gaming monster, the fastest card we've ever built to power the most extreme PC gaming rigs on the planet. Stacked with 5760 cores and 12 GB of memory, this dual GPU gives you the power to drive even the most insane multimonitor displays and 4K hyper PC machines

GTX TITAN Z GPU ENGINE SPECS:

CUDA Cores 5766

Base Clock (MHz) 705

Boost Clock (MHz) 876

Texture Fill Rate 338
(billion/sec)

GTX TITAN Z MEMORY SPECS:

Memory Speed 7.0 Gbps
Standard Memory Config 12288 MB

Memory Interface

Memory Interface Width

768-bit (384bit per GPU)

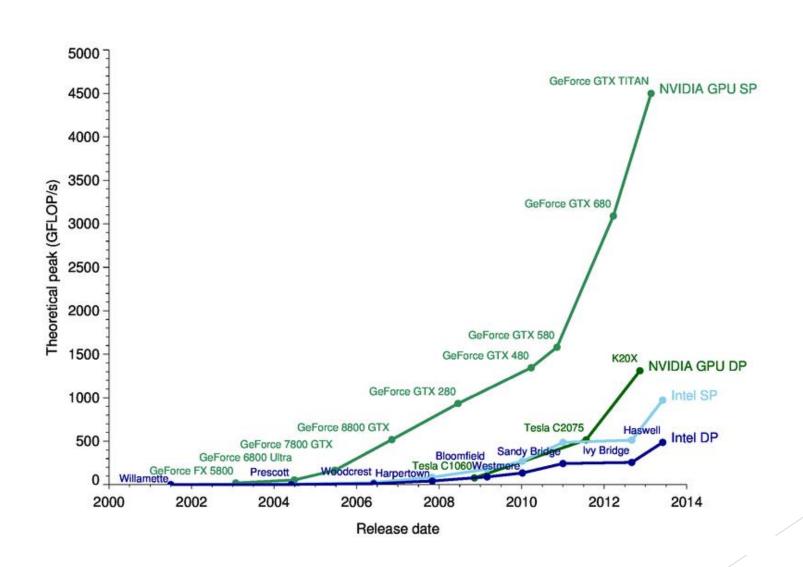
GDDR5

Memory Bandwidth 6

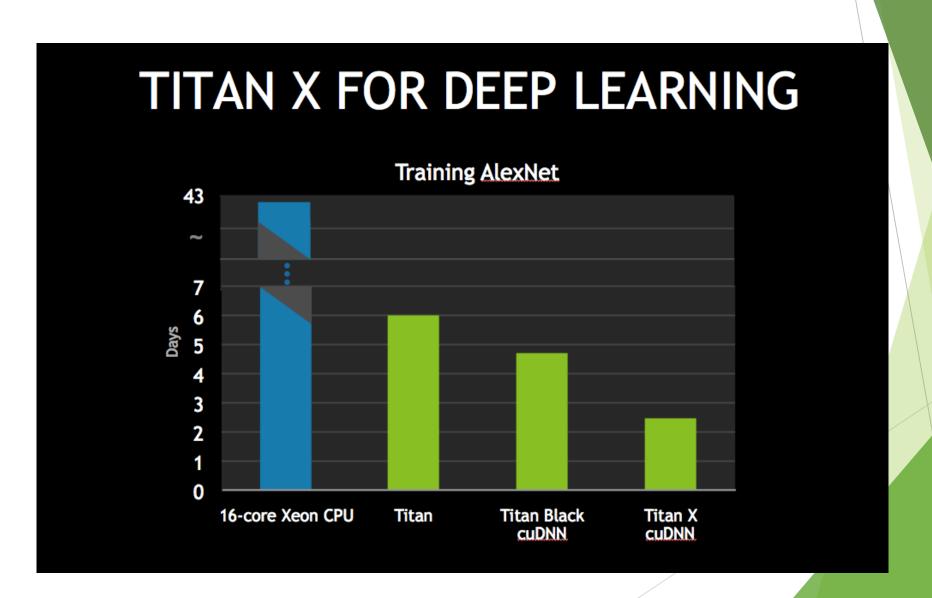
(GB/sec)

672

GPU vs CPU



GPU vs CPU



Refresh rates and Resolution

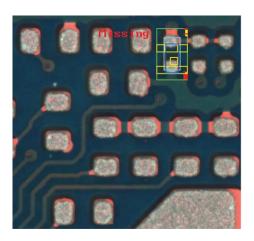
- Frequency:
 - ► How many frames per second
 - ► Typically with CRT, was 60 frames per second
 - What's the update frequency with 4K TV
 - https://www.cnet.com/uk/news/ultra-hd-4k-tv-refresh-rates/

Computer Vision

Introduction

- Computer Vision is about understanding images. These images can be
 - Greyscale or Colour
 - Snapshots or video sequences
 - ► Taken with a static or moving camera
 - ► Taken of a stationary or dynamic scene
 - ► Taken with a calibrated or un-calibrated camera
- ▶ What Computer Vision aims to do is to extract some useful information from these images for...
 - Inspection purposes
 - Analysis purposes
 - Control purposes

Applications - Industrial Inspection













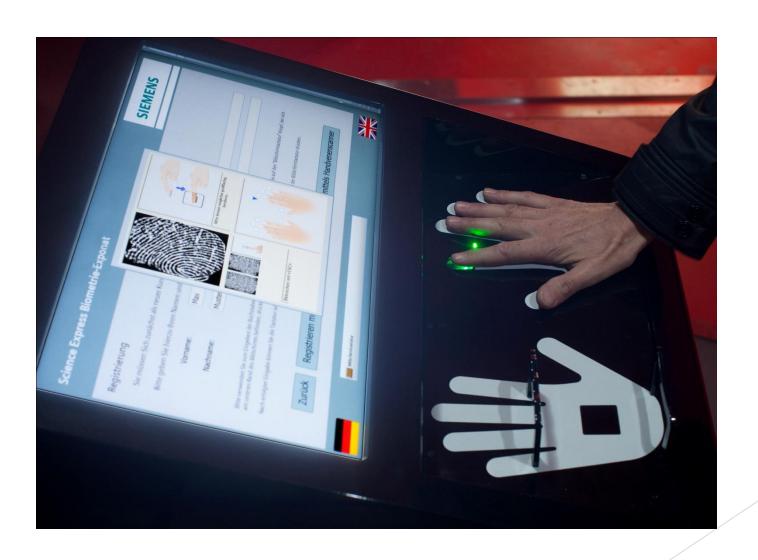


Applications - Surveillance / Forensics

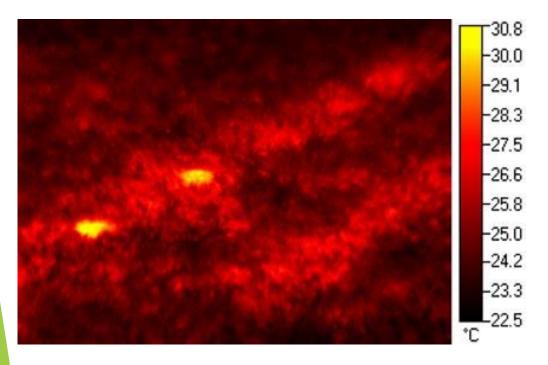
- ► In London there are 1,000,000+ cameras
- However, only 1 crime solved per 1,000 cameras (BBC, 2009)
- In July 2005 there was a major terrorist attack in London...
- Afterwards
 - ▶ 17,000 hours of video was studied
 - Over 1 year

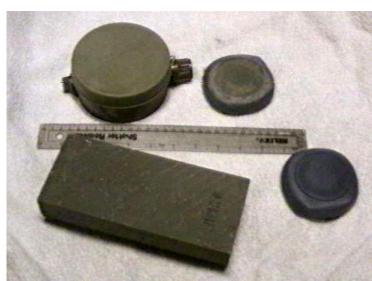


Applications - Biometrics



Applications - Landmine detection





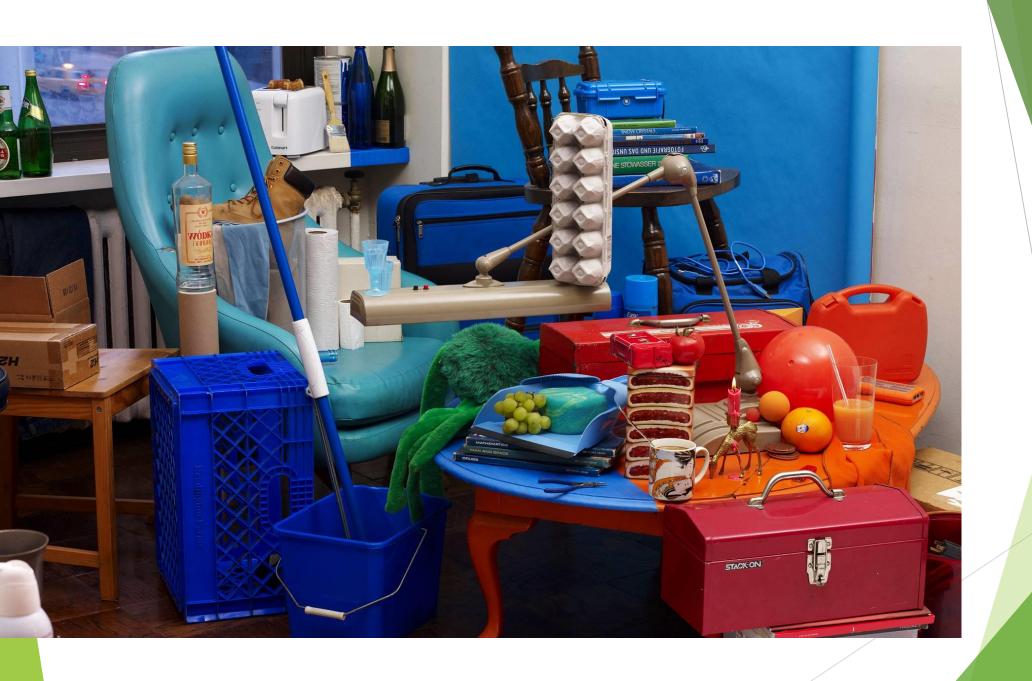
Vision is hard

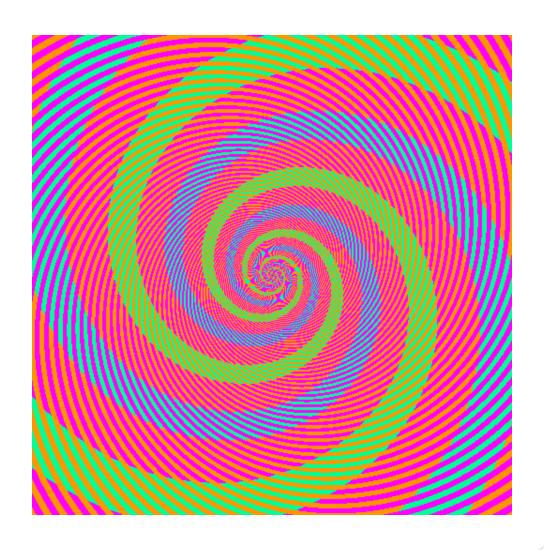
"What we experience, apparently directly, is actually very different from what is recorded by our sense organs."

[Perception: From Sense to Object by J. Wilding]

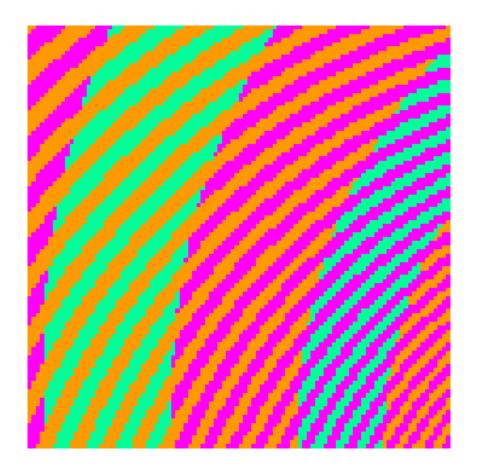
9cal97af64736d708a82808c81948f8e9488717b6d9e98a2989bdl717b7a876d 9b99a4ad687a6e778290878a8380818885818b816895989f9a9bd5647d7d864e a3b495a160656a74787a807d809ea3c580868d7c7785969a989697cf98502f35 ab969b9a5d626c7e807f8f74869ca9b4c9d37b7364819a9b958e978fd5393e38 9e619ch361616e727b7f6b86857aadadc7d2d860698e9a6f908d8ea33635365b 7556959a63686a77816a747781989faac6c6c3e5548c9c497992a12b4740479b 635f96ad5e5673729776828a898cbfc0d3d3clc6e67b9c5527d3773c3f428fa6 626198a655686b77b075767f6eb1adbfbec7c2cecd7c9a3dacd1da3b2b35a3a1 697196b760615d6e9f6f79788aa3aeaca8bbb9c8c4c4d7d8c7c5823f36a3a2a0 6e5e9db25c606865ab7e5f84a58fabaaa29db7a2a7d4c2c3d2822b398f9c9ba2 686balac5d636754b38d749e8f8f839a838a359eb8b9c7c154d732378fa5a09d 6a6996ae66606d6cbe8f827e8c823e763b2fcdb0b8ca6592bc322c70a4a8a6a1 6e5f96ae645963705d72847a424331505abcc2a7c2955fcc343e3c879ea5a293 626c98af686e6d809c60316f6d3283b9646d7ca3ca4848b3453e8ca3a59aa38d 6a6594b76d696fac37356639283fc47e69c7849a83ad37b13630a299999c8e8a 6b6088c1736f7832863b843635bb5f9aaeae898cb57f34a22f90a19a9982acc9 50548fb26e7232318e36328f68a7678ba6b2908e9f893e6c3f969a9a9584bbd2 5d498bb7756d709b9b382d2bb133628a9ca27876b18c3e2f24a29b9284bed4cb 31348dbd6f3d4c2d66852b4035313e3b3789b28e2e376a36998e9d8b7cd1d5b8 813c7ab3702626654366a93f303a366a7b9597c4cb606a4897816b6e93d49f4d 4a9b73b0a932243e6d647b693e3e4a5f8a869bb6c0d4834a978c9073cdd05d7c 2e9e72afa2363227307a71a73c3c6764898e8fa2abcad322988b77bade746a65 27b75dbca1303848265c652f5737806e91849a9fb9cbcf31999280cdd27c6c6c 28b759b8823f4c4b308c412c2b5c7e85928e9494abbbdcde56698b8fc154746f







The blue and green colors are actually the same



http://blogs.discovermagazine.com/badastronomy/2009/06/24/the-blue-and-the-green/

A Few Quotes

- "A breakthrough in machine learning would be worth ten Microsofts" (Bill Gates, Chairman, Microsoft)
- "Machine learning is the next Internet" (Tony Tether, Director, DARPA)
- Machine learning is the hot new thing" (John Hennessy, President, Stanford)
- "Web rankings today are mostly a matter of machine learning" (Prabhakar Raghavan, Dir. Research, Yahoo)
- "Machine learning is going to result in a real revolution" (Greg Papadopoulos, CTO, Sun)
- "Machine learning is today's discontinuity" (Jerry Yang, CEO, Yahoo)



The Ultimate Goal

"If our long-sought quest to create autonomous anthropomorphic automata is to succeed, we must first impart perceptual abilities to machines".

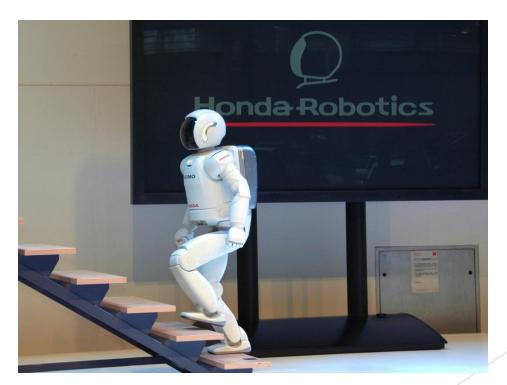
[A Guided Tour of Computer Vision, by V. Nalwa]

Applications - Robot Vision

Ultimately emulating this...



So, how are we doing?



Where we are now:

- https://www.theverge.com/circuitbreaker/2017/11/17/16671328/boston-dynamics-backflip-robot-atlas
- Google image search
 - Demo Search by image

Process of Edge Detection

- Edge detection is the most common approach for detecting meaningful discontinuities.
 - ▶ An edge is a set of connected pixels that lie on the boundary between two regions.
 - ► Edge detection is commonly used for image segmentation in computer vision tasks (e.g. face recognition)
- ► The magnitude of the first derivative in a particular direction can be used to detect the presence of an edge.

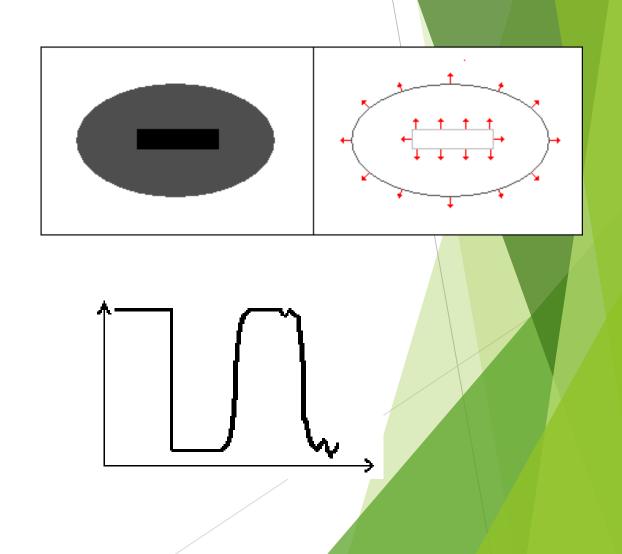


Edge Detection - Topics

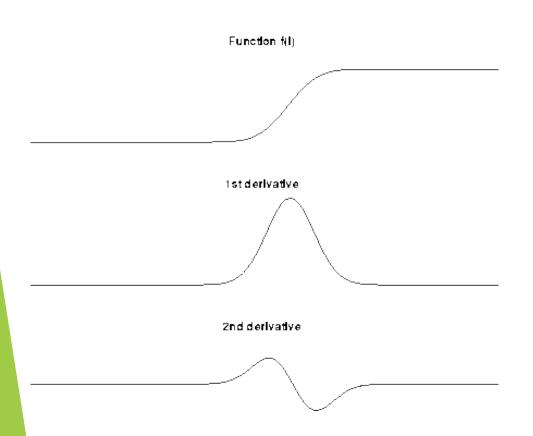
- ▶ 1st derivative edge detection Sobel
- ▶ 2nd derivative edge detection Canny
- Overview of Multispectral edge detection
- ► Application: Image sharpening

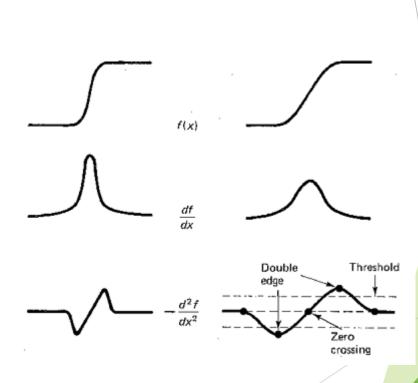
Edge Detection - What is an edge?

- Where brightness changes abruptly
- Edges have
 - Magnitude (Gradient)
 - Direction (Orientation)
- Edge Profiles
 - Step
 - Real
 - Noisy



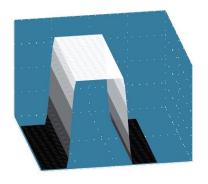
Edge Detection - derivatives

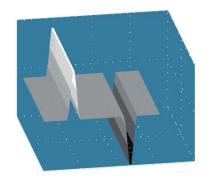


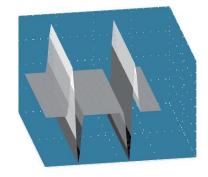


Edge Detection - 1st derivative definitions

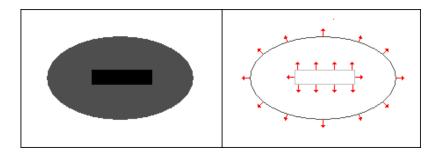








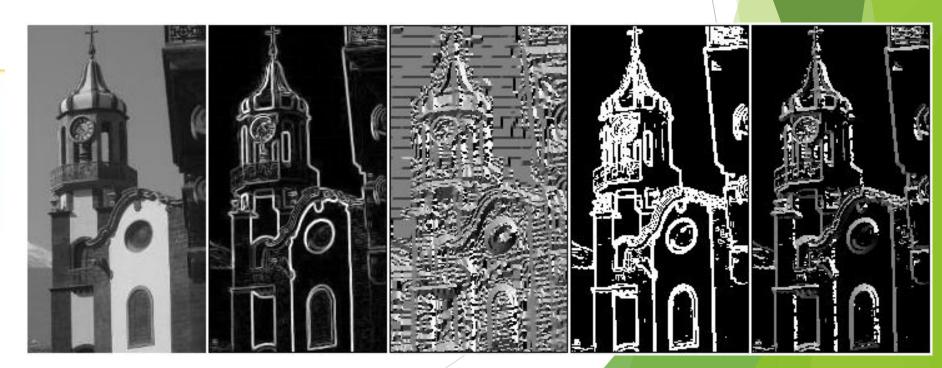
- Recall: Vector variable -
 - Gradient Magnitude
 - Orientation (0 degrees is East)



Edge detection - 1st derivative - Sobel

- The Sobel operator is used to find the approximate absolute gradient magnitude at each point in an input gray-scale image.
- **Sobel detection** uses these two below 3*3 convolution kernels to find edge in an image.

0	0
0	U
2	1
	2 G



Edge detection - 2nd derivative - Canny algorithm

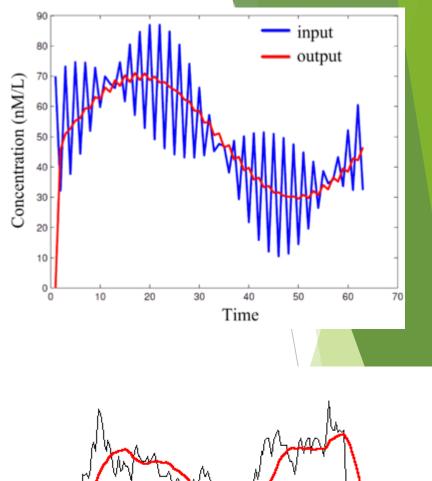
- Canny edge detection improved upon Sobel edge detection by:
 - ▶ Removing speckle noise with a low pass Gaussian filter first
 - ► Then applying a Sobel filter to detect edges
 - ► Then doing non-maximum suppression to pick out the best pixel for edges when there are multiple possibilities in a local neighborhood.
 - Offers more refined edges than Sobel
- ▶ Both can only function with **grey-scale images**

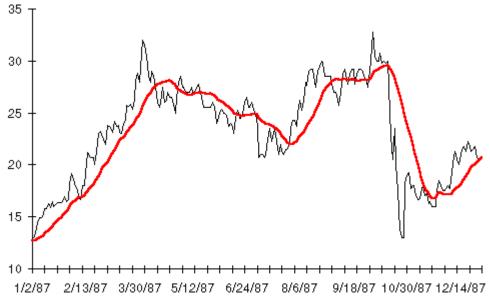


Low Pass Filter

- E.g. Moving average filter
 - https://www.youtube.com/watch?v=ZoaEDbivmOE







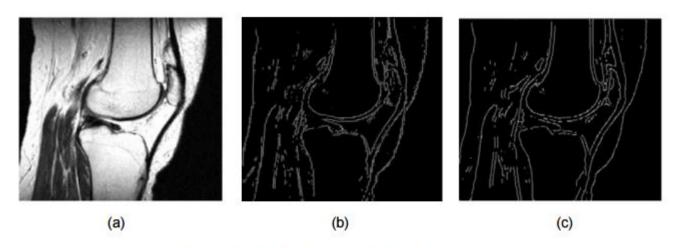


Figure 6: (a) Original image. Sobel (b). Canny (c)

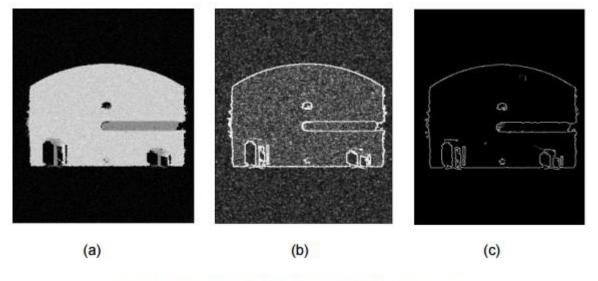
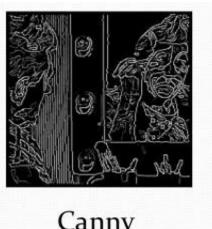


Figure 7: (a) Original image. Sobel (b). Canny (c)



Canny



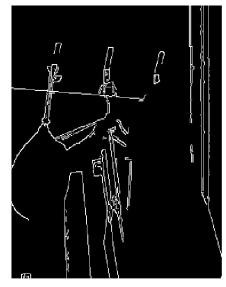
Sobel

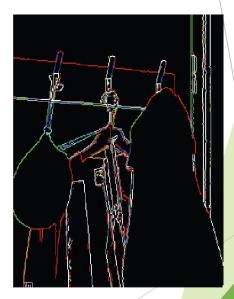
Overview: Multispectral edge detection

- Detect edges separately in each spectral band
 - ▶ Use maximal value OR some linear combination



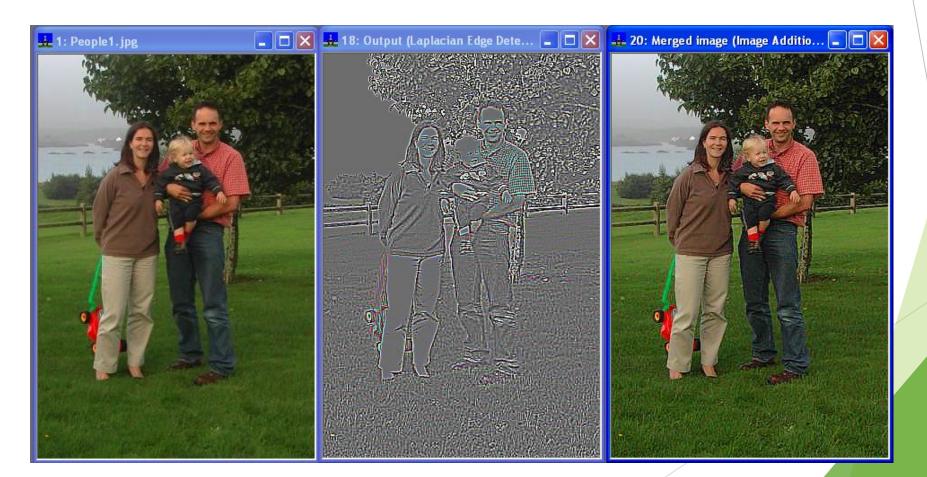






Application: Image sharpening

Making edges steeper.



Application: Image sharpening

▶ Subtract a multiple (e.g. 0.3) of the Laplacian from the image.

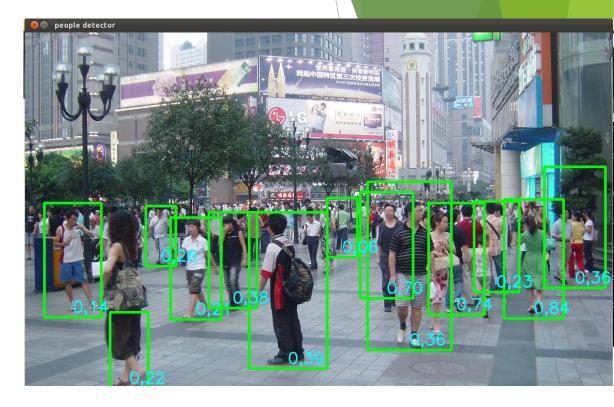






OpenCV

- Open Source Computer Vision is a library of programming functions mainly aimed at real-time computer vision.
- Originally developed by Intel
- OpenCV is released under a BSD license free for commercial use.
- ► Has C++, C, Python and Java interfaces and supports Windows, Linux, Mac OS, iOS and Android.
- Designed for computational efficiency. Written in optimized C/C++, the library can take advantage of multi-core/GPU processing.







Lab this week

- Intro to OpenCV
- Edge detection with Sobel and Canny