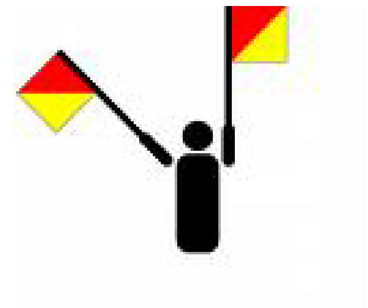
Introduction
Signal representation

Class 1. 30 August 2011

Instructor: Bhiksha Raj

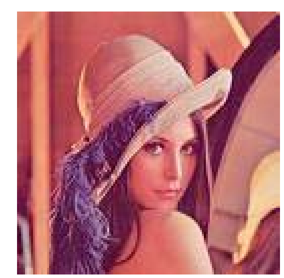
What is a signal

- A mechanism for conveying information
  - Semaphores, gestures, traffic lights..

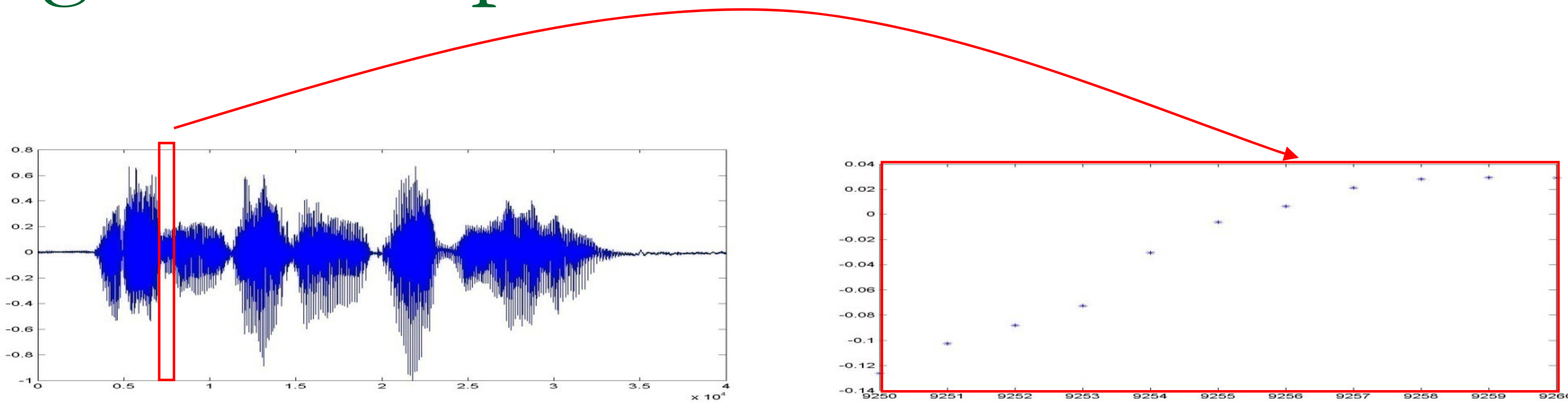


- Electrical engineering: currents, voltages

- Digital signals: Ordered collections of numbers that convey information
  - from a source to a destination
  - about a real world phenomenon
    - Sounds, images

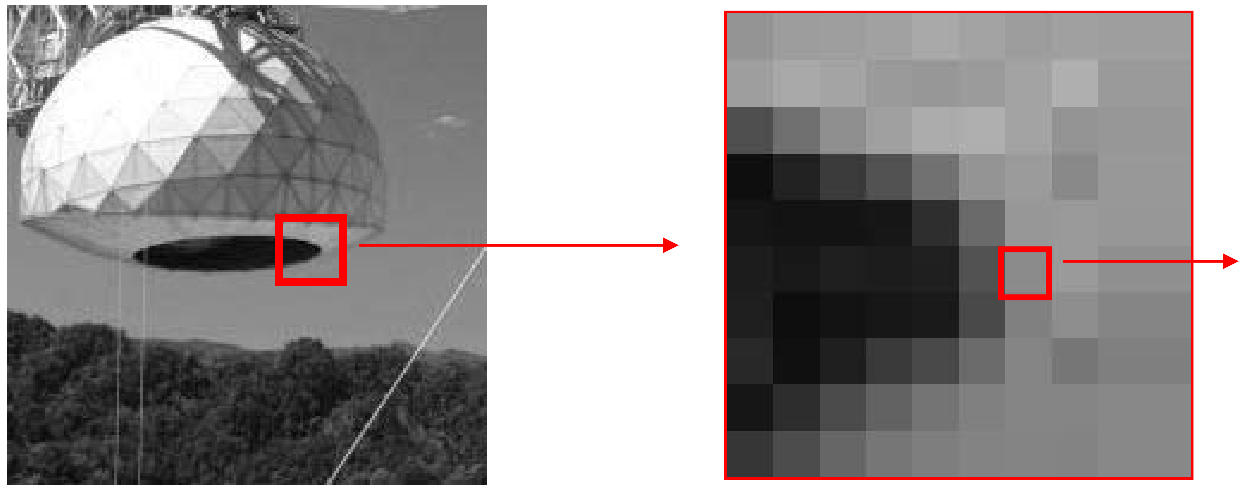


Signal Examples: Audio



- A sequence of numbers
  - [n1 n2 n3 n4 …]
  - The order in which the numbers occur is important
    - Ordered
  - Represent a perceivable sound

Example: Images



- A rectangular arrangement (matrix) of numbers
  - Or sets of numbers (for color images)
- Each pixel represents a visual representation of one of these numbers
  - 0 is minimum / black, 1 is maximum / white
  - Position / order is important

What is Signal Processing

- Analysis, Interpretation, and Manipulation of signals.
  - Decomposition: Fourier transforms, wavelet transforms
  - Denoising signals
  - Coding: GSM, LPC, Mpeg, Ogg Vorbis
  - Detection: Radars, Sonars
  - Pattern matching: Biometrics, Iris recognition, finger print recognition
  - Etc.

What is Machine Learning

- The science that deals with the development of algorithms that can learn from data
  - Learning patterns in data
    - Automatic categorization of text into categories; Market basket analysis
  - Learning to classify between different kinds of data
    - Spam filtering: Valid email or junk?
  - Learning to predict data
    - Weather prediction, movie recommendation
- Statistical analysis and pattern recognition when performed by a computer scientist..

MLSP

- The application of Machine Learning techniques to the analysis of signals such as audio, images and video
- Learning to characterize signals in a data driven manner
  - What are they composed of?
    - Can we automatically deduce that the fifth symphony is composed of notes?
    - Can we segment out components of images?
    - Can we learn the *sparsest* way to represent any signal
- Learning to *detect* signals
  - Radars. Face detection. Speaker verification
- Learning to *recognize* themes in signals
  - Face recognition. Speech recognition.
- Learning to: interpret; optimally represent etc

- In some sense, a combination of signal processing and machine learning
  - But also includes *learning based* methods (as opposed to deterministic methods) for data analysis

MLSP

- IEEE Signal Processing Society has an MLSP committee
  - The Machine Learning for Signal Processing Techinical Committee (MLSP TC) is at the interface between theory and application, developing novel theoretically-inspired methodologies targeting both longstanding and emergent signal processing applications. Central to MLSP is on-line/adaptive nonlinear signal processing and data-driven learning methodologies. Since application domains provide unique problem constraints/assumptions and thus motivate and drive signal processing advances, it is only natural that MLSP research has a broad application base. MLSP thus encompasses new theoretical frameworks for statistical signal processing (e.g. machine learning-based and information-theoretic signal processing), new and emerging paradigms in statistical signal processing (e.g. independent component analysis (ICA), kernel-based methods, cognitive signal processing) and novel developments in these areas specialized to the processing of a variety of signals, including audio, speech, image, multispectral, industrial, biomedical, and genomic signals.

MLSP: Fast growing field

- IEEE Workshop on Machine Learning for Signal Processing
  - Held this year in Beijing. Sep 18-21, http/mlsp2011.conwiz.dk/
- Several special interest groups
  - IEEE : multimedia and audio processing, machine learning and speech processing
  - ACM
  - ISCA
- Books
  - In work: MLSP, P. Smaragdis and B. Raj
- Courses (18797 was one of the first)
- Used everywhere
  - Biometrics: Face recognition, speaker identification
  - User interfaces: Gesture UIs, voice UIs, music retrieval
  - Data capture: OCR,. Compressive sensing
  - Network traffic analysis: Routing algorithms, vehicular traffic..
- Synergy with other topics (text / genome)

In this course

- Jetting through fundamentals:
  - Signal Processing, Linear Algebra, Probability
- Machine learning concepts
  - EM, various relevant estimation and classification techniques

- Sounds:
  - Characterizing sounds
  - Denoising speech
  - Synthesizing speech
  - Separating sounds in mixtures
  - Processing music.

- Images:
  - Characterization
  - Denoising
  - Object detection and recognition
  - Biometrics

- Representation:
  - Transform methods
  - Compressive sensing.

- Topics covered are representative
- Actual list to be covered may change, depending on how the course progresses

Required Background

- DSP
  - Fourier transforms, linear systems, basic statistical signal processing
- Linear Algebra
  - Definitions, vectors, matrices, operations, properties
- Probability
  - Basics: what is an random variable, probability distributions, functions of a random variable
- Machine learning
  - Learning, modelling and classification techniques

Guest Lectures

- Several guest lectures by experts in the topics
  - Alan Black (CMU)
    - Statistical speech synthesis and Voice morphing
  - Fernando de la Torre (CMU)
    - Data representations
  - Marios Savvides
    - Iris recognition
  - Paris Smaragdis (UIUC)
    - Independent component analysis
  - Petros Boufounos (Mitsubishi)
    - Compressive Sensing

Guest Lectures

- Several guest lectures by experts in the topics
  - Rahul Sukhtankar (Google)
    - Music retrieval
  - Mario Berges
    - Load monitoring
  - Roger Dannenberg
    - Music processing
  - Iain Matthews (Disney)
    - Active appearance models
  - John McDonough
    - Microphone arrays

- Subject to change
  - Guest lecturers are notorious for having schedule changes □
  - If the guest lecturer is unavailable, the topic will be covered by me

Schedule of Other Lectures

- Early Lectures (the few weeks)
- Sep 1 : Linear algebra refresher
- Sep 5: More linear algebra
- Sep 8: Project ideas
- Sep 13: Representing sounds and images (DSP)
- Sep 15 : Eigen faces
- Sep 20: Boosting, Face detection
- Sep 22: Expectation Maximization
- Sep 27: Expectation Maximization and Clustering
- Sep 29: Latent Variable Models for Audio

Schedule of Other Lectures

- Early Lectures (the few weeks)
- Oct 4 : Speech Synthesis (Black)
- Oct 6: Latent variable models: Shift invariance etc.
- Oct 11: Iris Recognition (Marios)
- Oct 13: Component Analysis (De La Torre) (2?)
- Oct 18: Linear classifiers and regressions
- Oct 20: Sound Modification, Denoising
- Oct 25: Hidden Markov Models
- Oct 27: NMF, NMF for sounds, images, etc.
- Nov 1: Tracking and prediction: Kalman filters

Schedule of Other Lectures

- Early Lectures (the few weeks)
- Nov 3: Paris Smaragdis Seminar
- Nov 8: Extended Kalman filtering
- Nov 10: Microphone array processing (McDonough)
- Nov 15: Active appearance models (Matthews)
- Nov 17: Bayes nets and Belief Propagation
- Nov 22: Independent Component Analysis
- Nov 24: Thanksgiving (no class)
- Nov 29: Compressive Sensing (Boufounos)
- Dec 1: Music Retrieval (Sukthankar)

Grading

- Homework assignments : 50%
  - Mini projects
  - Will be assigned during course
  - 3 in all
  - You *will not catch up* if you slack on any homework
    - Those who didn't slack will also do the next homework
- Final project: 50%
  - Will be assigned early in course
  - Dec 6: Poster presentation for all projects, with demos (if possible)
    - Partially graded by visitors to the poster

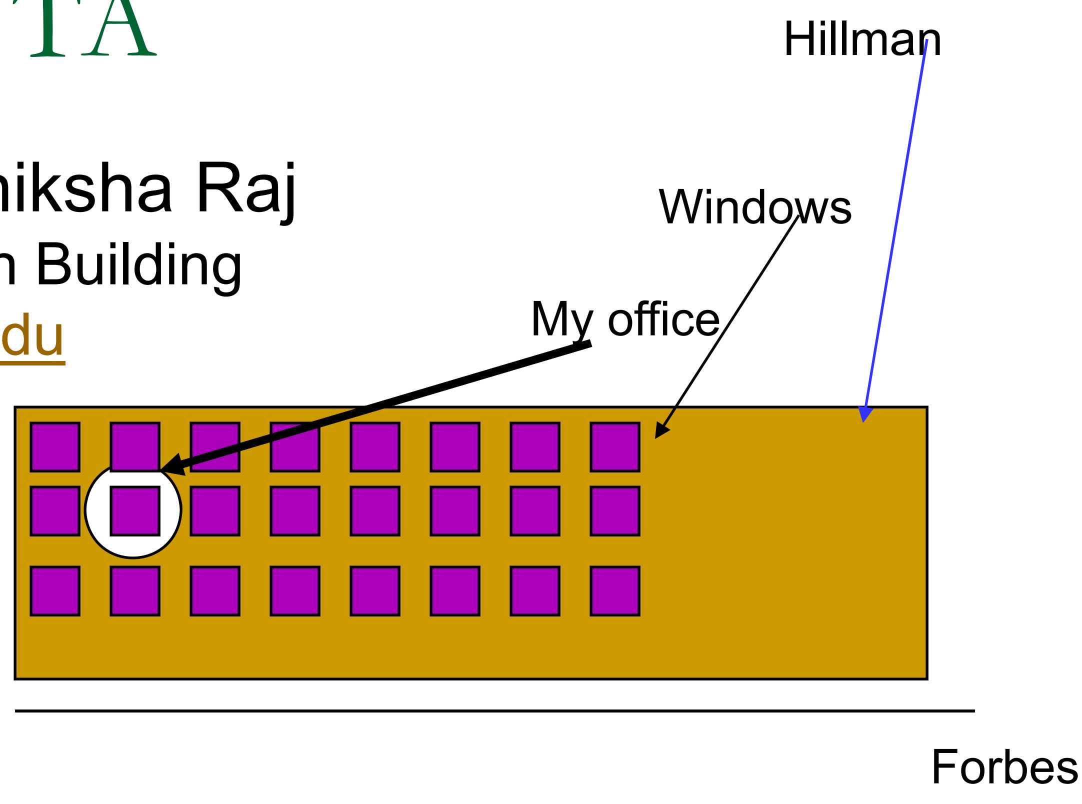
Projects

- 2010 list given as handout
  - Multiple publications and one thesis problem

- 2011: Exciting set of projects
  - Project from NASA
  - Sarnoff Labs
  - Deputy Coroner of Fayette..

Instructor and TA

- Instructor: Prof. Bhiksha Raj
  - Room 6705 Hillman Building
  - bhiksha@cs.cmu.edu
  - 412 268 9826



- TA:
  - Manuel Tragut
  - Anoop Ramakrishna
- Office Hours:
  - Bhiksha Raj: Mon 3:00-4.00
  - TA: TBD
  - Available by email: bhiksha@cs.cmu.edu

Additional Administrivia

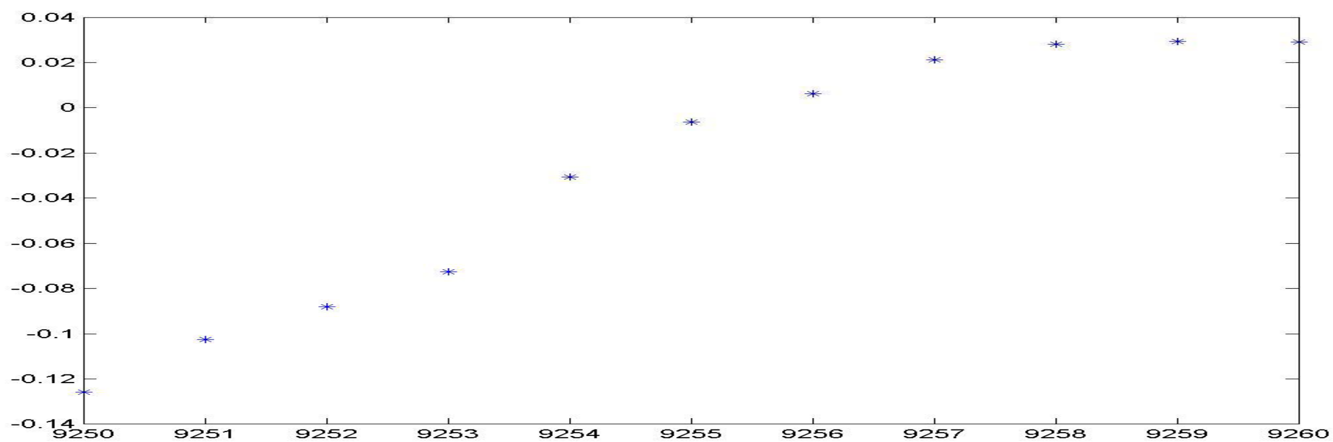
- Website:
  - http://mlsp.cs.cmu.edu/courses/fall2011/
  - Lecture material will be posted on the day of each class on the website
  - Reading material and pointers to additional information will be on the website
- Discussion board
  - blackboard.andrew.cmu.edu/

Representing Data

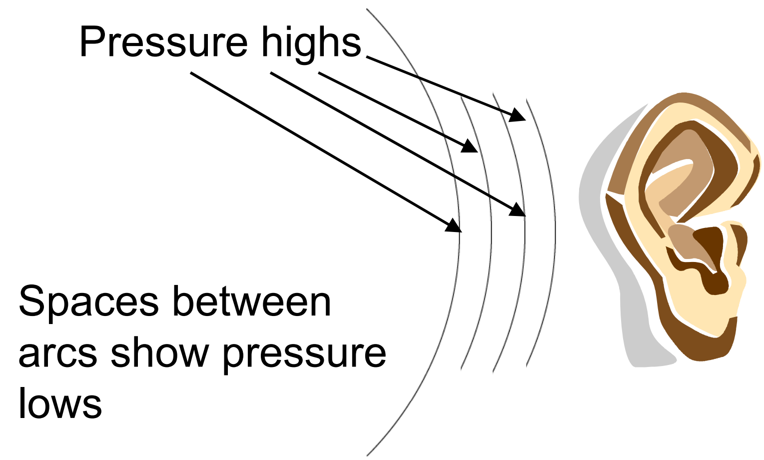
- Audio
- Images
  - Video
- Other types of signals
  - In a manner similar to one of the above

What is an audio signal

- A typical audio signal
  - It's a sequence of points

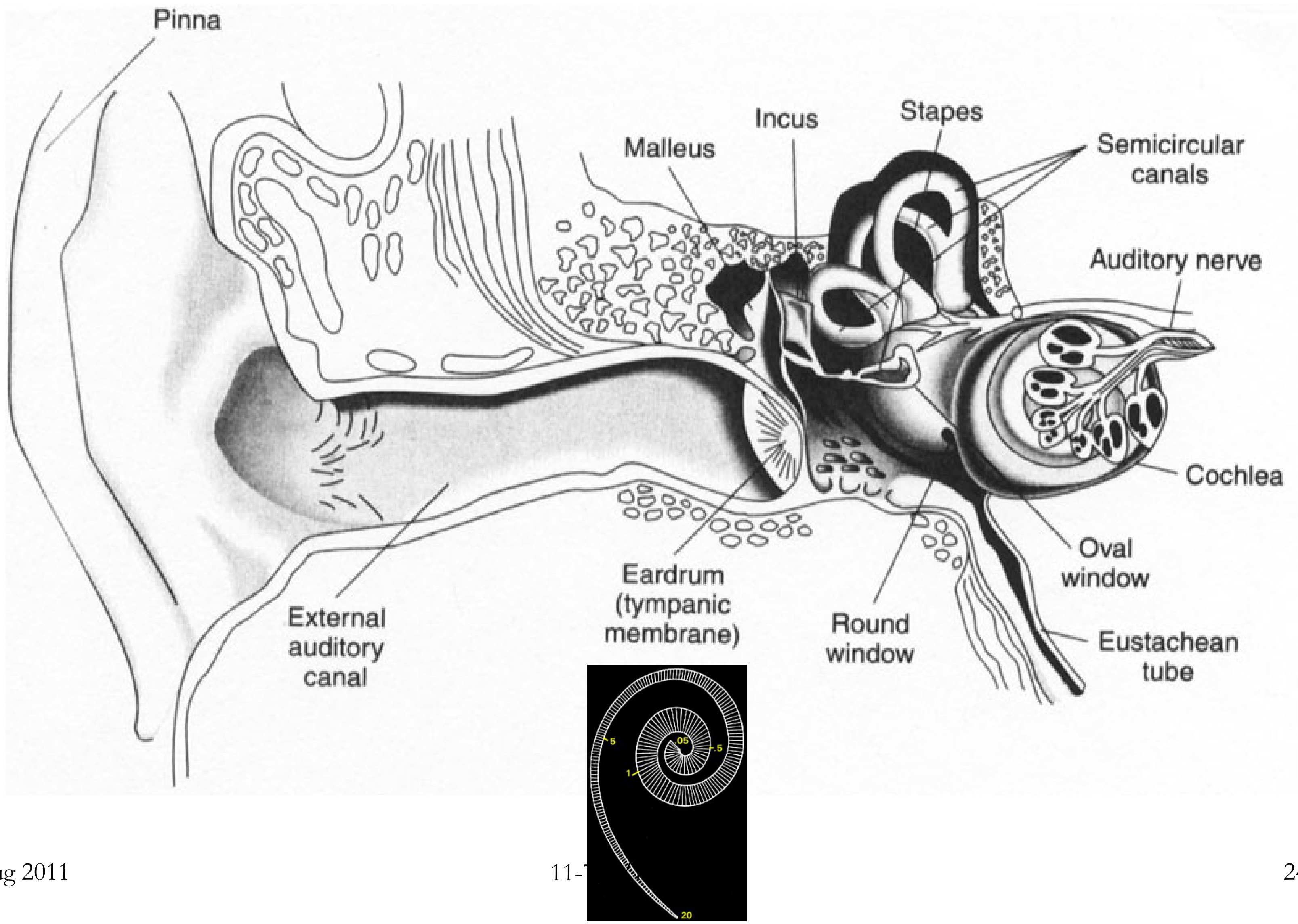


Where do these numbers come from?



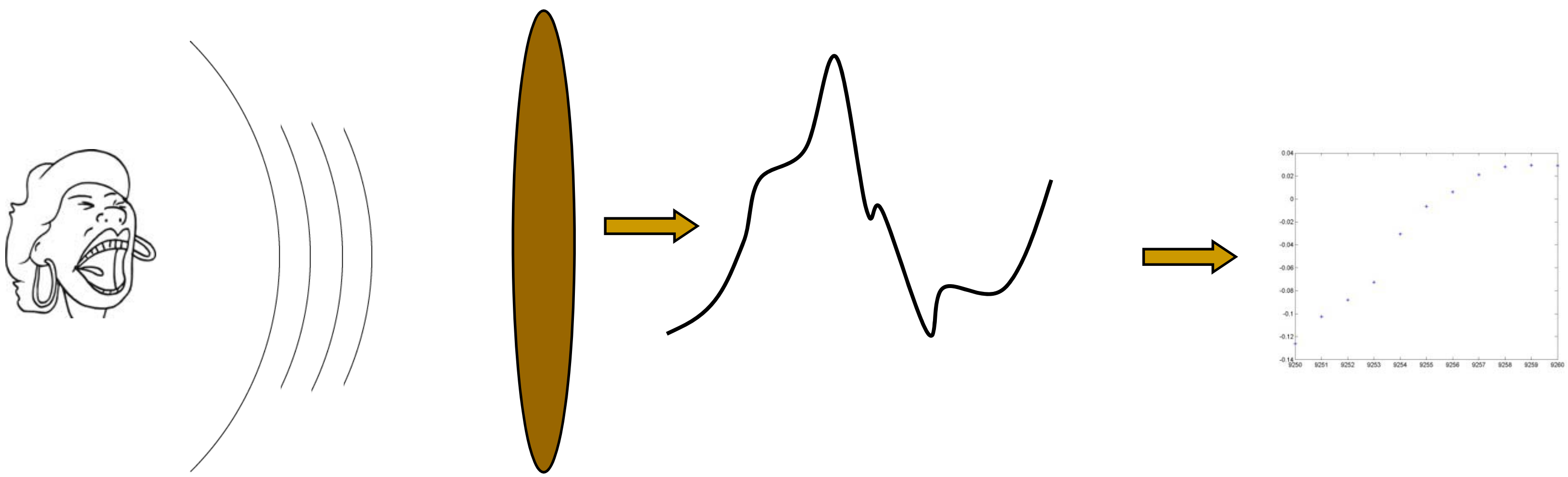
- Any sound is a pressure wave: alternating highs and lows of air pressure moving through the air
- When we speak, we produce these pressure waves
  - Essentially by producing puff after puff of air
  - Any sound producing mechanism actually produces pressure waves
- These pressure waves move the eardrum
  - Highs push it in, lows suck it out
  - We sense these motions of our eardrum as “sound”

SOUND PERCEPTION



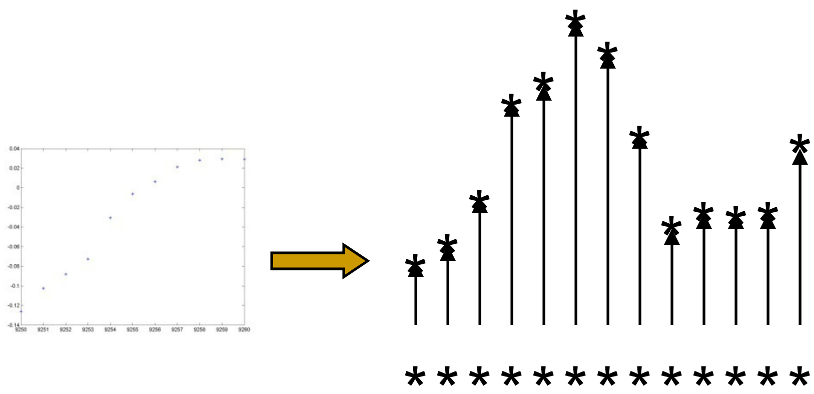
Storing pressure waves on a computer

- The pressure wave moves a diaphragm
  - On the microphone
- The motion of the diaphragm is converted to continuous variations of an electrical signal
  - Many ways to do this
- A "sampler" samples the continuous signal at regular intervals of time and stores the numbers



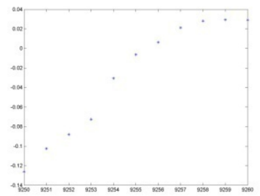
Are these numbers sound?

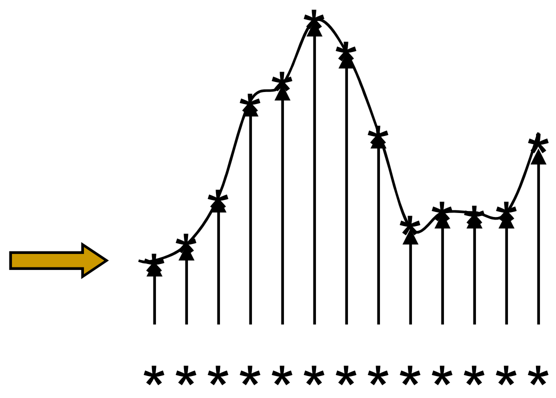
- How do we even know that the numbers we store on the computer have anything to do with speech really?
  - Recreate the sense of sound
- The numbers are used to control the levels of an electrical signal



Are these numbers sound?

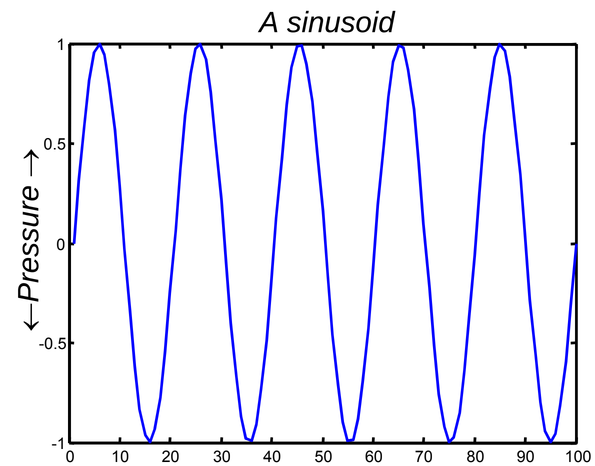
- How do we even know that the numbers we store on the computer have anything to do with speech really?
  - Recreate the sense of sound
- The numbers are used to control the levels of an electrical signal
- The electrical signal moves a diaphragm back and forth to produce a pressure wave
  - That we sense as sound

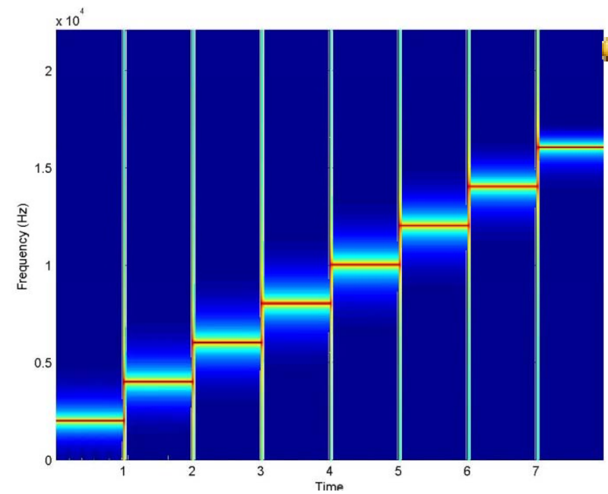




How many samples a second

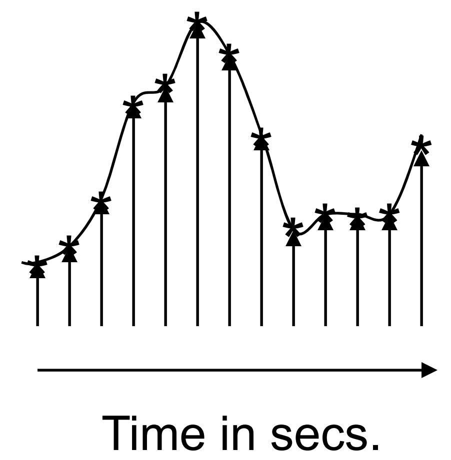
- Convenient to think of sound in terms of sinusoids with frequency
- Sounds may be modelled as the sum of many sinusoids of different frequencies
  - Frequency is a physically motivated unit
  - Each hair cell in our inner ear is tuned to specific frequency
- Any sound has many frequency components
  - We can hear frequencies up to 16000Hz
    - Frequency components above 16000Hz can be heard by children and some young adults
    - Nearly nobody can hear over 20000Hz.





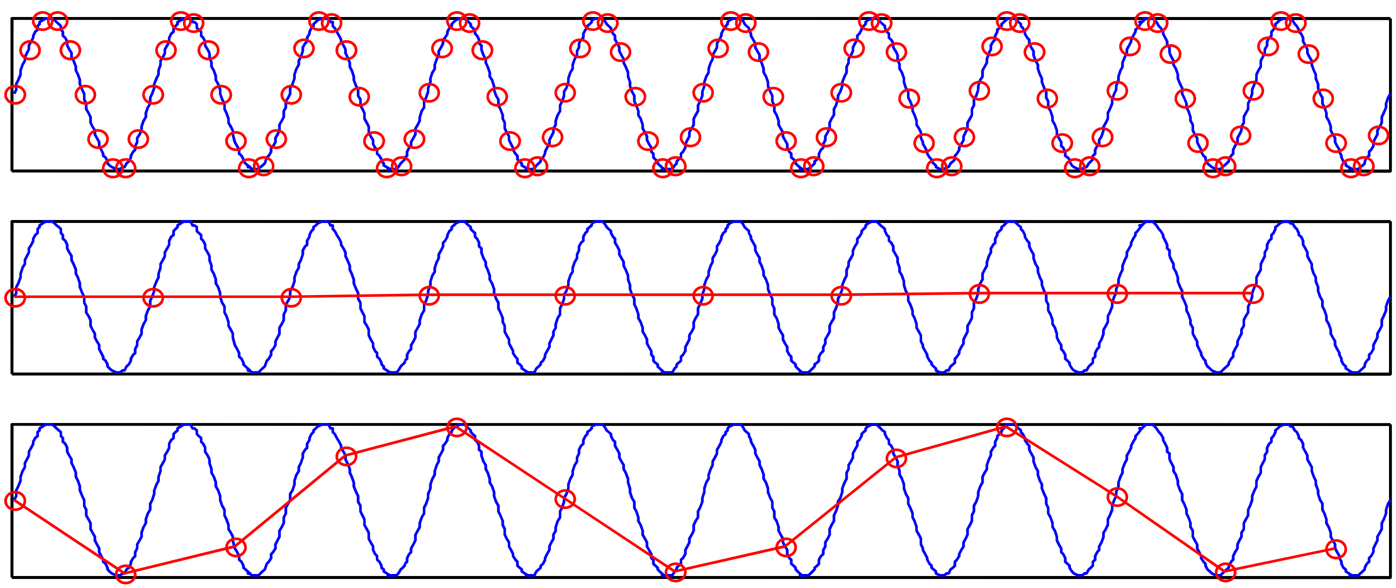
Signal representation - Sampling

- *Sampling frequency* (or *sampling* rate) refers to the number of samples taken a second
- Sampling is measured in Hz
  - We need a sample rate twice as high as the highest frequency we want to represent (Nyquist freq)
- For our ears this means a sample rate of at least 40kHz
  - Cause we hear up to 20kHz
- Common sample rates
  - For speech 8kHz to 16kHz
  - For music 32kHz to 44.1kHz
  - Pro-equipment 96kHz
  - When in doubt use 44.1kHz



Aliasing

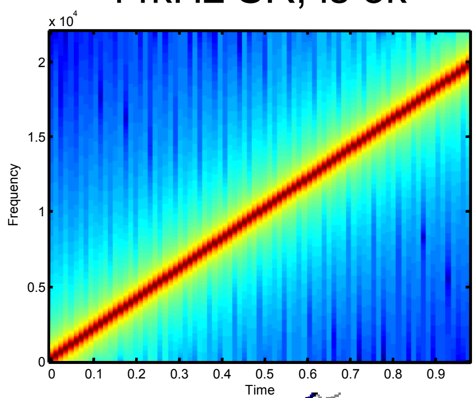
- Low sample rates result in *aliasing*
  - High frequencies are misrepresented
  - Frequency f_1 will become (sample rate – f_1)
  - In video also when you see wheels go backwards



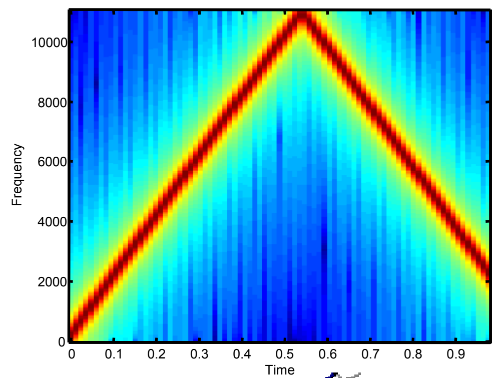
Aliasing examples

Sinusoid sweeping from 0Hz to 20kHz

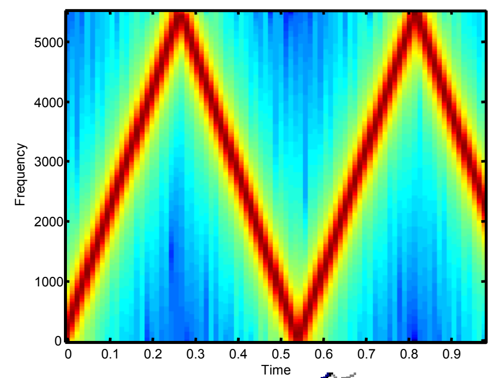
44kHz SR, is ok



22kHz SR, aliasing!



11kHz SR, double aliasing!

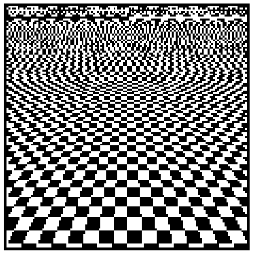


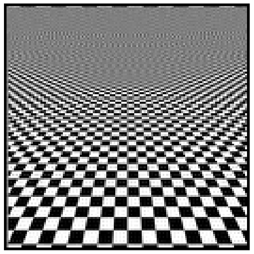
On real sounds

at 44kHz *at 11kHz* *at 4kHz*

at 22kHz *at 5kHz* *at 3kHz*

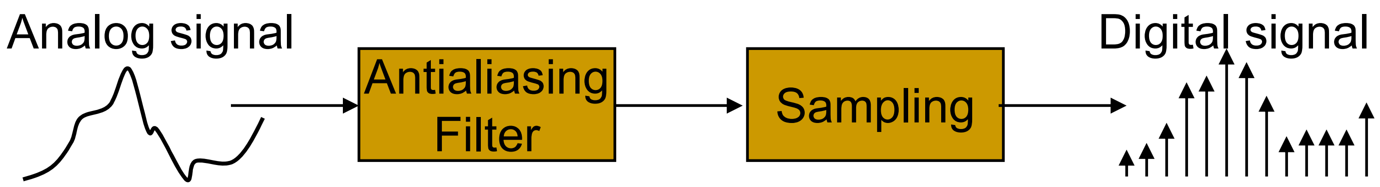
On images





On video

Avoiding Aliasing



- Sound naturally has all perceivable frequencies
  - And then some
  - Cannot control the rate of variation of pressure waves in nature
- Sampling at *any* rate *will* result in aliasing
- Solution: *Filter the electrical signal* before sampling it
  - Cut off all frequencies above samplingfrequency/2
  - E.g., to sample at 44.1Khz, filter the signal to eliminate all frequencies above 22050 Hz

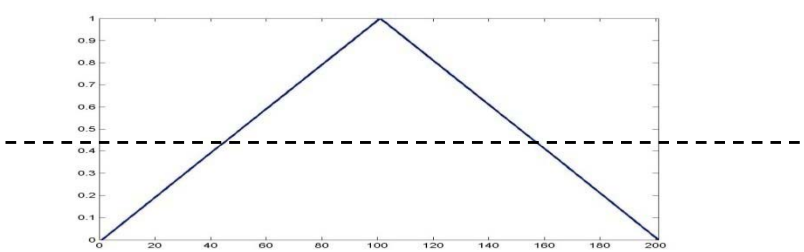
Storing numbers on the Computer

- Sound is the outcome of a continuous range of variations
  - The pressure wave can take any value (within limit)
  - The diaphragm can also move continuously
  - The electrical signal from the diaphragm has continuous variations
- A computer has finite resolution
  - Numbers can only be stored to finite resolution
  - E.g. a 16-bit number can store only 65536 values, while a 4-bit number can store only 16 values
  - To store the sound wave on the computer, the continuous variation must be "mapped" on to the discrete set of numbers we can store

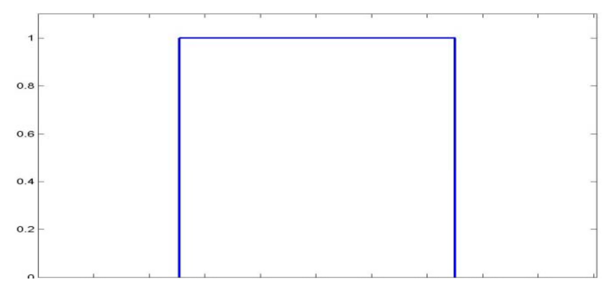
Mapping signals into bits

- Example of 1-bit sampling table

Signal Value	Bit sequence	Mapped to
S > 2.5v	1	1 * const
S <=2.5v	0	0



Original Signal

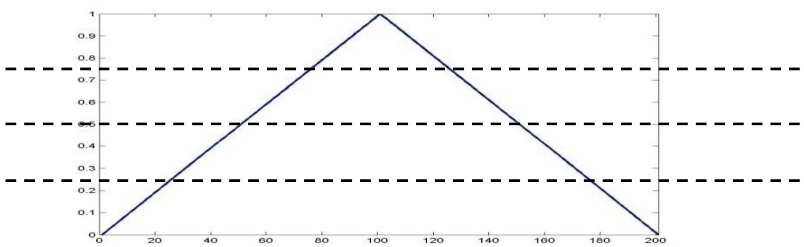


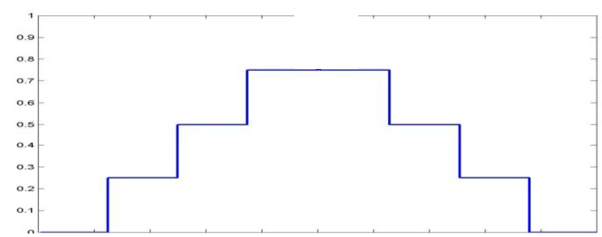
Quantized approximation

Mapping signals into bits

- Example of 2-bit sampling table

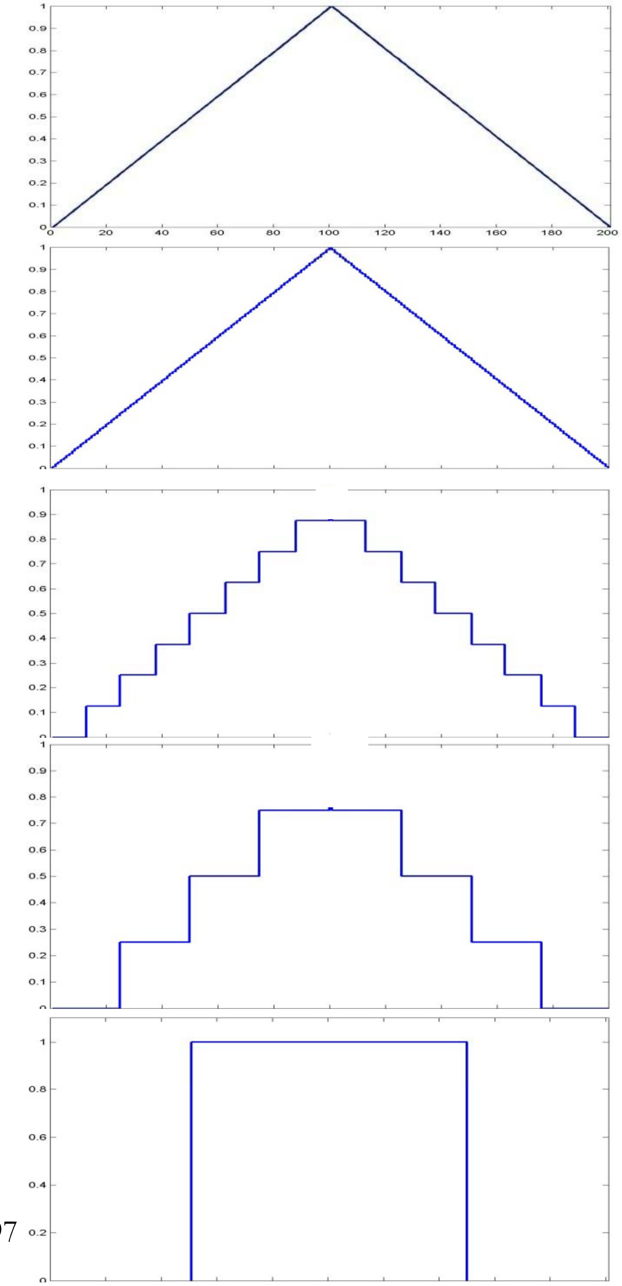
Signal Value	Bit sequence	Mapped to
S >= 3.75v	11	3 * const
3.75v > S >= 2.5v	10	2 * const
2.5v > S >= 1.25v	01	1 * const
1.25v > S >= 0v	0	0


Original Signal


Quantized approximation

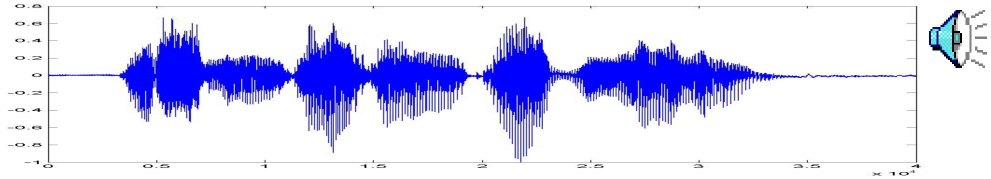
Storing the signal on a computer

- The original signal
- 8 bit quantization
- 3 bit quantization
- 2 bit quantization
- 1 bit quantization

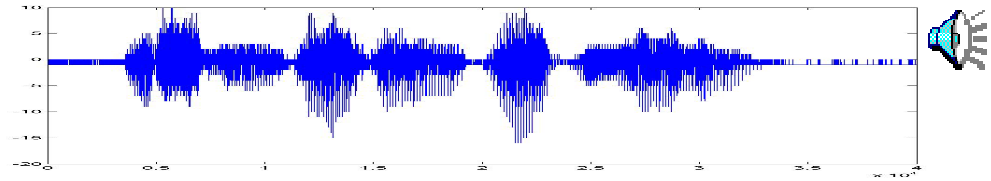


Tom Sullivan Says his Name

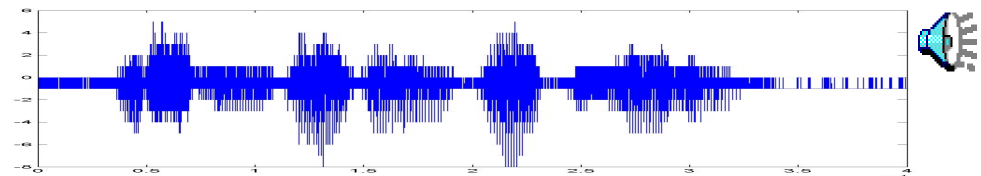
- 16 bit sampling



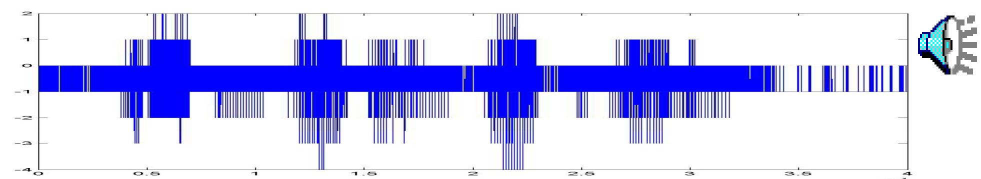
- 5 bit sampling



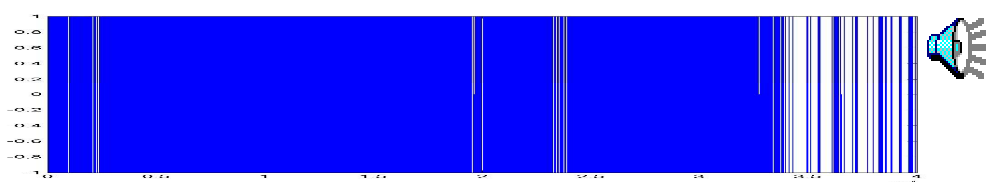
- 4 bit sampling



- 3 bit sampling

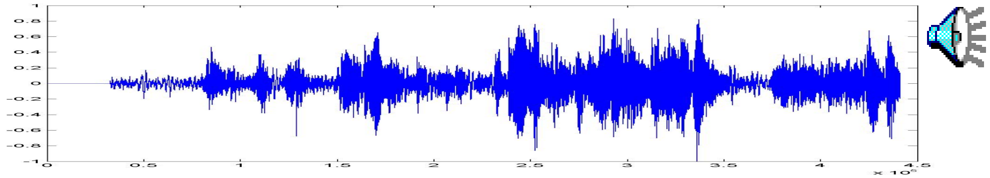


- 1 bit sampling

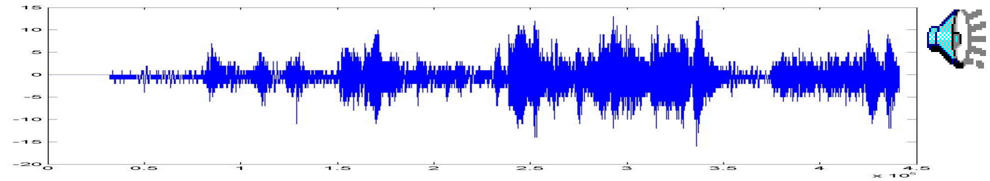


A Schubert Piece

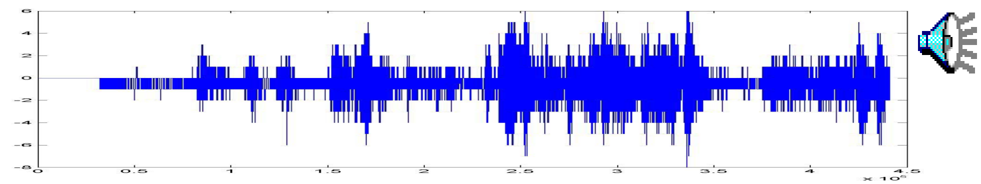
- 16 bit sampling



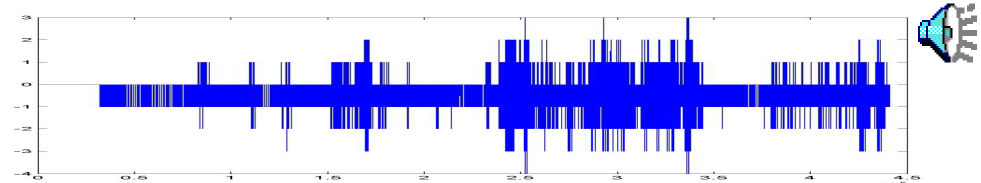
- 5 bit sampling



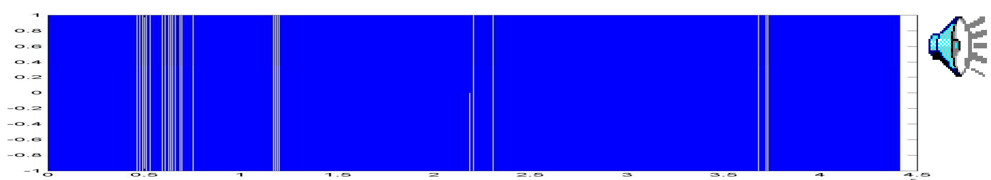
- 4 bit sampling



- 3 bit sampling



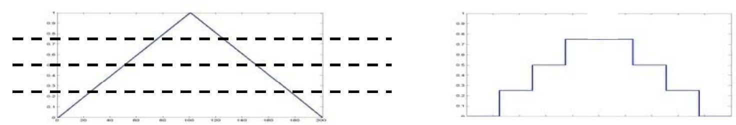
- 1 bit sampling



Quantization Formats

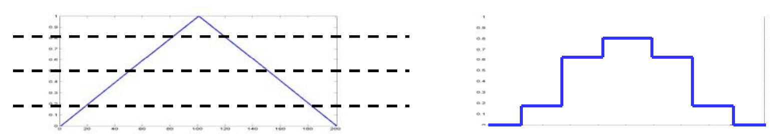
- Sampling can be uniform
  - Sample values equally spaced out

Signal Value	Bits	Mapped to
S >= 3.75v	11	3 * const
3.75v > S >= 2.5v	10	2 * const
2.5v > S >= 1.25v	01	1 * const
1.25v > S >= 0v	0	0

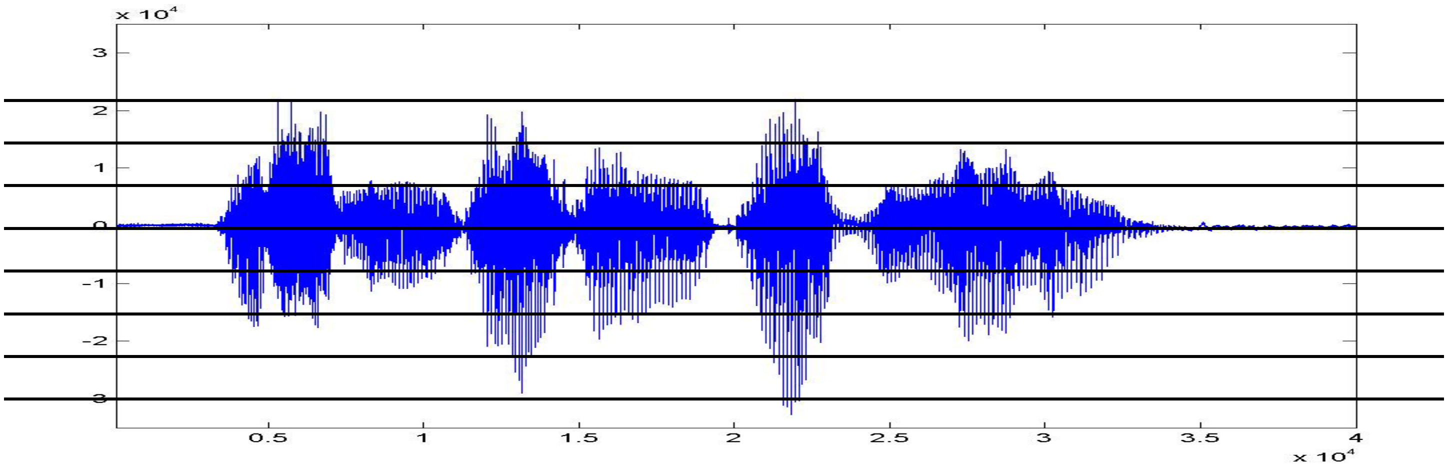


- Or nonuniform

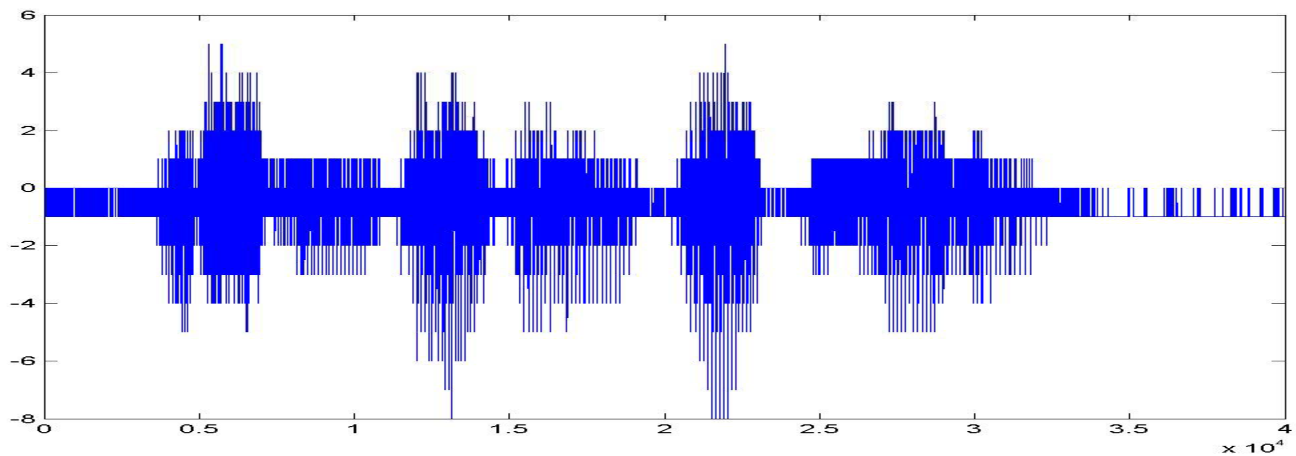
Signal Value	Bits	Mapped to
S >= 4v	11	4.5 * const
4v > S >= 2.5v	10	3.25 * const
2.5v > S >= 1v	01	1.25 * const
1.0v > S >= 0v	0	0.5 * const



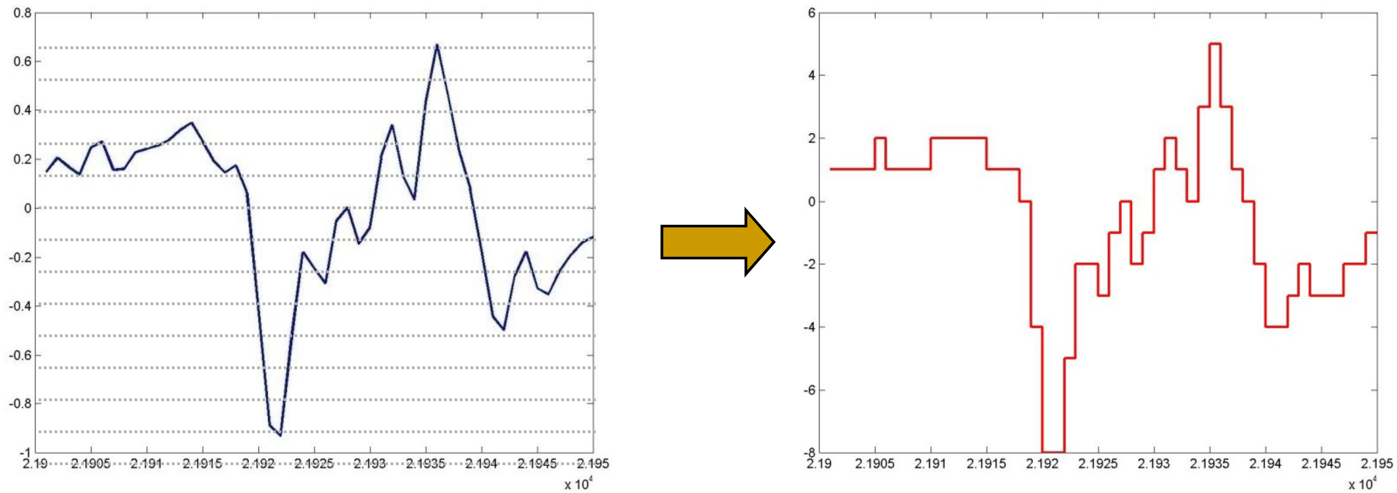
Uniform Quantization



UPON BEING SAMPLED AT ONLY 3 BITS (8 LEVELS)

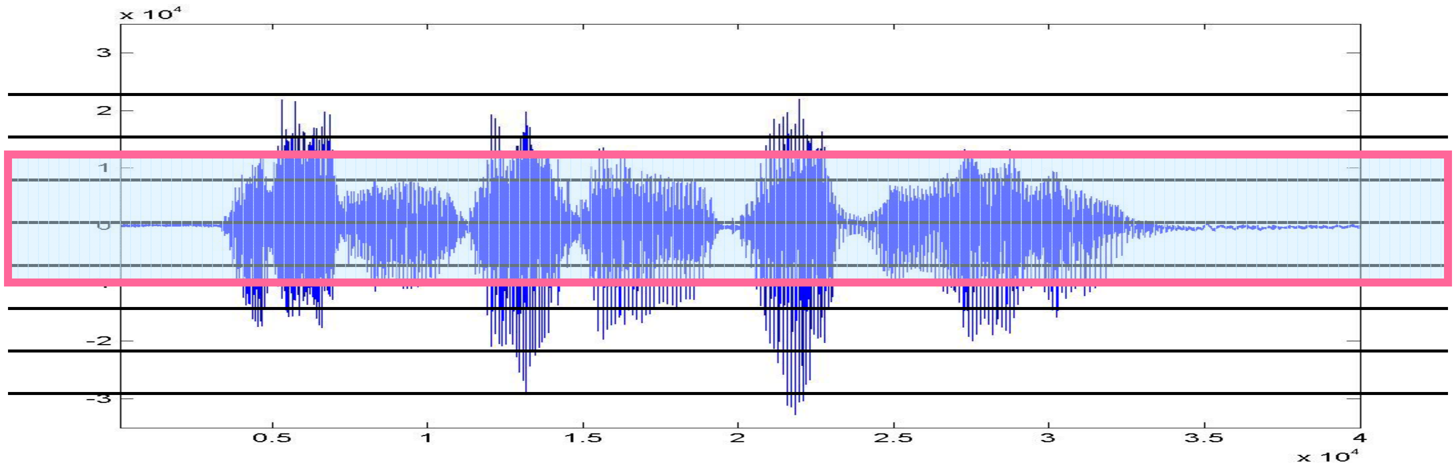


Uniform Quantization



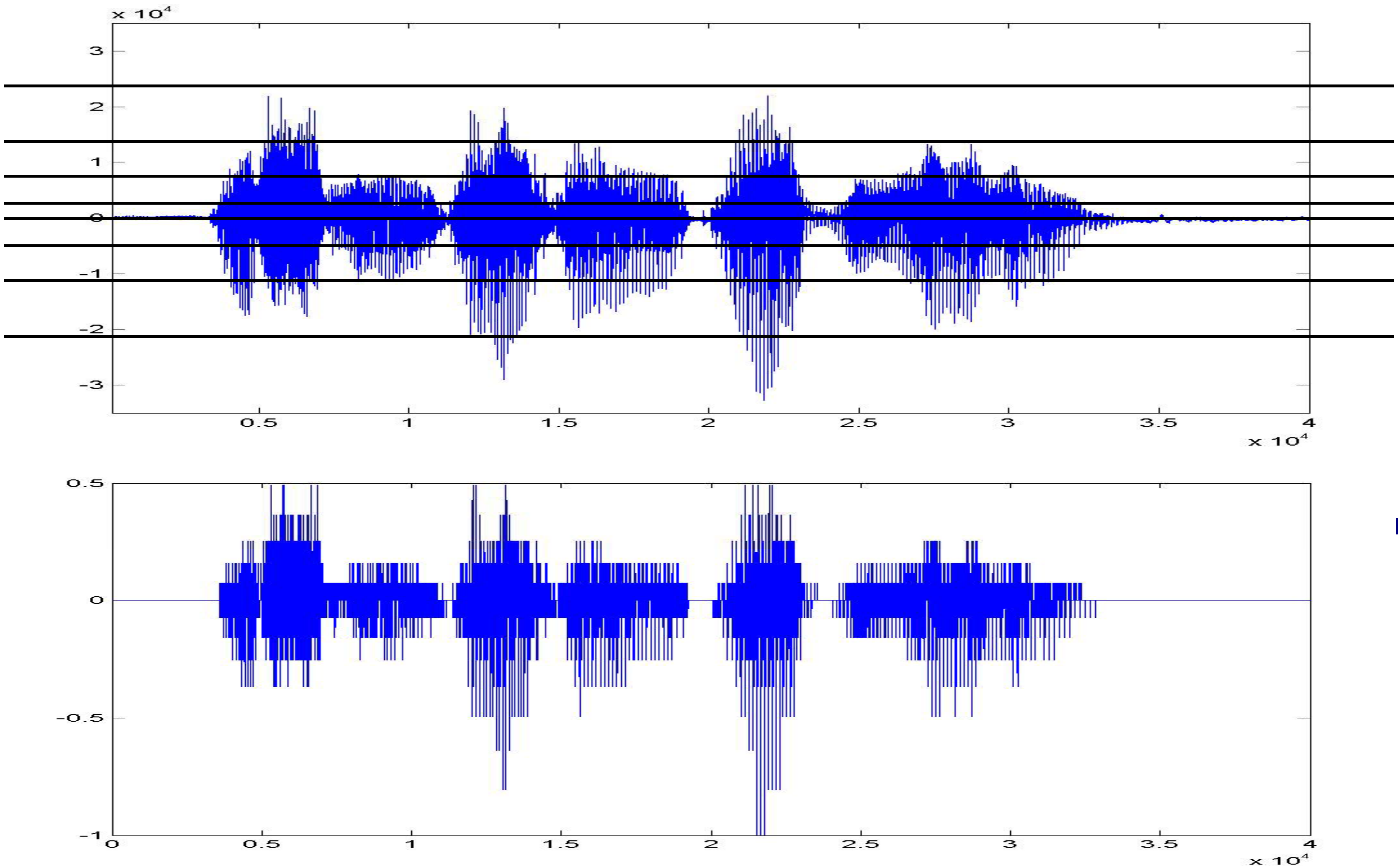
- At the sampling instant, the actual value of the waveform is rounded off to the nearest level permitted by the quantization
- Values entirely outside the range are quantized to either the highest or lowest values

Uniform Quantization



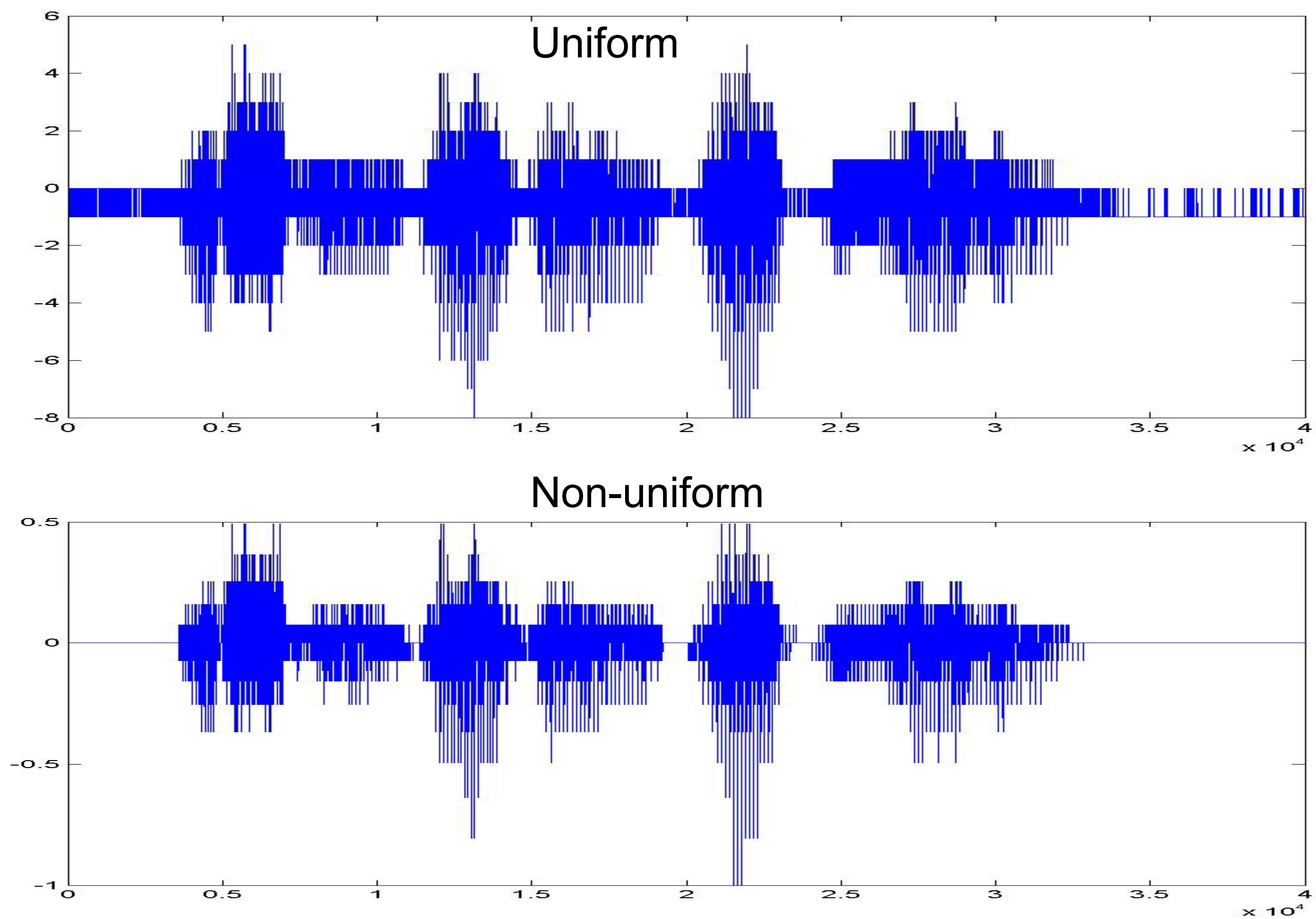
- There is a lot more action in the central region than outside.
- Assigning only four levels to the busy central region and four entire levels to the sparse outer region is inefficient
- Assigning more levels to the central region and less to the outer region can give better fidelity
  - for the same storage

Non-uniform Quantization



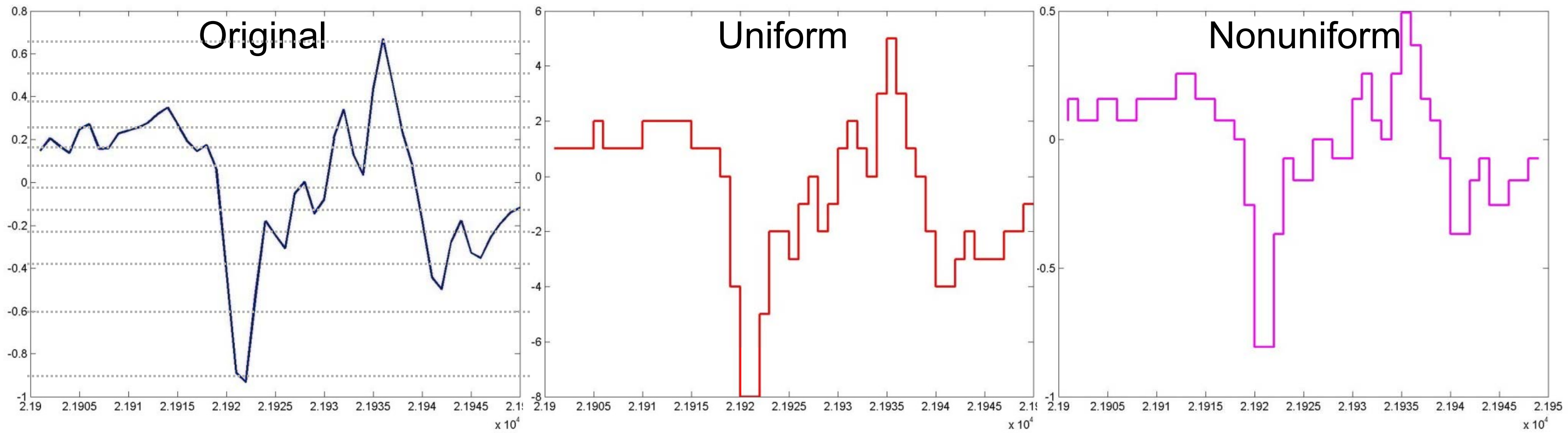
- Assigning more levels to the central region and less to the outer region can give better fidelity for the same storage

Non-uniform Quantization



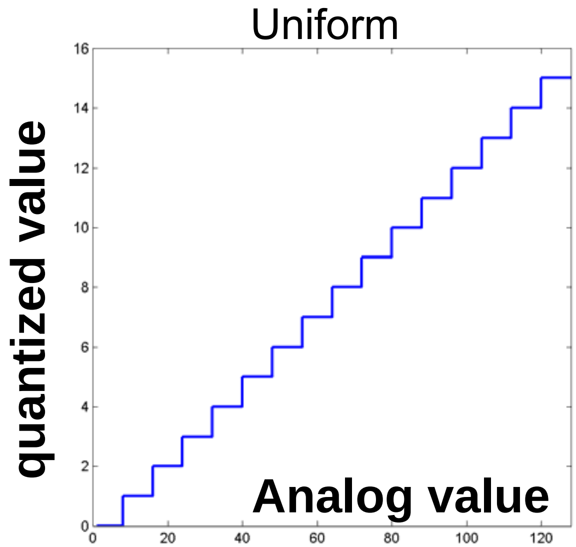
- Assigning more levels to the central region and less to the outer region can give better fidelity for the same storage

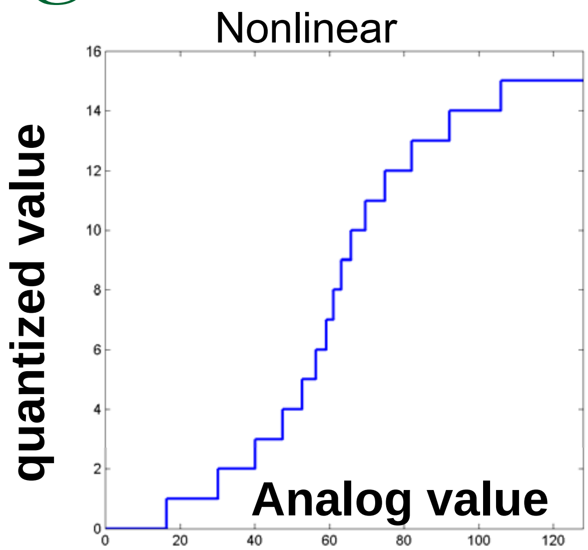
Non-uniform Sampling



- At the sampling instant, the actual value of the waveform is rounded off to the nearest level permitted by the quantization
- Values entirely outside the range are quantized to either the highest or lowest values

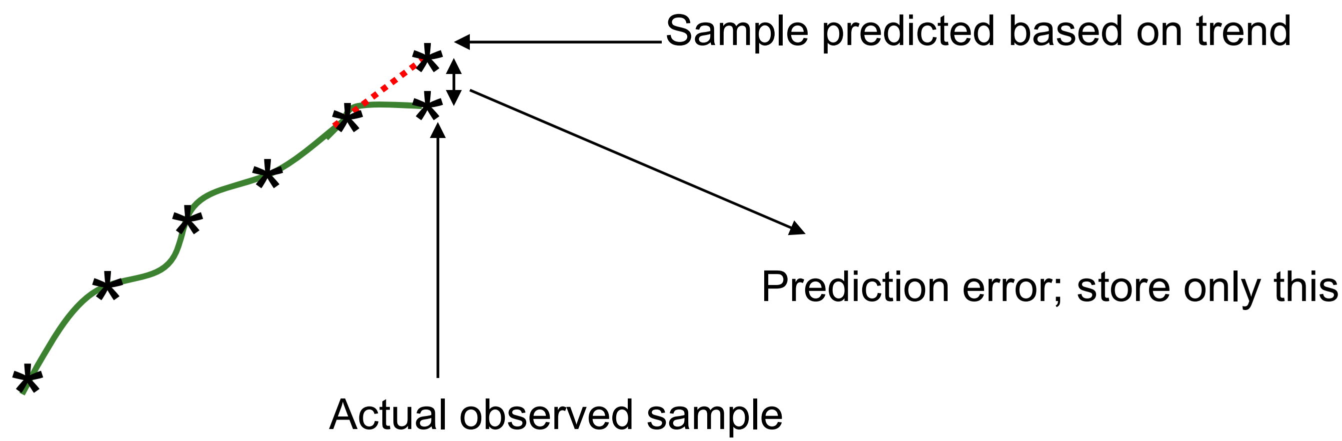
Non-uniform Sampling





- Uniform sampling maps uniform widths of the analog signal to units steps of the quantized signal
- In non-uniform sampling the step sizes are smaller near 0 and wider farther away
  - The curve that the steps are drawn on follow a logarithmic law:
    - Mu Law: $Y = C. \log(1 + \mu X/C)/(1+\mu)$
    - A Law: $Y = C. (1 + \log(a.X)/C)/(1+a)$
- One can get the same perceptual effect with 8bits of non-linear sampling as 12bits of linear sampling

Storage based on prediction



- "Predict" the next sample and store the difference between the value we predict and what we actually see using a small number of bits
- To reconstruct, predict the next sample and add the stored difference back in
- Variety of formats: DPCM, ADPCM.
- Coding schemes: LPC based methods (G728,G729), Mpeg, Ogg Vorbis, …

Dealing with audio

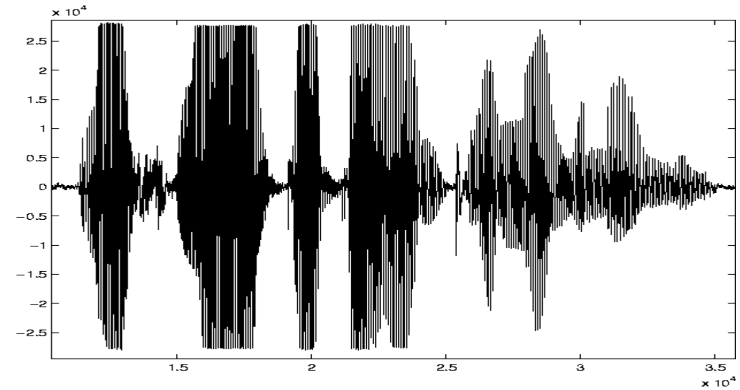
Signal Value	Bits	Mapped to
S >= 3.75v	11	3
3.75v > S >= 2.5v	10	2
2.5v > S >= 1.25v	01	1
1.25v > S >= 0v	0	0

Signal Value	Bits	Mapped to
S >= 4v	11	4.5
4v > S >= 2.5v	10	3.25
2.5v > S >= 1v	01	1.25
1.0v > S >= 0v	0	0.5

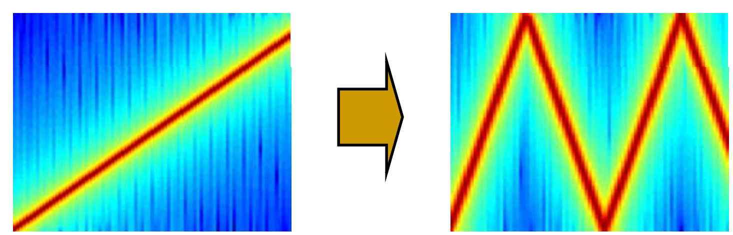
- Capture / read audio in the format provided by the file or hardware
  - Linear PCM, Mu-law, A-law, Coded
- Convert to 16-bit PCM value
  - I.e. map the bits onto the number on the right column
  - This mapping is typically provided by a table computed from the sample compression function
  - No lookup for data stored in PCM
- Conversion from Mu law:
  - http://www.speech.cs.cmu.edu/comp.speech/Section2/Q2.7.html

Common Audio Capture Errors

- Gain/Clipping: High gain levels in A/D can result in distortion of the audio

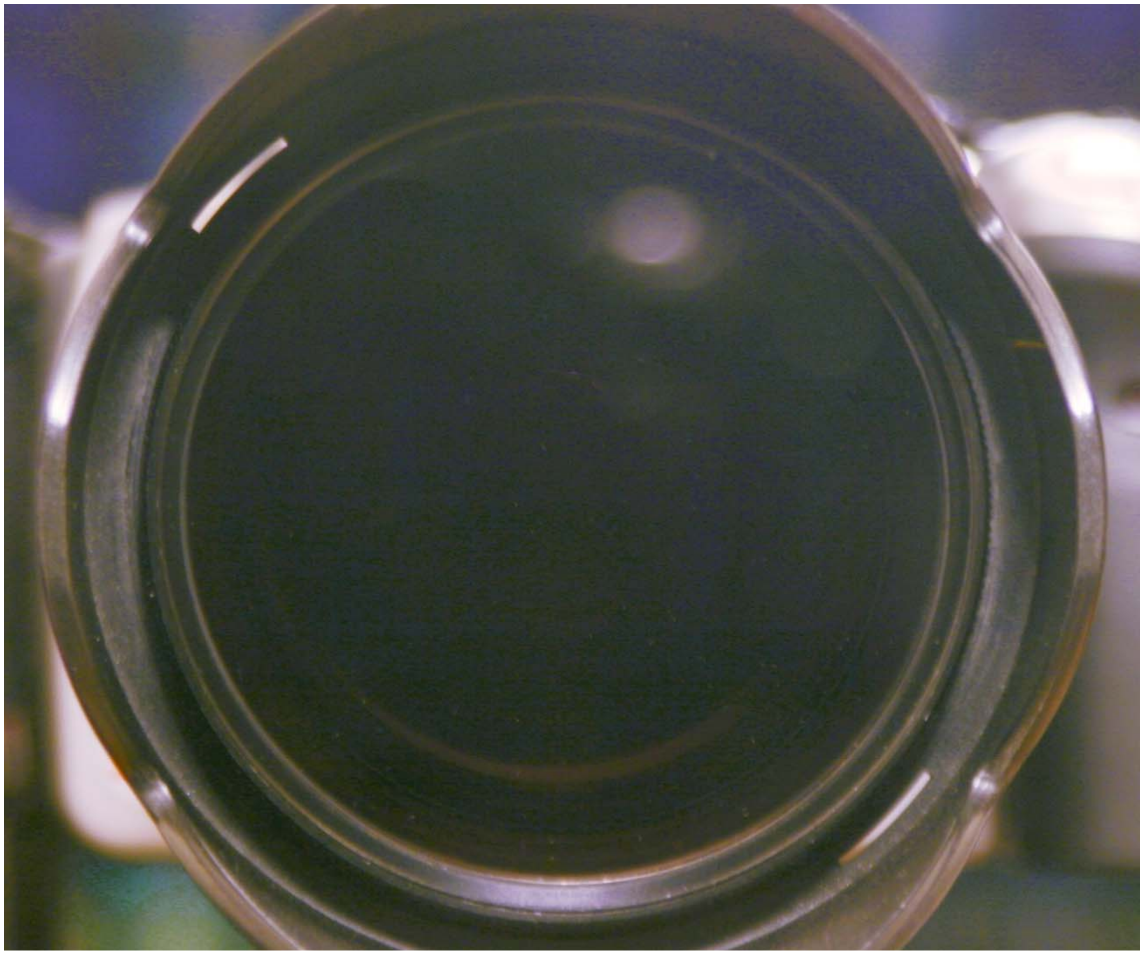


- Antialiasing: If the audio is sampled at N kHz, it *must* first be low-pass filtered at < N/2 kHz
  - Otherwise high-frequency components will alias into lower frequencies and distort them

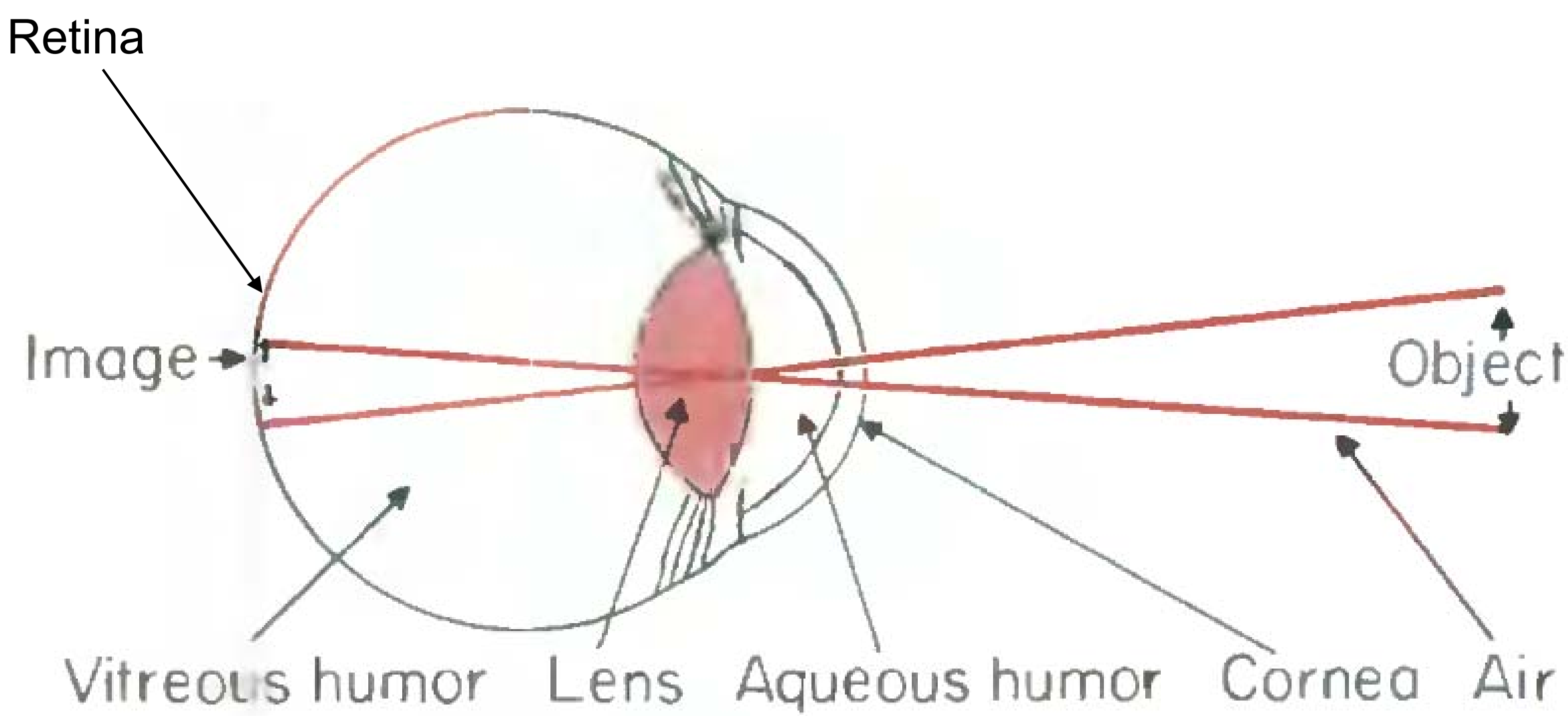


Images

Images

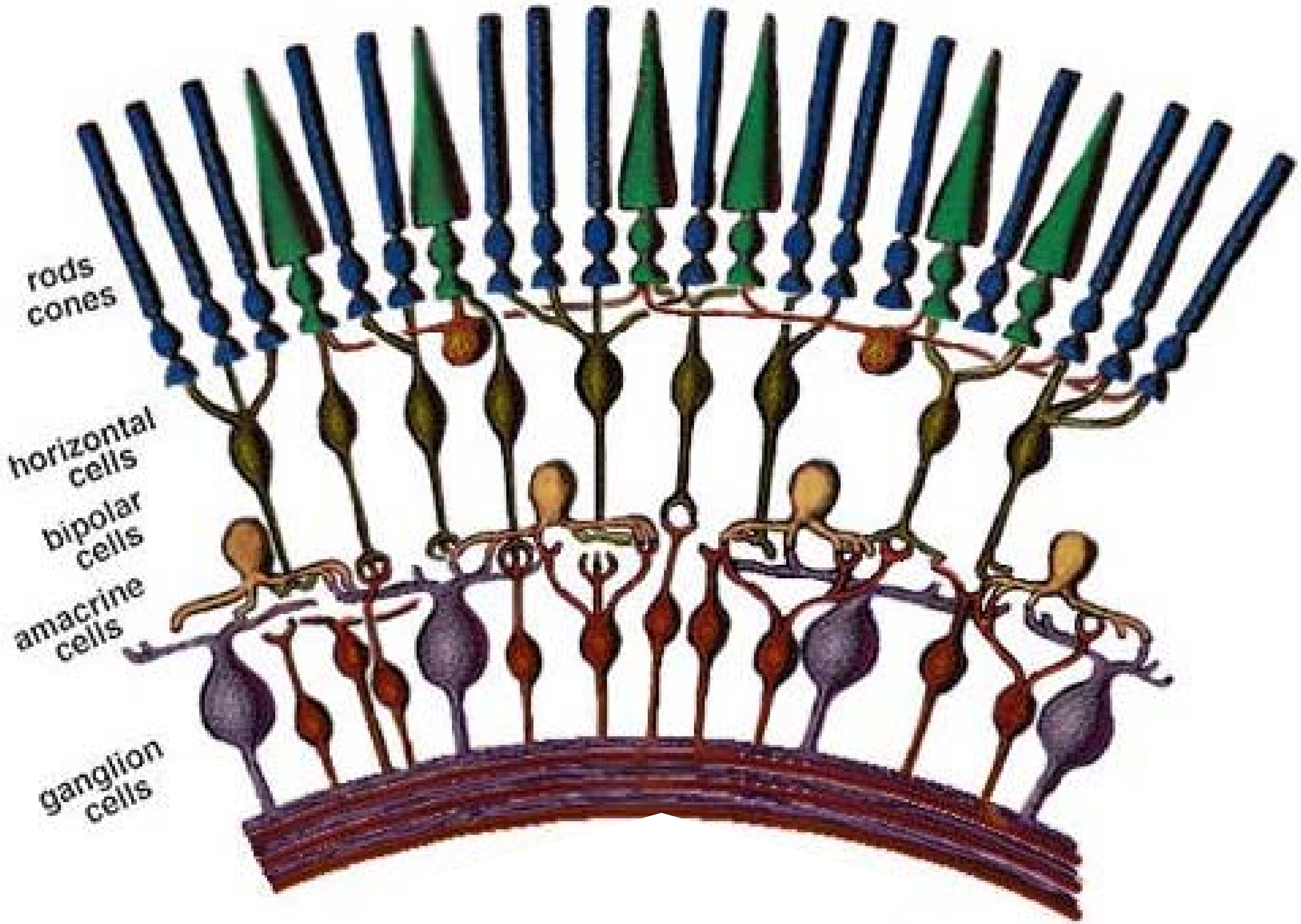


The Eye



Basic Neuroscience: Anatomy and Physiology Arthur C. Guyton, M.D. 1987 W.B.Saunders Co.

The Retina



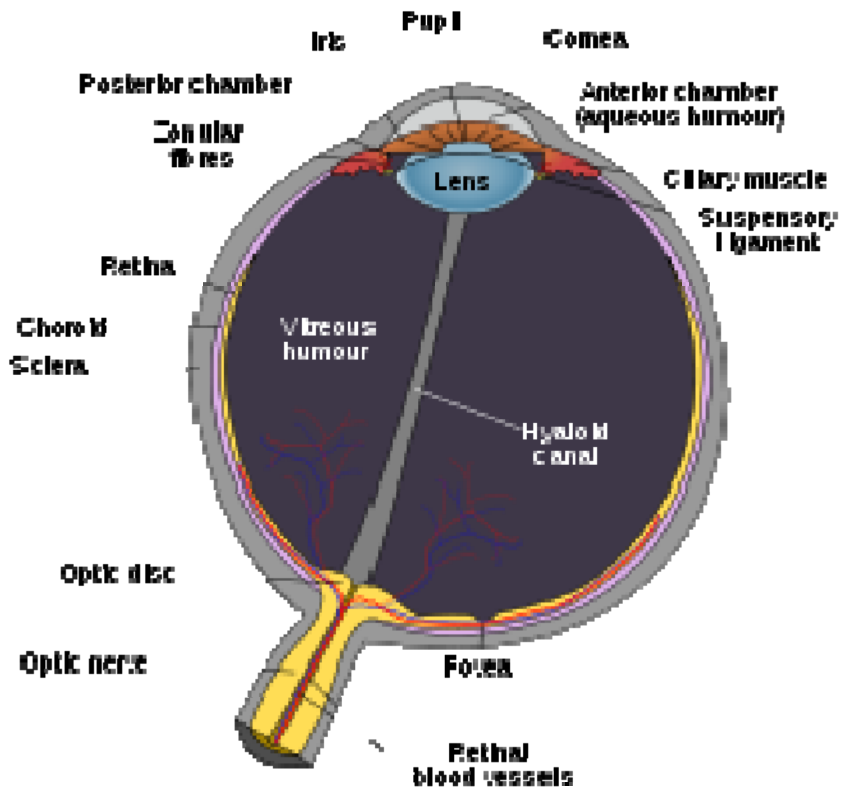
http://www.brad.ac.uk/acad/lifesci/optometry/resources/modules/stage1/pvp1/Retina.html

Rods and Cones

- Separate Systems
- Rods
  - Fast
  - Sensitive
  - predominate in the periphery
- Cones
  - Slow
  - Not so sensitive
  - Fovea / Macula
  - **COLOR**!

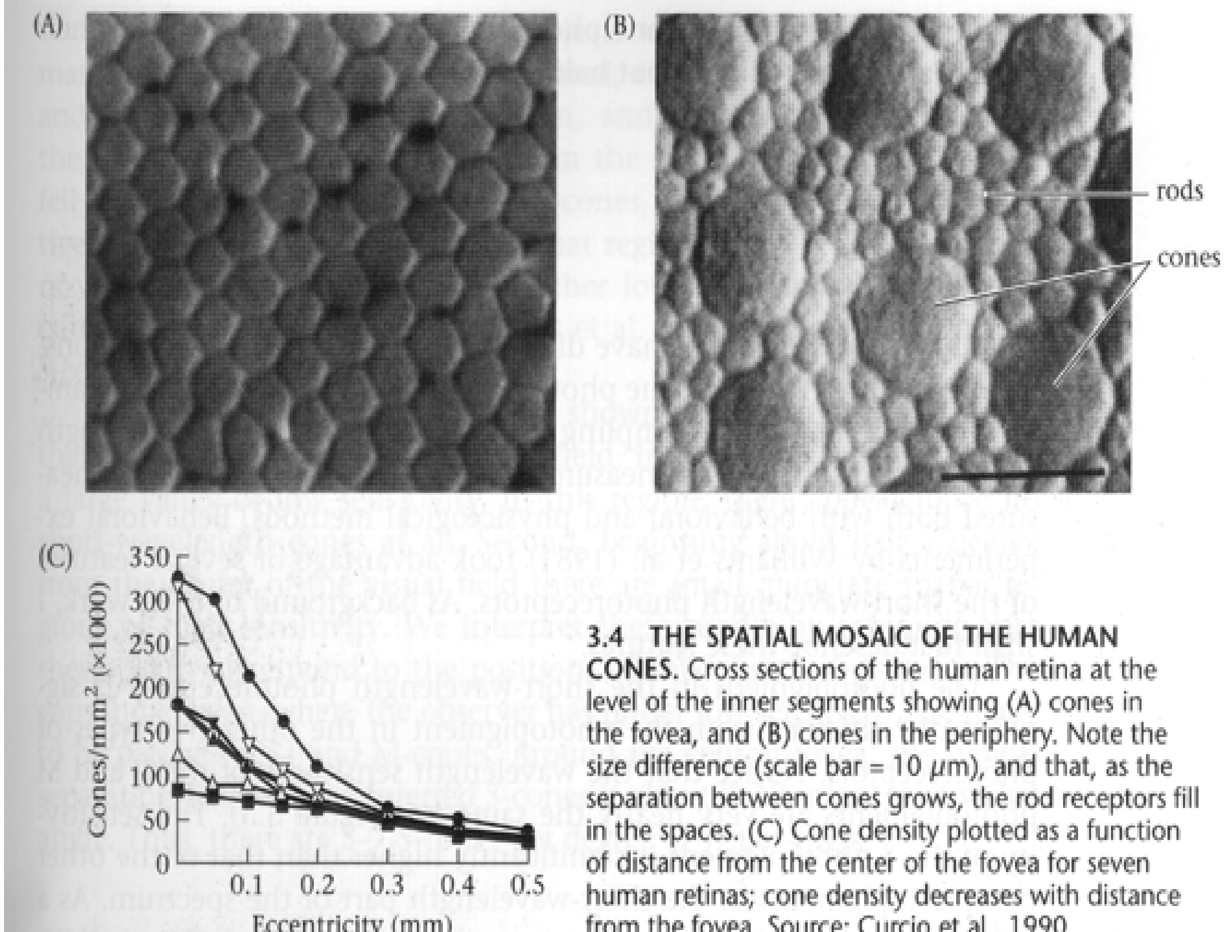
Basic Neuroscience: Anatomy and Physiology Arthur C. Guyton, M.D. 1987 W.B.Saunders Co.

The Eye



- The density of cones is highest at the fovea
  - The region immediately surrounding the fovea is the macula
    - The most important part of your eye: damage == blindness
- Peripheral vision is almost entirely black and white
- Eagles are bifoveate
- Dogs and cats have no fovea, instead they have an elongated slit

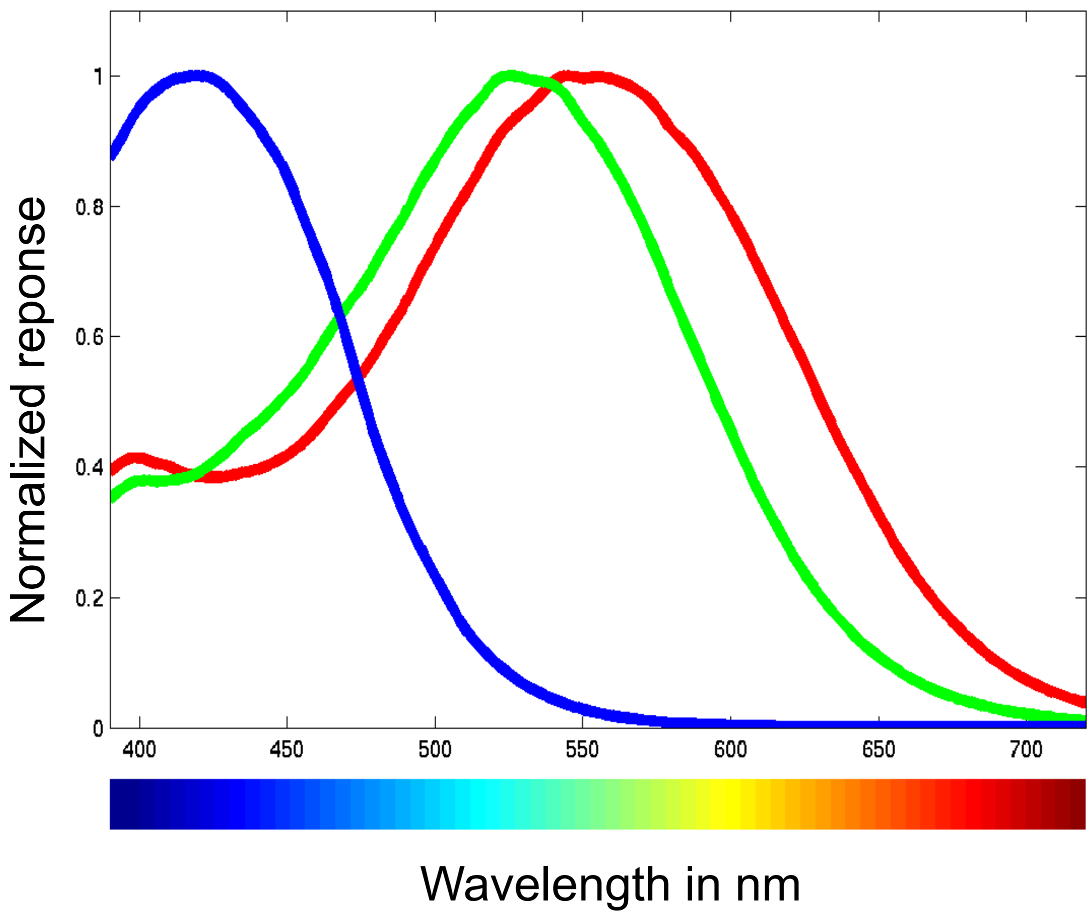
Spatial Arrangement of the Retina



3.4 THE SPATIAL MOSAIC OF THE HUMAN CONES. Cross sections of the human retina at the level of the inner segments showing (A) cones in the fovea, and (B) cones in the periphery. Note the size difference (scale bar = 10 μm), and that, as the separation between cones grows, the rod receptors fill in the spaces. (C) Cone density plotted as a function of distance from the center of the fovea for seven human retinas; cone density decreases with distance from the fovea. Source: Curcio et al., 1990.

(From Foundations of Vision, by Brian Wandell, Sinauer Assoc.)

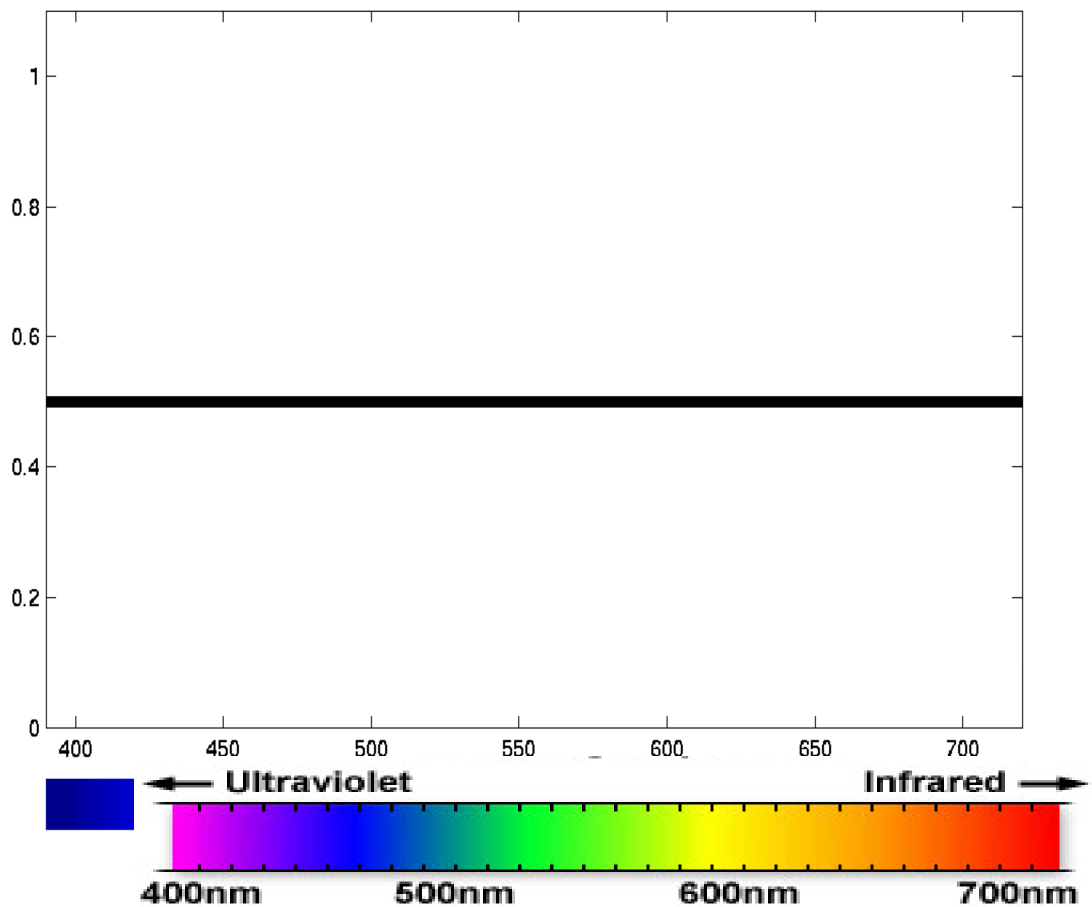
Three Types of Cones (trichromatic vision)



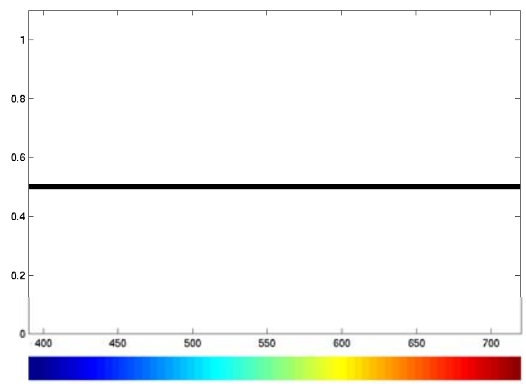
Trichromatic Vision

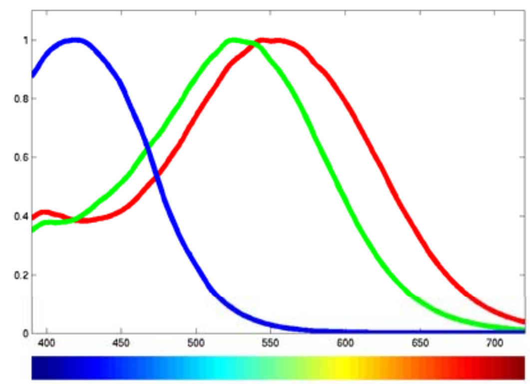
- So-called "blue" light sensors respond to an entire range of frequencies
  - Including in the so-called "green" and "red" regions
- The difference in response of "green" and "red" sensors is small
  - Varies from person to person
    - Each person really sees the world in a different color
  - If the two curves get too close, we have color blindness
    - Ideally traffic lights should be red and blue

White Light


1
0.8
0.6
0.4
0.2
0
400
450
500
550
600
650
700
Ultraviolet
Infrared
400nm
500nm
600nm
700nm

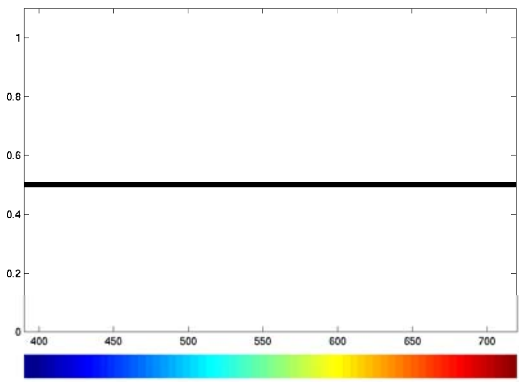
Response to White Light

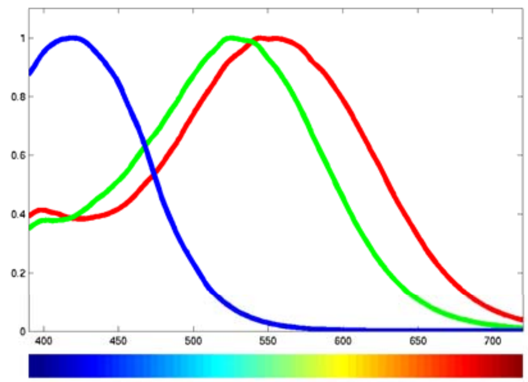


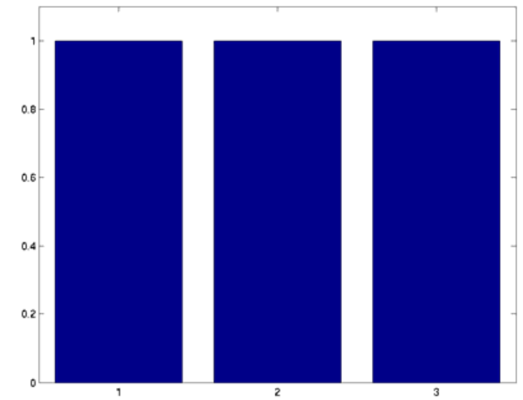


?

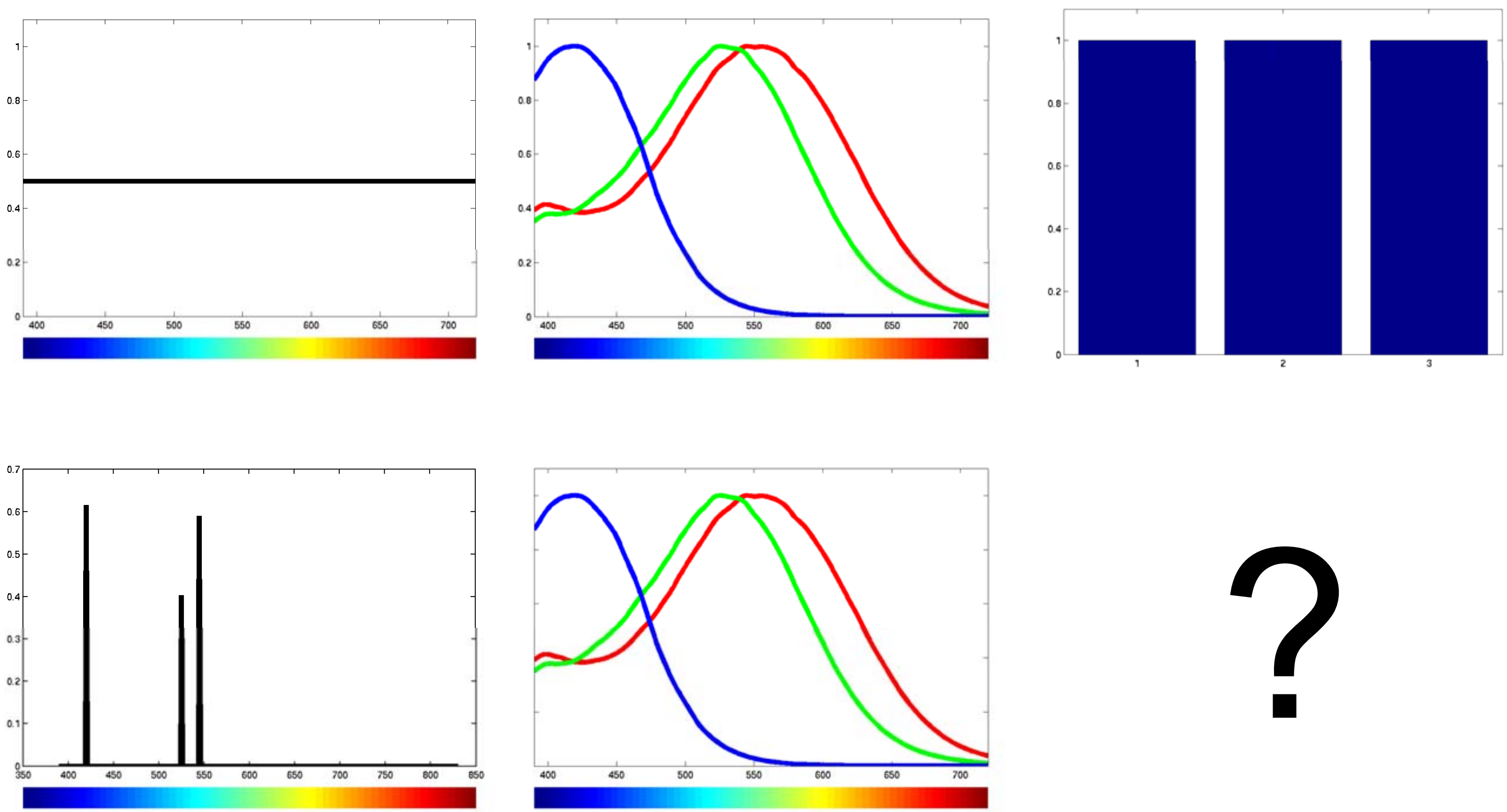
Response to White Light



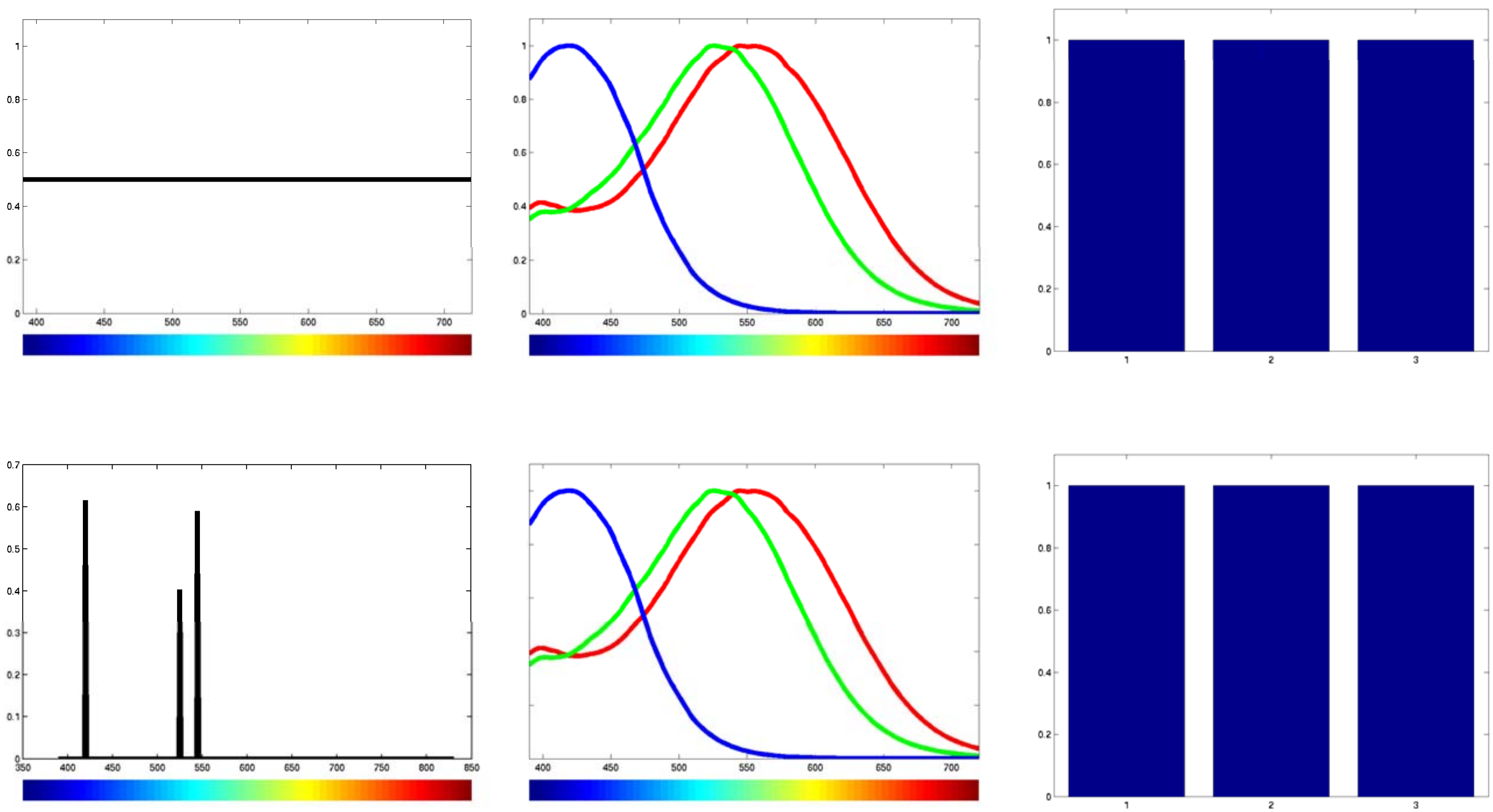




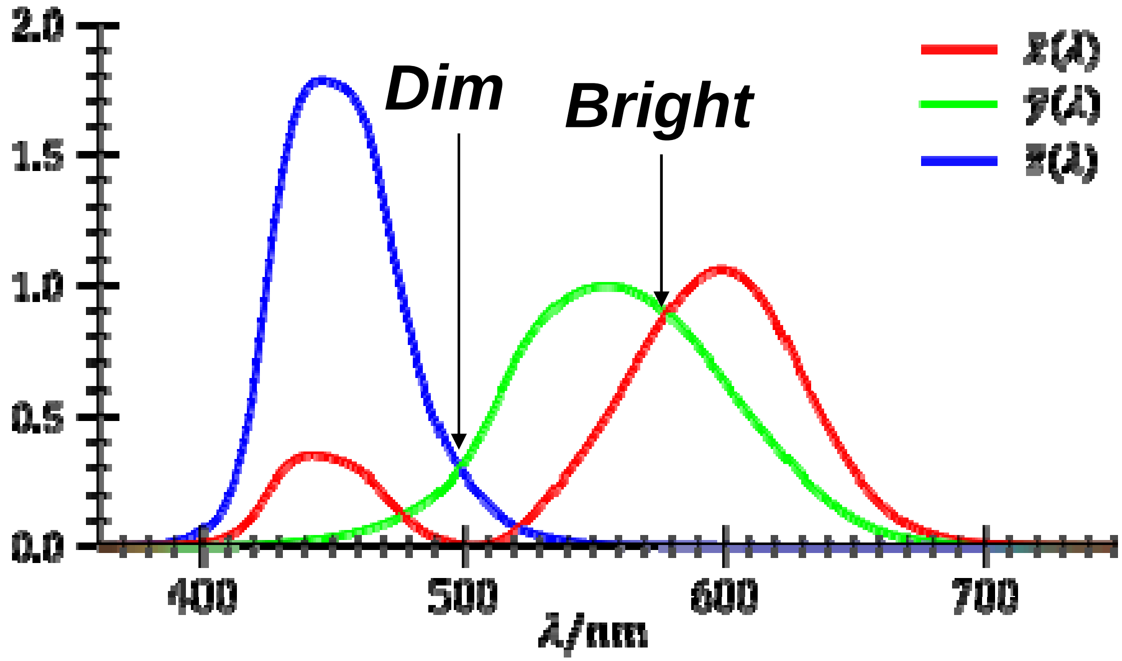
Response to Sparse Light



Response to Sparse Light

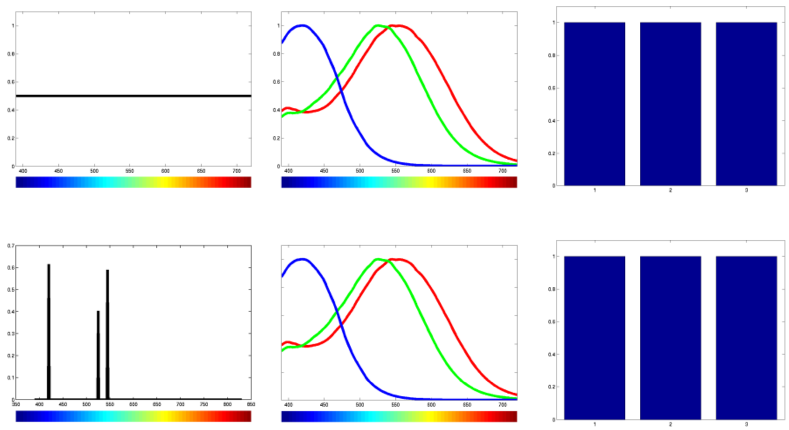


Human perception anomalies



- The same intensity of monochromatic light will result in different *perceived* brightness at different wavelengths
- Many combinations of wavelengths can produce the same sensation of colour.
- Yet humans can distinguish 10 *million* colours

Representing Images

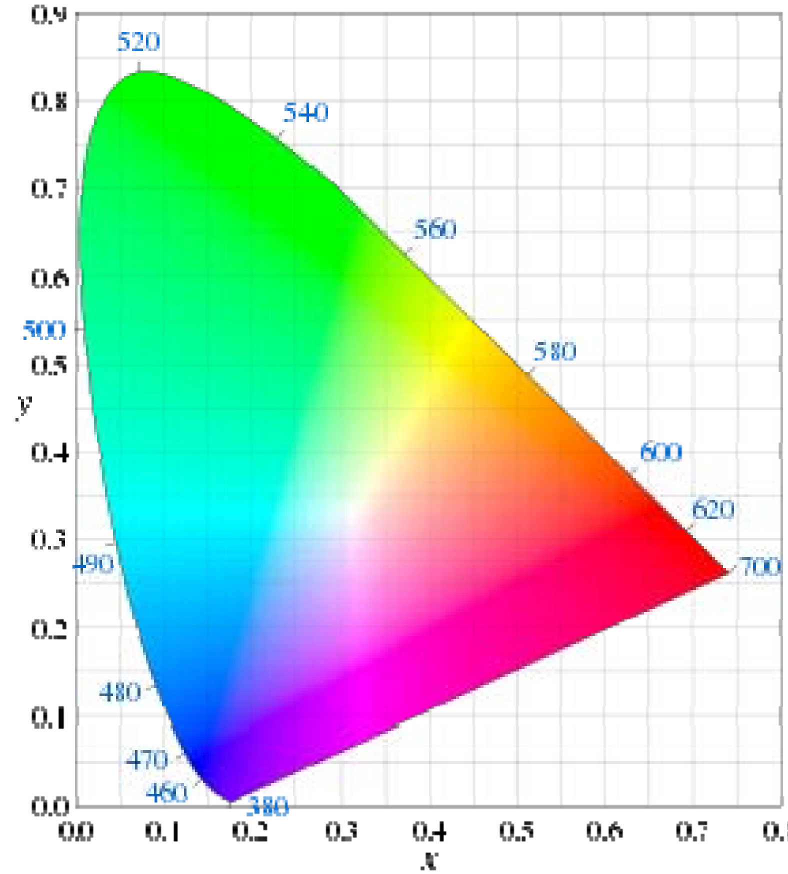


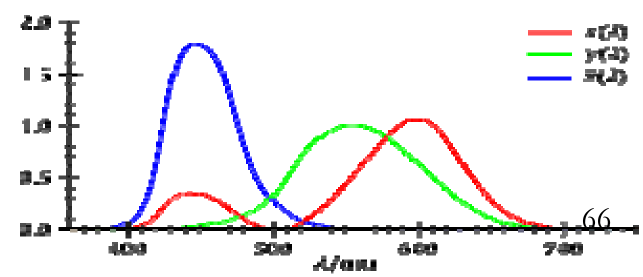
- Utilize trichromatic nature of human vision
  - Sufficient to trigger each of the three cone types in a manner that produces the sensation of the desired color
    - A *tetrachromatic* animal would be very confused by our computer images
      - Some new-world monkeys are tetrachromatic
- The three “chosen” colors are red (650nm), green (510nm) and blue (475nm)
  - By appropriate combinations of these colors, the cones can be excited to produce a very large set of colours
    - Which is still a small fraction of what we can actually see
  - How many colours? …

The “CIE” colour space

- From experiments done in the 1920s by W. David Wright and John Guild
  - Subjects adjusted x,y,and z on the right of a circular screen to match a colour on the left
- X, Y and Z are normalized responses of the three sensors
  - X + Y + Z is 1.0
    - Normalized to have to total net intensity
- The image represents all colours a person can see
  - The outer curved locus represents monochromatic light
    - X,Y and Z as a function of λ
  - The lower line is the line of purples
    - End of visual spectrum
- The CIE chart was updated in 1960 and 1976
  - The newer charts are less popular

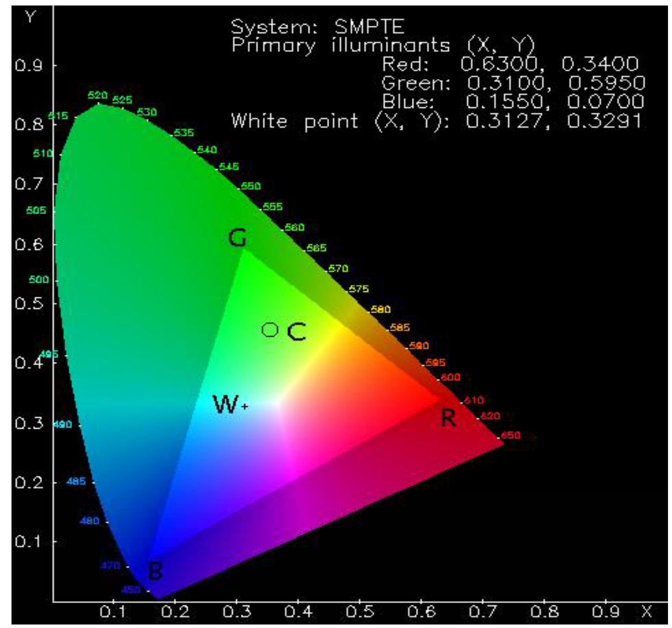
International council on illumination, 1931





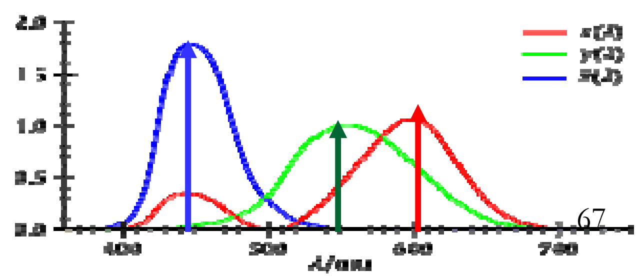
What is displayed

- The RGB triangle
  - Colours outside this area cannot be matched by combining only 3 colours
    - Any other set of monochromatic colours would have a differently restricted area
    - TV images can never be like the real world



- Each corner represents the (X,Y,Z) coordinate of one of the three “primary” colours used in images

- In reality, this represents a very tiny fraction of our visual acuity
  - Also affected by the quantization of levels of the colours



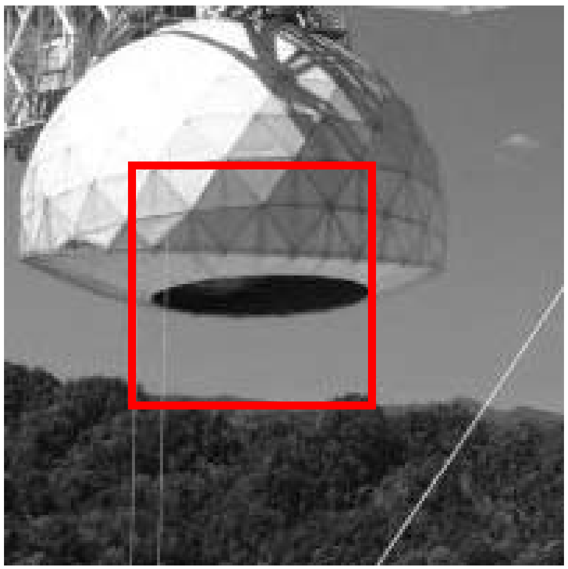
Representing Images on Computers

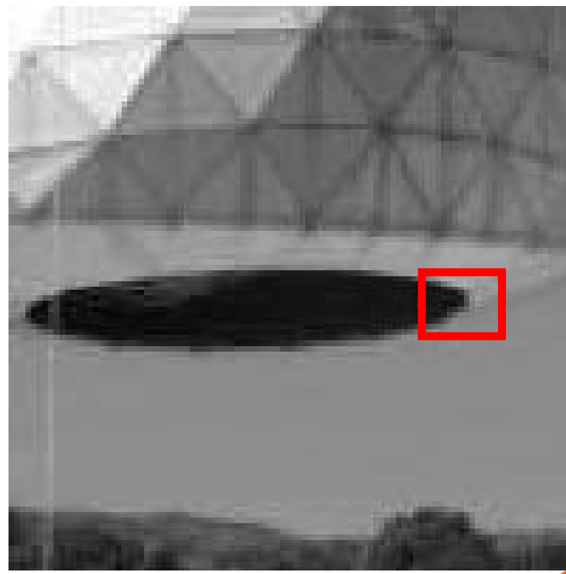
- Greyscale: a single matrix of numbers
  - Each number represents the intensity of the image at a specific location in the image
  - Implicitly, R = G = B at all locations

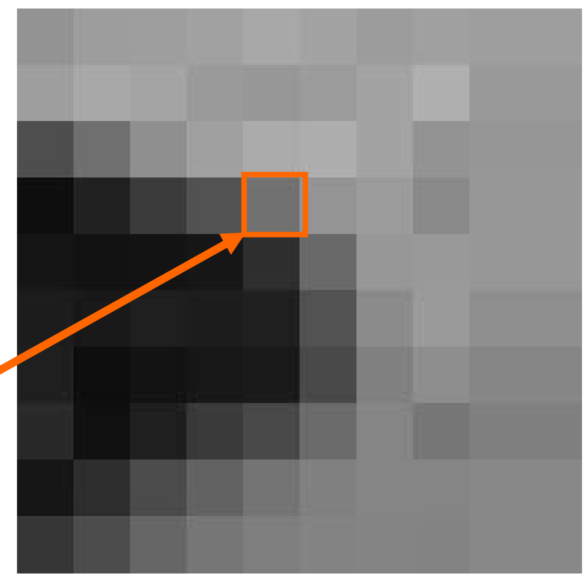
- Color: 3 matrices of numbers
  - The matrices represent different things in different representations
  - RGB Colorspace: Matrices represent intensity of Red, Green and Blue
  - CMYK Colorspace: Cyan, Magenta, Yellow
  - YIQ Colorspace..
  - HSV Colorspace..

Computer Images: Grey Scale

R = G = B. Only a single number need be stored per pixel

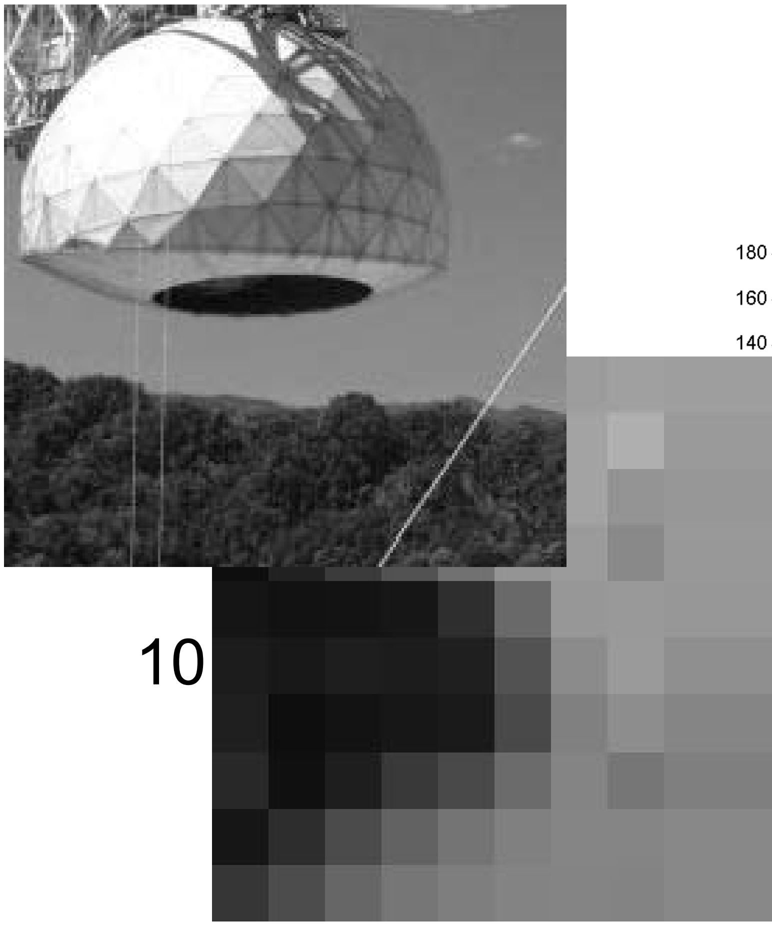






Picture Element (PIXEL)
Position & gray value (scalar)

What we see



10

10

What the computer “sees”

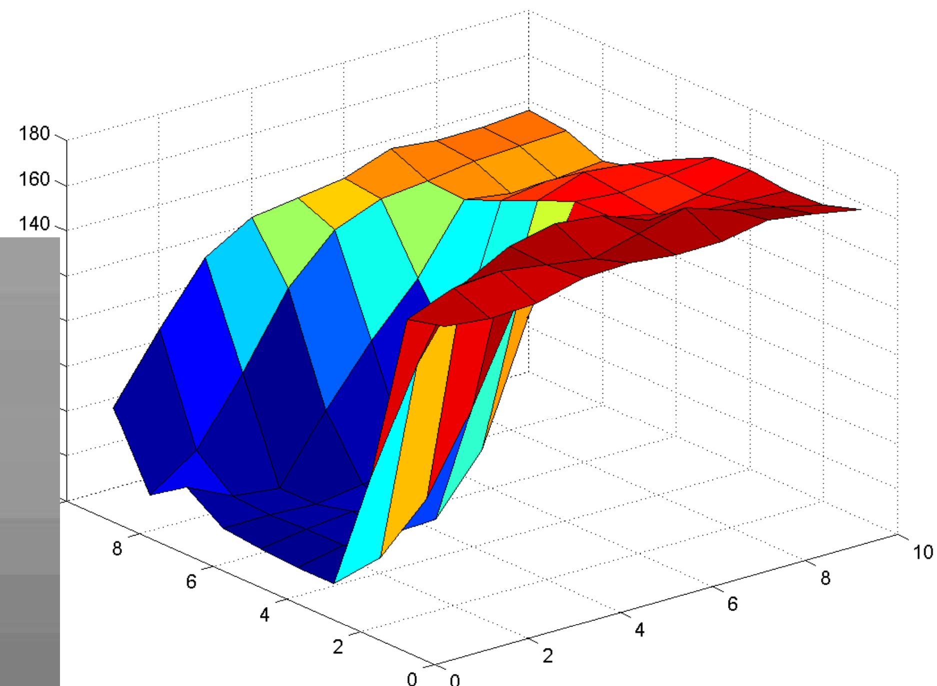
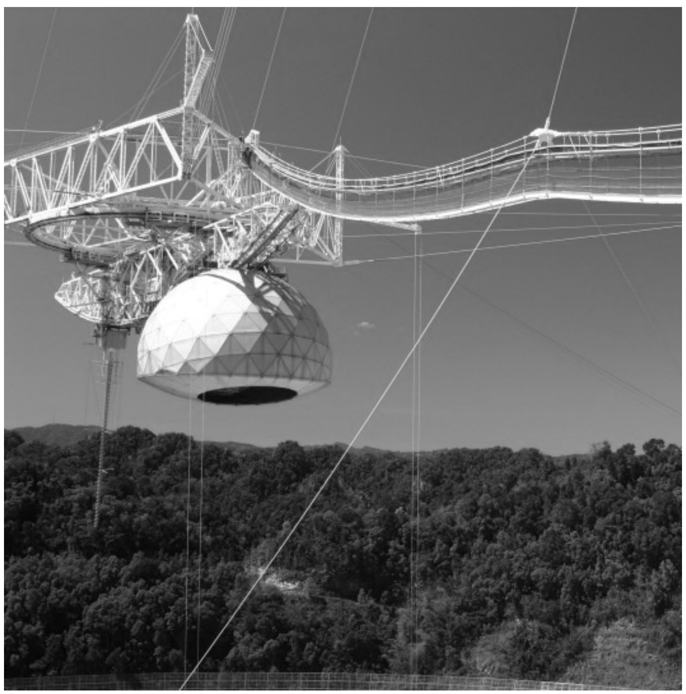
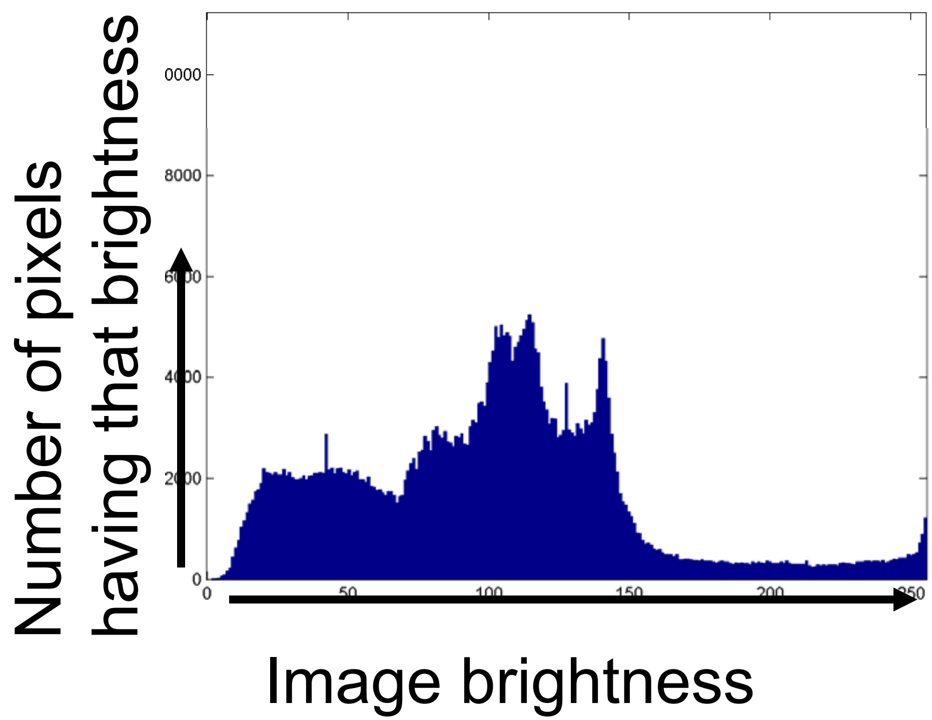


Image Histograms





Example histograms

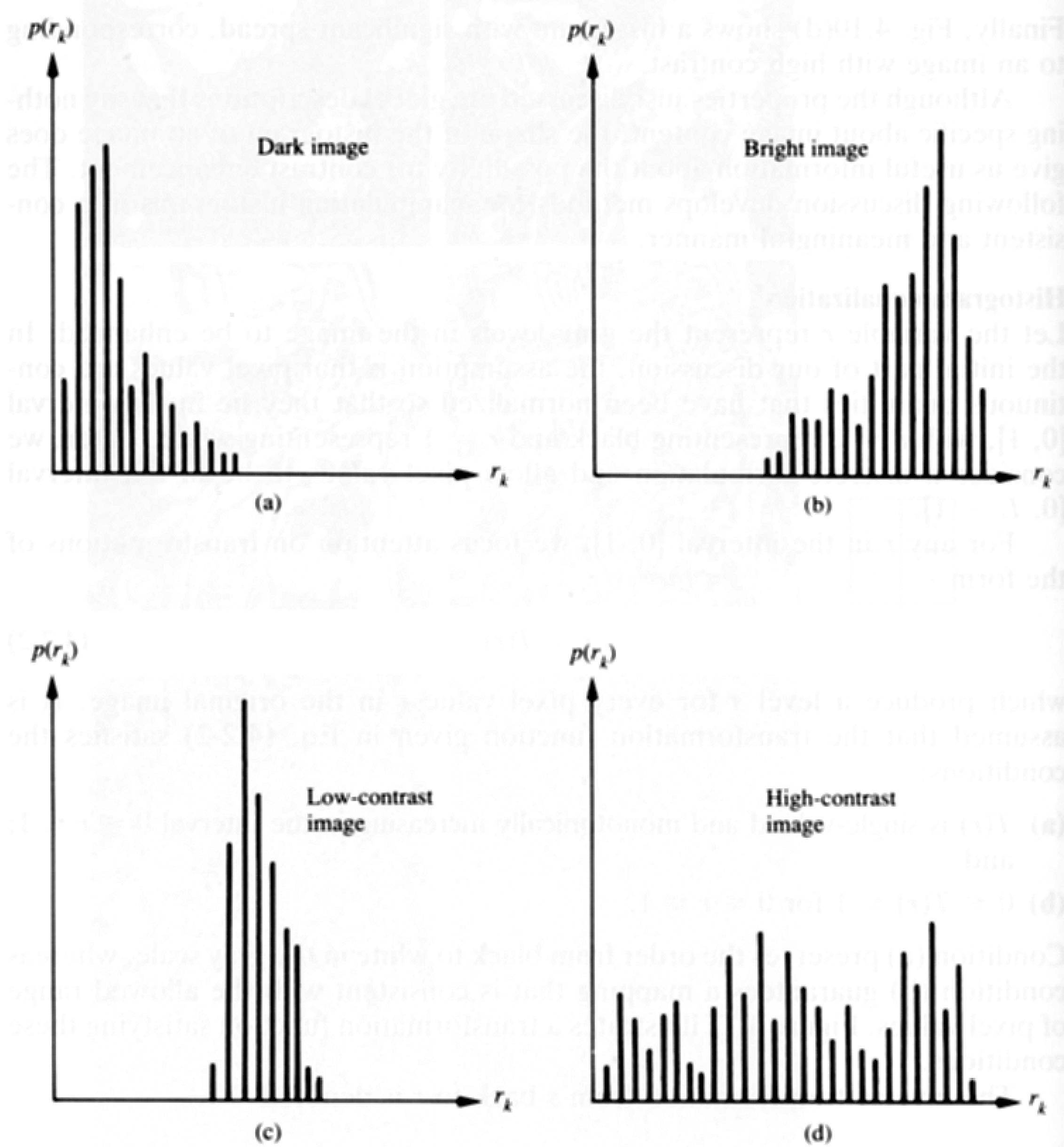


Figure 4.10 *Histograms corresponding to four basic image types.*

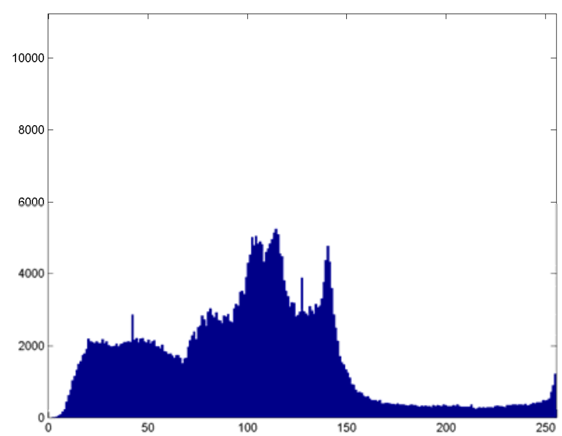
From: Digital Image Processing,

by Gonzales and Woods, Addison Wesley, 1992

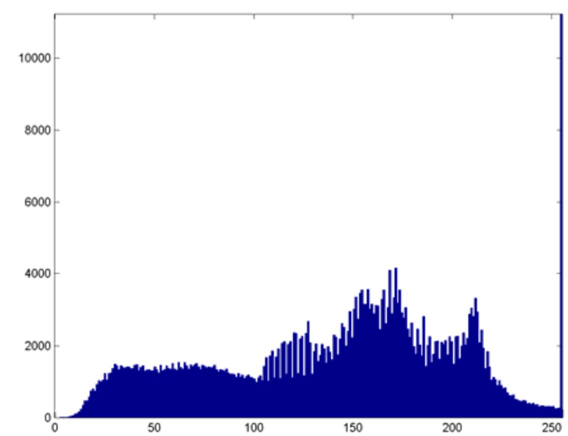
Pixel operations

- New value is a function of the old value
  - Tonescale to change image brightness
  - Threshold to reduce the information in an image
  - Colorspace operations

J=1.5*I

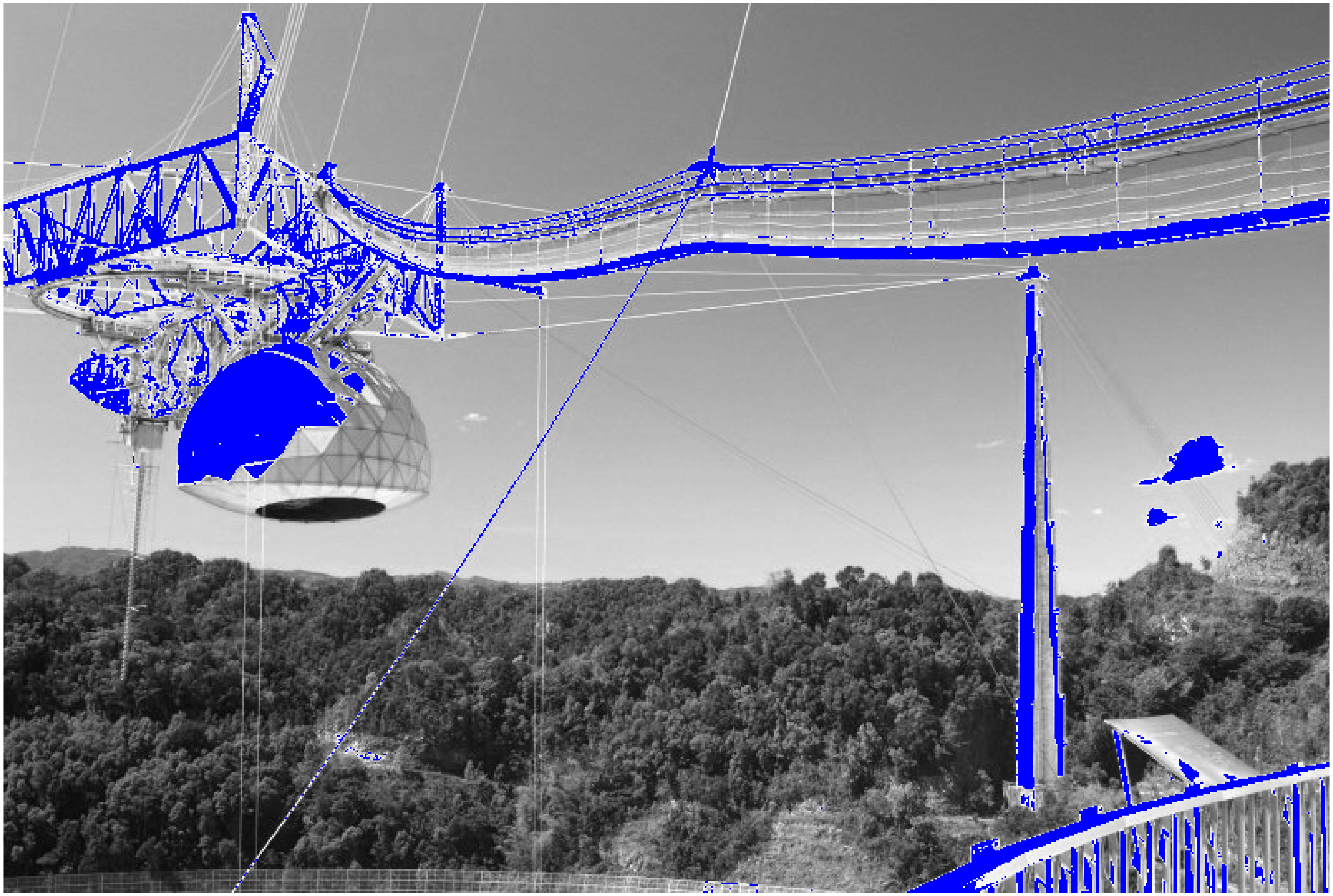


0
2000
4000
6000
8000
10000
0
50
100
150
200
250

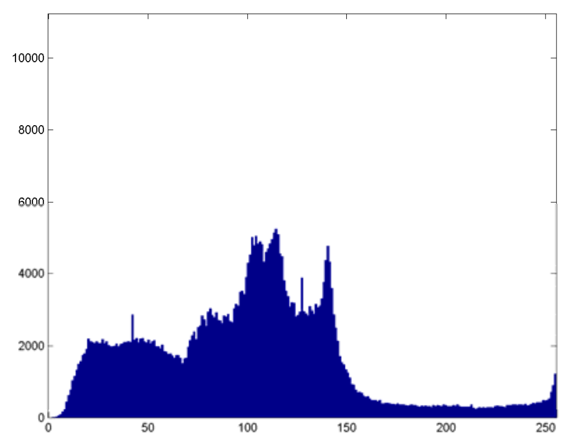


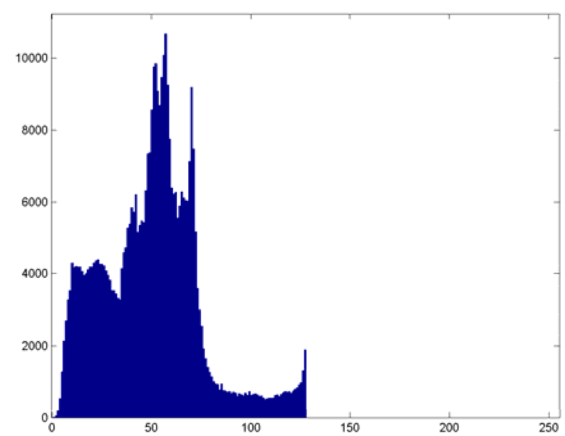
0
2000
4000
6000
8000
10000
0
50
100
150
200
250

Saturation

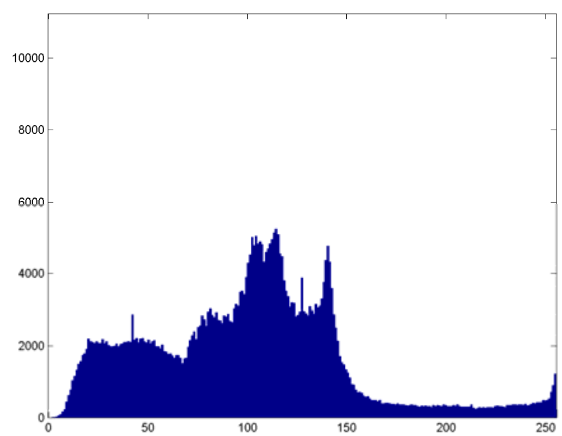


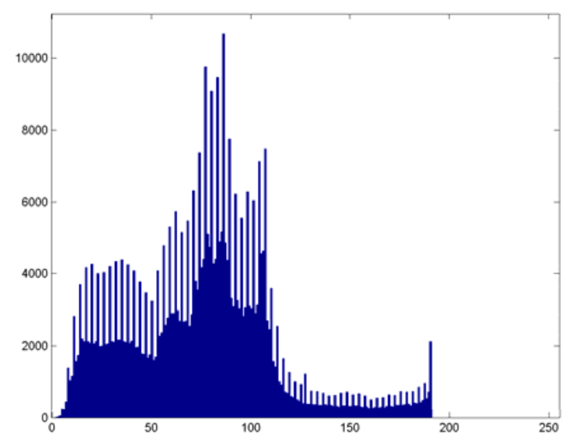
J=0.5*I


10000
8000
6000
4000
2000
0
0
50
100
150
200
250

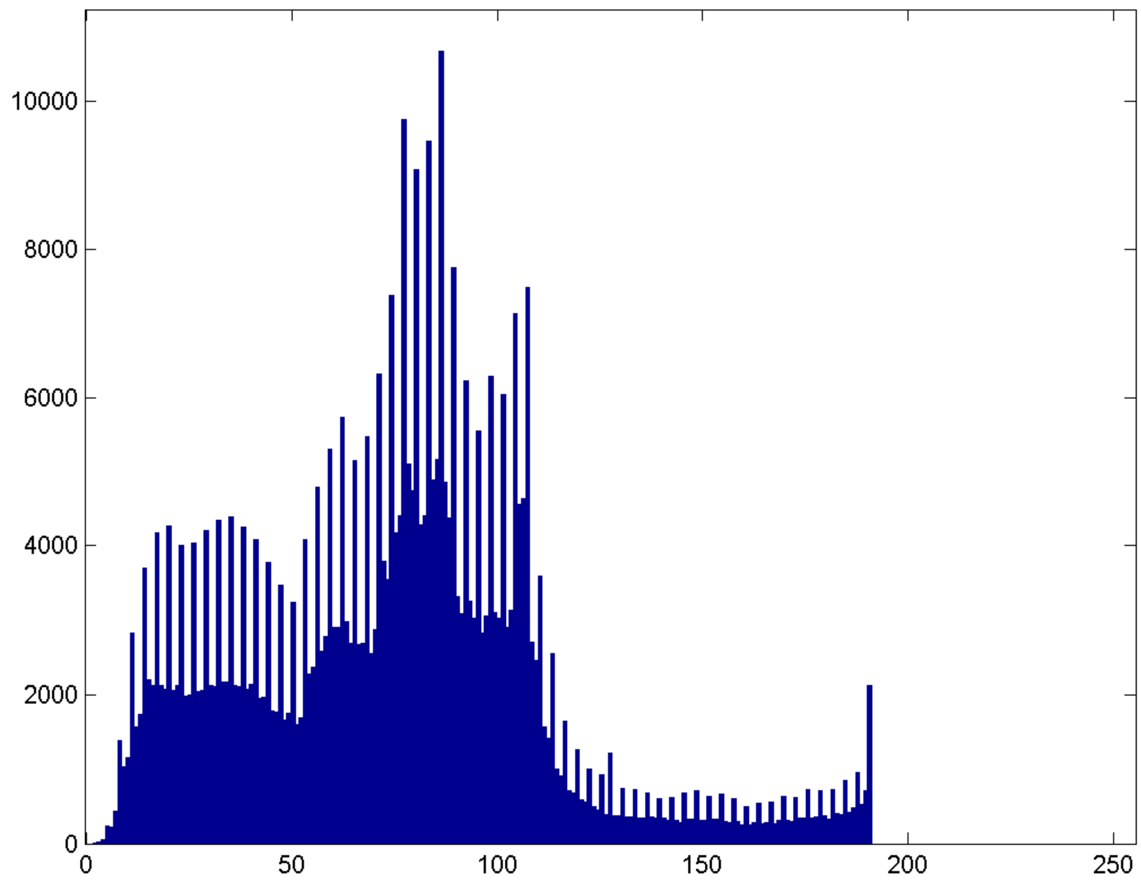

10000
8000
6000
4000
2000
0
0
50
100
150
200
250

J=uint8(0.75*I)

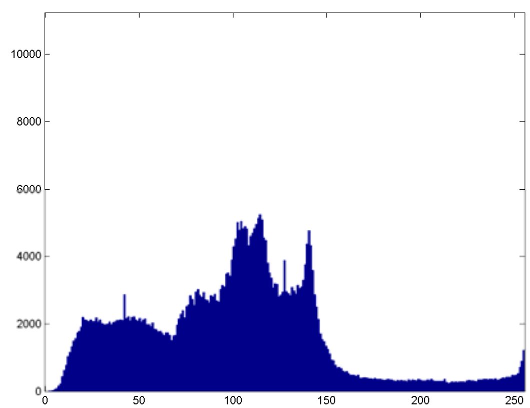

10000
8000
6000
4000
2000
0
0
50
100
150
200
250

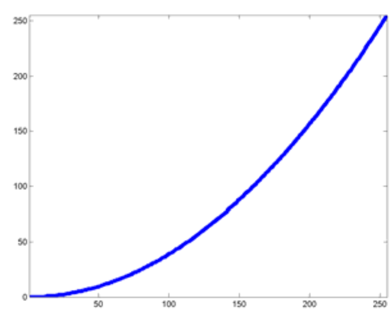

10000
8000
6000
4000
2000
0
0
50
100
150
200
250

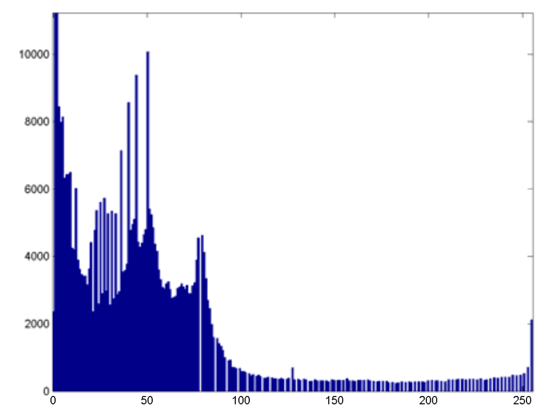
What's this?



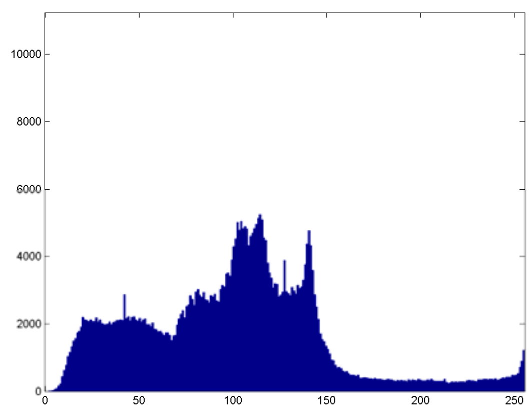
Non-Linear Darken

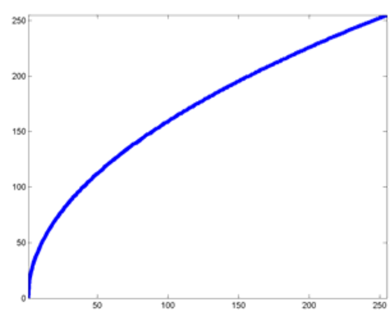


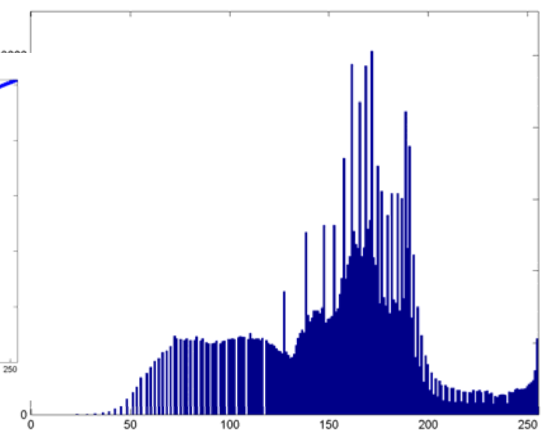




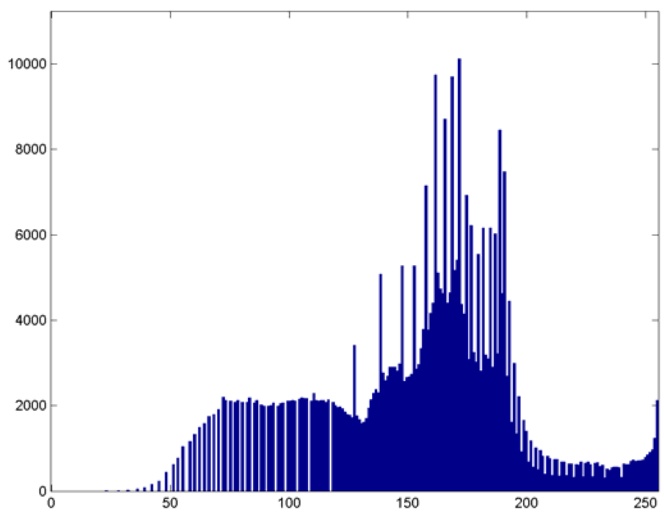
Non-Linear Lighten

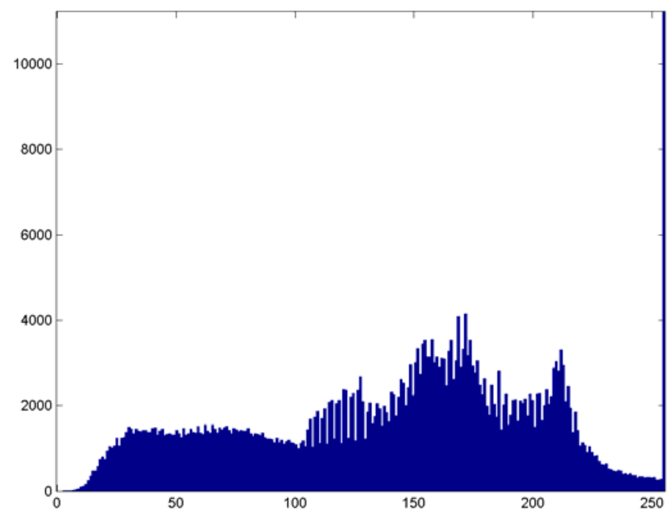




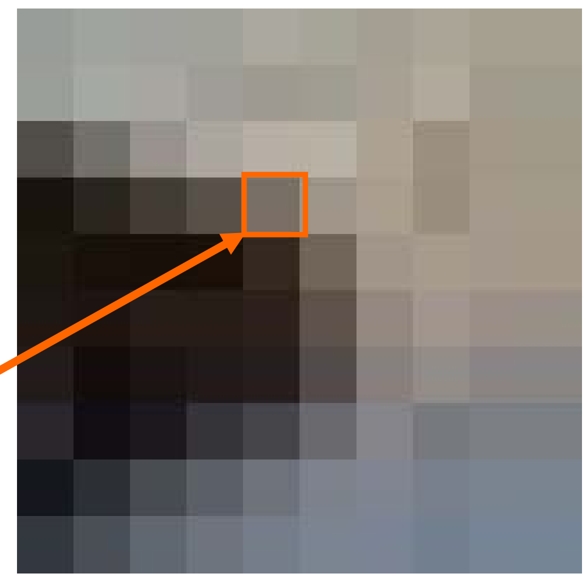


Linear vs. Non-Linear





Color Images



Picture Element (PIXEL)
Position & color value (red, green, blue)

RGB Representation

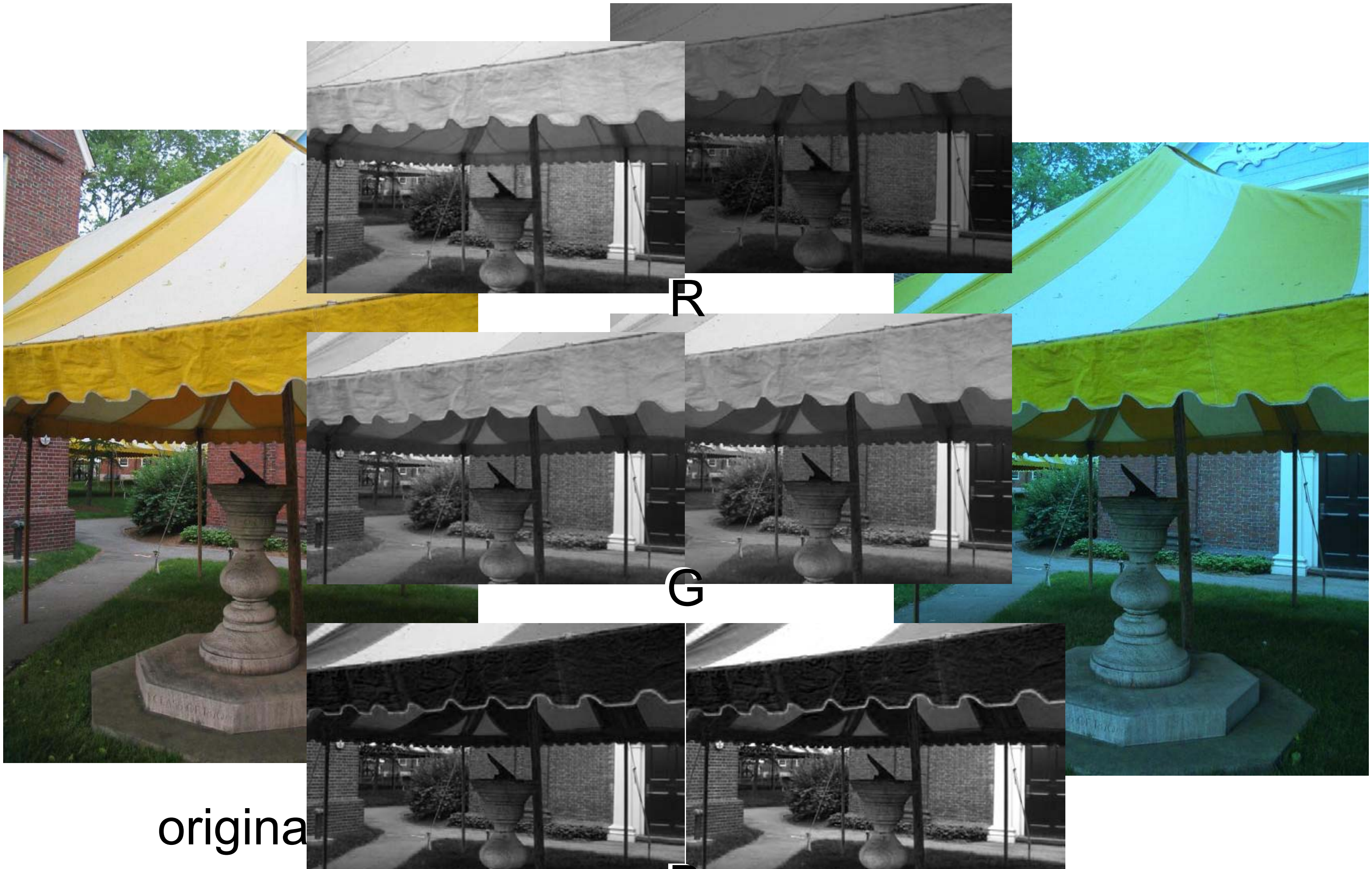
original

R

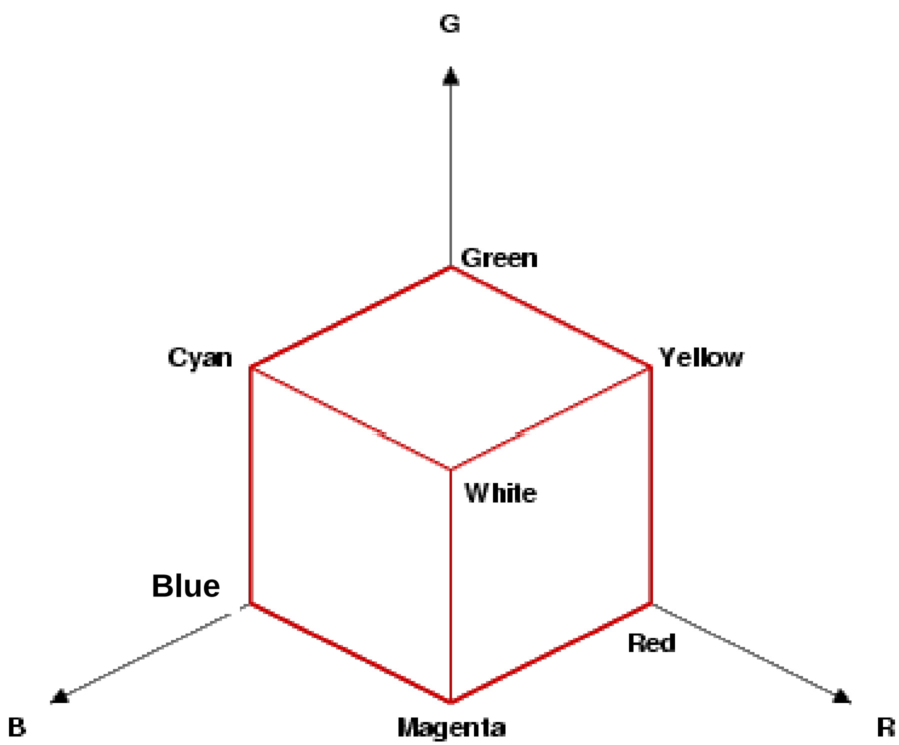
G

B

RGB Manipulation Example: Color Balance

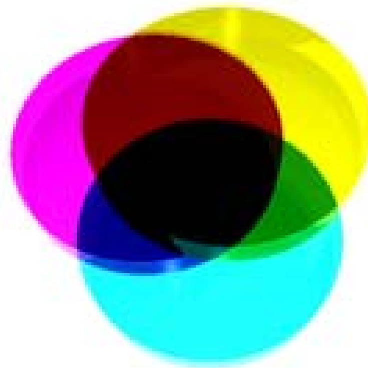


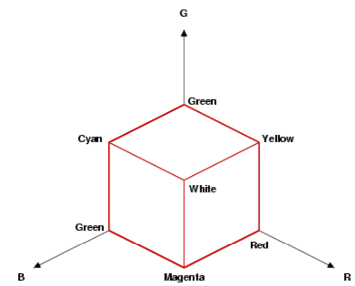
The CMYK color space



- Represent colors in terms of cyan, yellow and magenta
  - The “K” stands for “Key”, not “black”

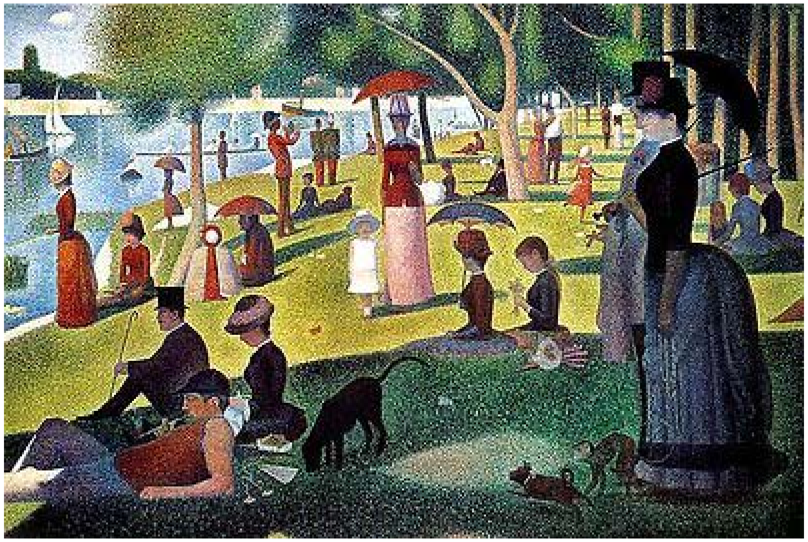
CMYK is a *subtractive* representation





- RGB is based on *composition*, i.e. it is an additive representation
  - Adding equal parts of red, green and blue creates white
- What happens when you mix red, green and blue paint?
  - Clue – paint colouring is subtractive..
- CMYK is based on *masking,* i.e. it is subtractive
  - The base is white
  - Masking it with equal parts of C, M and Y creates Black
  - Masking it with C and Y creates Green
    - Yellow masks blue
  - Masking it with M and Y creates Red
    - Magenta masks green
  - Masking it with M and C creates Blue
    - Cyan masks green
  - Designed specifically for *printing*
    - As opposed to rendering

An Interesting Aside



- Paints create subtractive coloring
  - Each paint masks out some colours
  - Mixing paint subtracts combinations of colors
  - Paintings represent subtractive colour masks
- In the 1880s Georges-Pierre Seurat pioneered an *additive-colour* technique for painting based on “pointilism”
  - How do you think he did it?

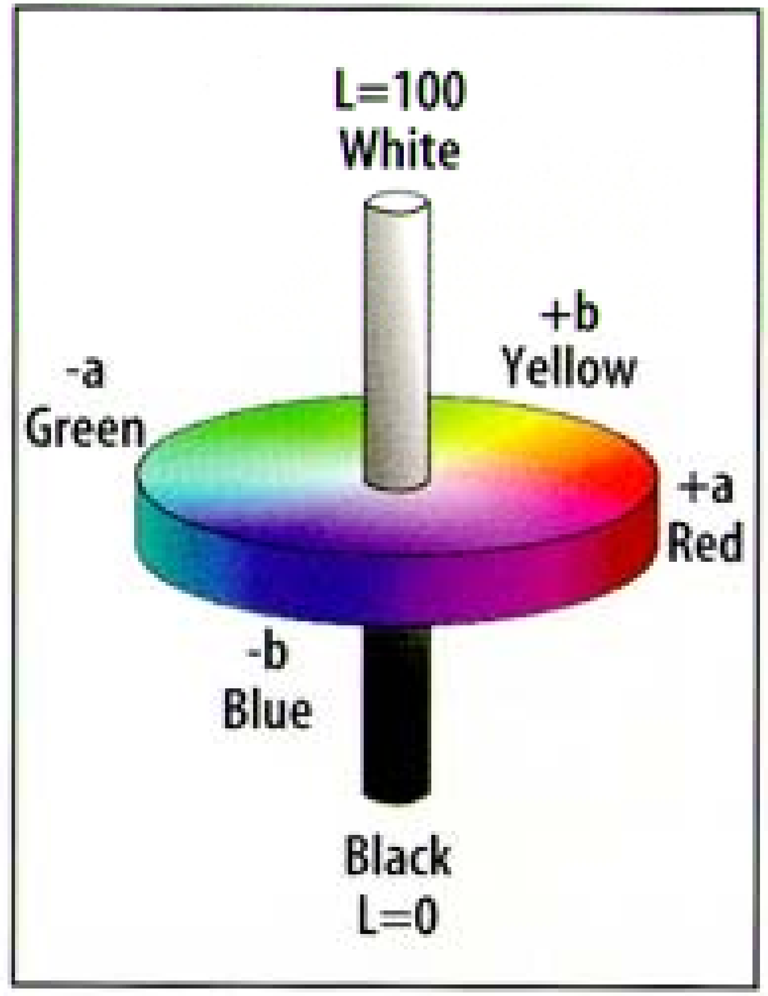
NTSC color components

Y = “luminance”

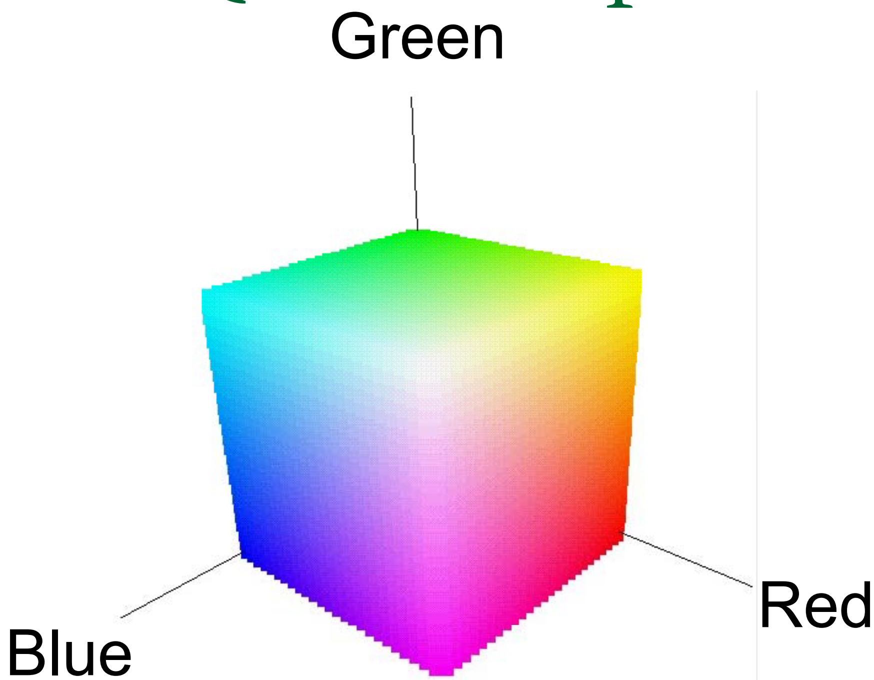
I = “red-green”

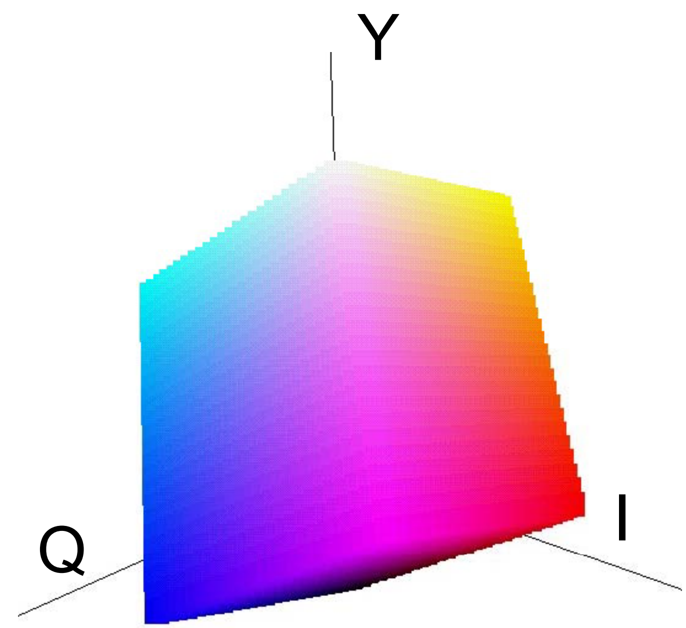
Q = “blue-yellow”

a.k.a. YUV although YUV is actually the color specification for PAL video



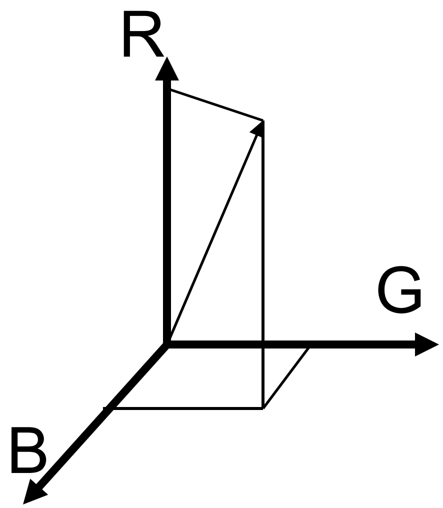
YIQ Color Space

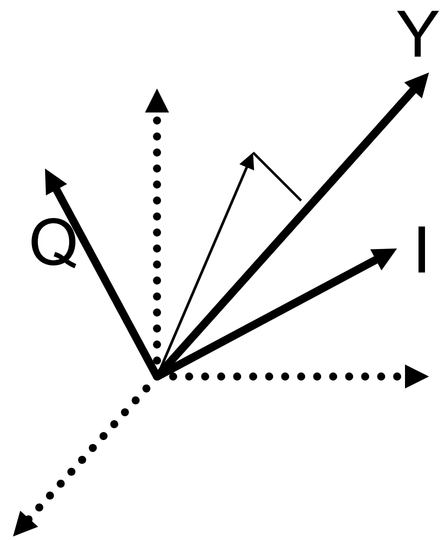




$$\begin{bmatrix} Y \\ I \\ Q \end{bmatrix} = \begin{bmatrix} .299 & .587 & .114 \\ .596 & -.275 & -.321 \\ .212 & -.523 & .311 \end{bmatrix} \begin{bmatrix} R \\ G \\ B \end{bmatrix}$$

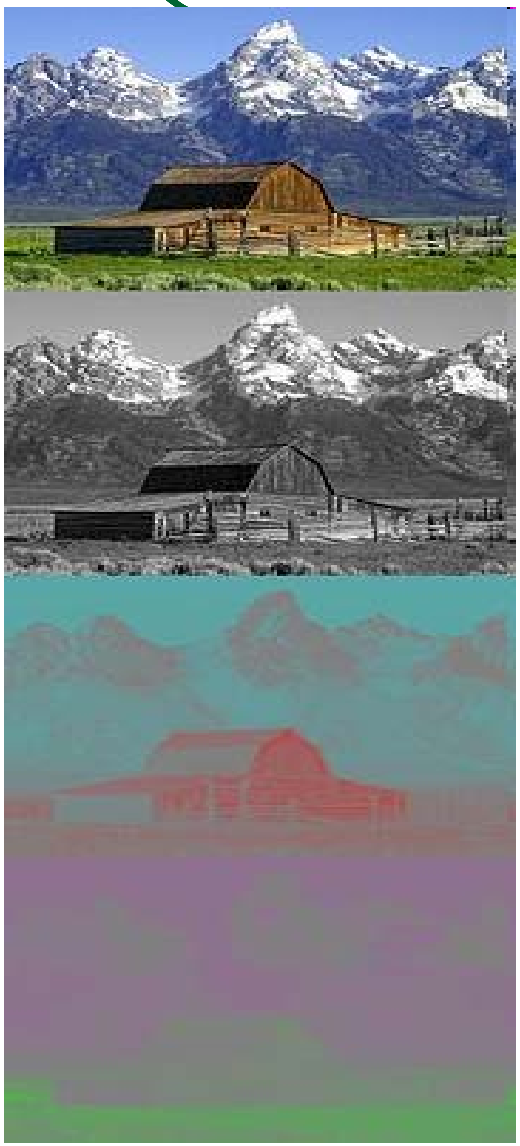
Color Representations





- Y value lies in the same range as R,G,B ([0,1])
- I is to [-0.59 0.59]
- Q is limited to [-0.52 0.52]
- Takes advantage of lower human sensitivity to I and Q axes

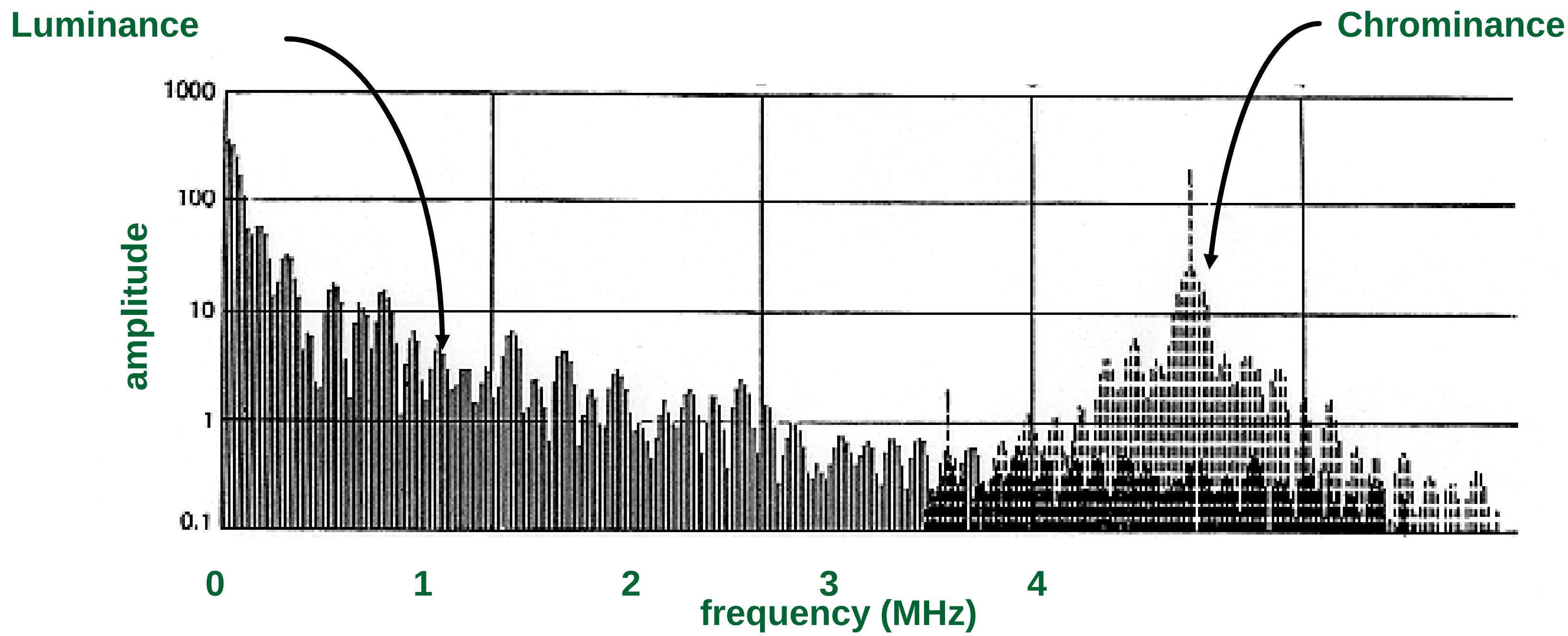
YIQ



- Top: Original image
- Second: Y
- Third: I (displayed as red-cyan)
- Fourth: Q (displayed as green-magenta)
  - From http://wikipedia.org/

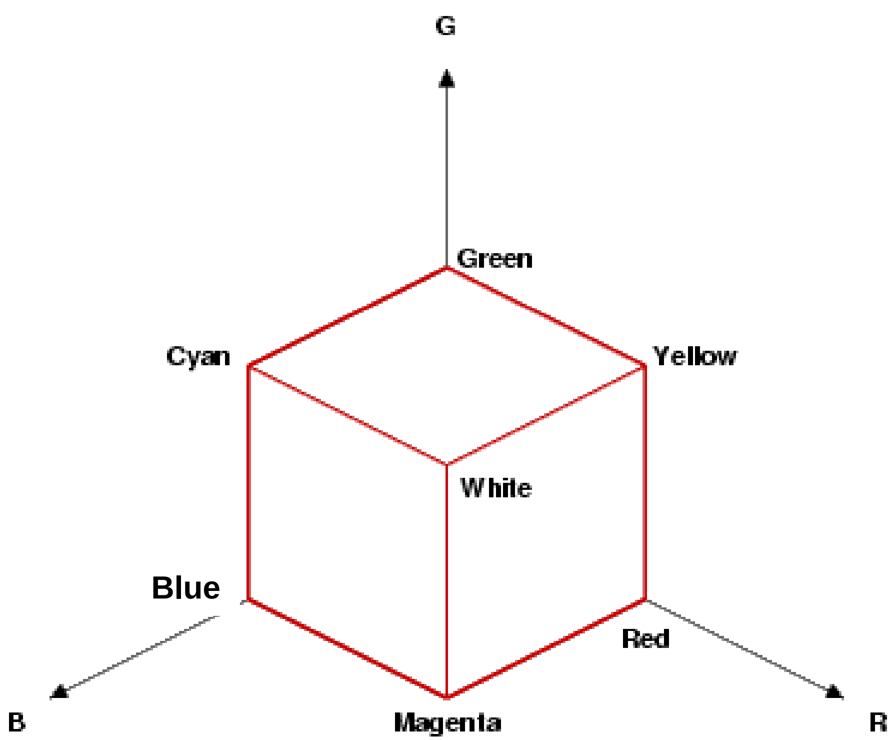
- Processing (e.g. histogram equalization) only needed on Y
  - In RGB must be done on all three colors. Can distort image colors
  - A black and white TV only needs Y

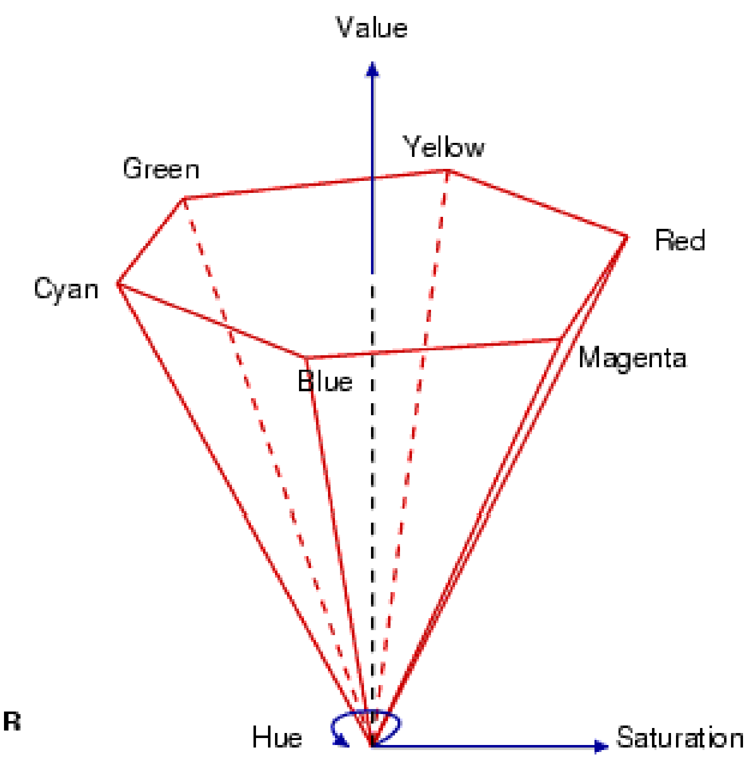
Bandwidth (transmission resources) for the components of the television signal



Understanding image perception allowed NTSC to add color to the black and white television signal. The eye is more sensitive to I than Q, so lesser bandwidth is needed for Q. Both together used much less than Y, allowing for color to be added for minimal increase in transmission bandwidth.

Hue, Saturation, Value



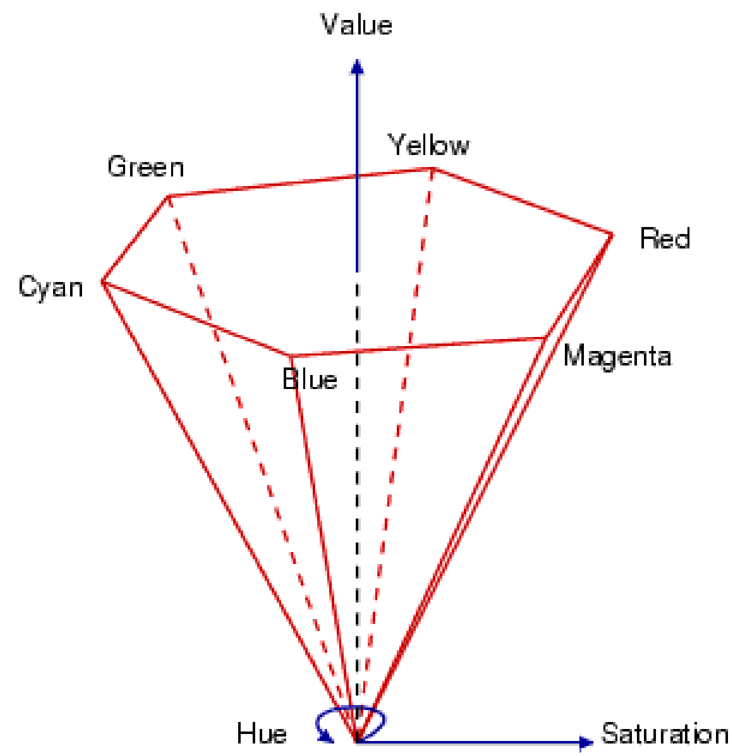


The HSV Colour Model By Mark Roberts
http://www.cs.bham.ac.uk/~mer/colour/hsv.html

V = [0,1], S = [0,1]
H = [0,360]

HSV

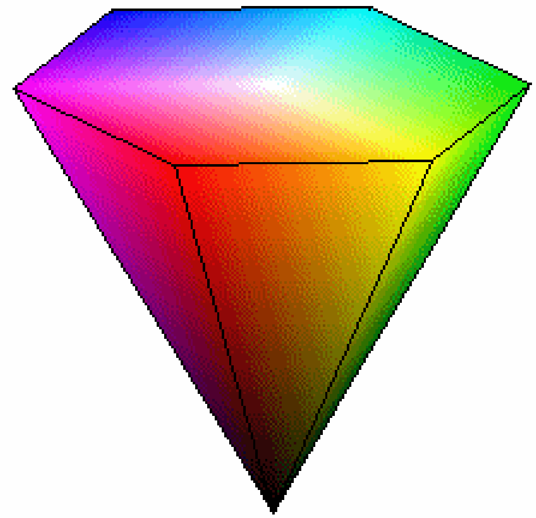
- V = Intensity
  - 0 = Black
  - 1 = Max (white at S = 0)
- S = 1:
  - As H goes from 0 (Red) to 360, it represents a different combinations of 2 colors
- As S->0, the color components from the opposite side of the polygon increase



V = [0,1], S = [0,1]
H = [0,360]

Hue, Saturation, Value

$$h = \begin{cases} 0, & \text{if } \max = \min \\ (60^\circ \times \frac{g-b}{\max - \min} + 360^\circ) \bmod 360^\circ, & \text{if } \max = r \\ 60^\circ \times \frac{b-r}{\max - \min} + 120^\circ, & \text{if } \max = g \\ 60^\circ \times \frac{r-g}{\max - \min} + 240^\circ, & \text{if } \max = b \end{cases}$$



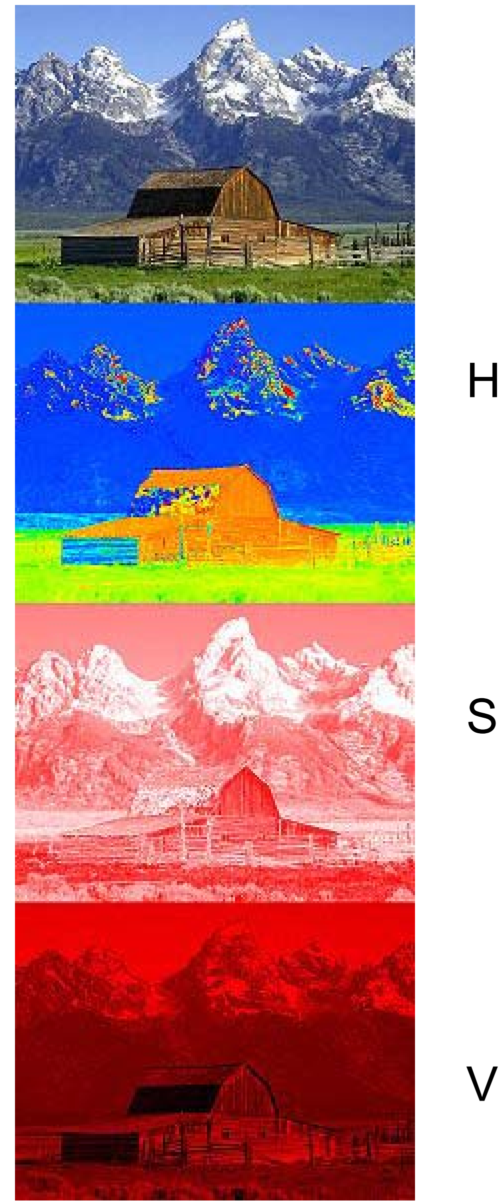
$$s = \begin{cases} 0, & \text{if } \max = 0 \\ \frac{\max - \min}{\max} = 1 - \frac{\min}{\max}, & \text{otherwise} \end{cases}$$

$$v = \max$$

Max is the maximum of (R,G,B)
Min is the minimum of (R,G,B)

HSV

- Top: Original image
- Second H (assuming S = 1, V = 1)
- Third S (H=0, V=1)
- Fourth V (H=0, S=1)



Quantization and Saturation

- Captured images are typically quantized to N-bits
- Standard value: 8 bits
- 8-bits is not very much < 1000:1
- Humans can easily accept 100,000:1
- And most cameras will give you 6-bits anyway…

Processing Colour Images

- Typically work only on the Grey Scale image
  - Decode image from whatever representation to RGB
  - GS = R + G + B
- The Y of YIQ may also be used
  - Y is a linear combination of R,G and B
- For specific algorithms that deal with colour, individual colours may be maintained
  - Or any linear combination that makes sense may be maintained.

Reference Info

- Many books
  - Digital Image Processing, by Gonzales and Woods, Addison Wesley, 1992
  - Computer Vision: A Modern Approach, by David A. Forsyth and Jean Ponce
  - Spoken Language Processing: A Guide to Theory, Algorithm and System Development, by Xuedong Huang, Alex Acero and Hsiao-Wuen Hon