

# Majid. In his office.

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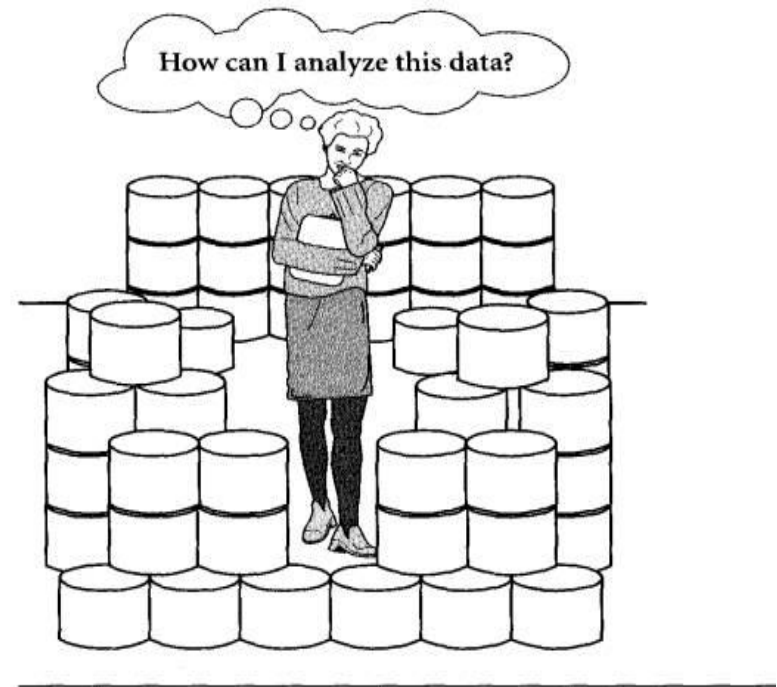
Drop in or email to  
arrange an appointment.



# SPS: The Story So Far

The sorts of ways we wish to manipulate and analyze data:

- Data Properties
- Data Modelling
- Classification and recognition
- Clustering and segmentation
- Estimation and detection



# SPS: *The Next Frontier!*


- Data Representations
- Data Transformations
- Feature extraction

- ❖ Fourier Space Analysis
- ❖ Convolutions
- ❖ Principal Component Analysis
- ❖ Coordinate Transformations

## This Lecture:

- Overview
- Intro to Signals

Maths:  
*nothing  
scary!*

$$\begin{aligned}
 c &= a + b + d \\
 c &= (\pi \cdot 5 \cdot (\Omega - 10^\circ) + 3\alpha + 2 \cdot 3 \ln 11)^{\frac{1}{2}} \\
 c &= (\pi \cdot 5 \cdot \log \frac{1}{x+y} + 3\alpha + 6 \ln 11)^{\frac{1}{2}} \\
 c &= \left[ \int_{x_1}^{x_2} \sum_{i=1}^n \alpha dx + \frac{3[(3+7x)^2 + 6 + 3\pi]}{(5+y)(8+z)+1} + 6 \ln 11 \right]^{\frac{1}{2}} \\
 c &= \left[ \int_{x_1}^{x_2} \sum_{i=1}^n \frac{(3+7x)^2 + 6 + 3\pi}{(5+y)(8+z)+1} dx + \frac{3[(3+7x)^2 + 6 + 3\pi]}{(5+y)(8+z)+1} + 6 \ln 11 \right]^{\frac{1}{2}} \\
 c &= \left[ \int_{x_1}^{x_2} \sum_{i=1}^n \frac{(3+7x)^2 + (\beta - 180^\circ) + 3\pi}{(5+y)(8+z)+1} dx + \frac{3[(3+7x)^2 + (\beta - 180^\circ) + 3\pi]}{(5+y)(8+z)+1} + 6 \ln 11 \right]^{\frac{1}{2}} \\
 c &= \left[ \int_{x_1}^{x_2} \sum_{i=1}^n \frac{\sqrt{3+7x} + (\beta - 180^\circ) + 3\pi}{\frac{(5+y)(8+z) + \log 8}{10\Omega - 6\pi - 1}} dx + \frac{3[\sqrt{3+7x} + (\beta - 180^\circ) + 3\pi]}{\frac{(5+y)(8+z)}{10\Omega - 6\pi - 1} + \log 8} + 6 \ln 11 \right]^{\frac{1}{2}} \\
 c &= \sqrt{\left[ \int_{x_1}^{x_2} \sum_{i=1}^n \alpha dx + \frac{3[\sqrt{3+7x} + (\beta - 180^\circ) + 3\pi]}{\frac{(5+y)(8+z)}{10\Omega - 6\pi - 1} + \log 8} + 6 \ln 11 \right]} \\
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 \end{aligned}$$


# Representing Data

To manipulate data properly we may have to represent it in a different way. *Why?*

- Sometimes we need to look at data in a different way.
- Sometimes we need to alter it to prepare it for the next stage of processing or data analysis. Because:
  - It is noisy (errors or outliers),
  - It is missing values,
  - It contains redundancies,
  - It contains inconsistencies
  - It reveals its substance or begins to make sense

# Representing Data

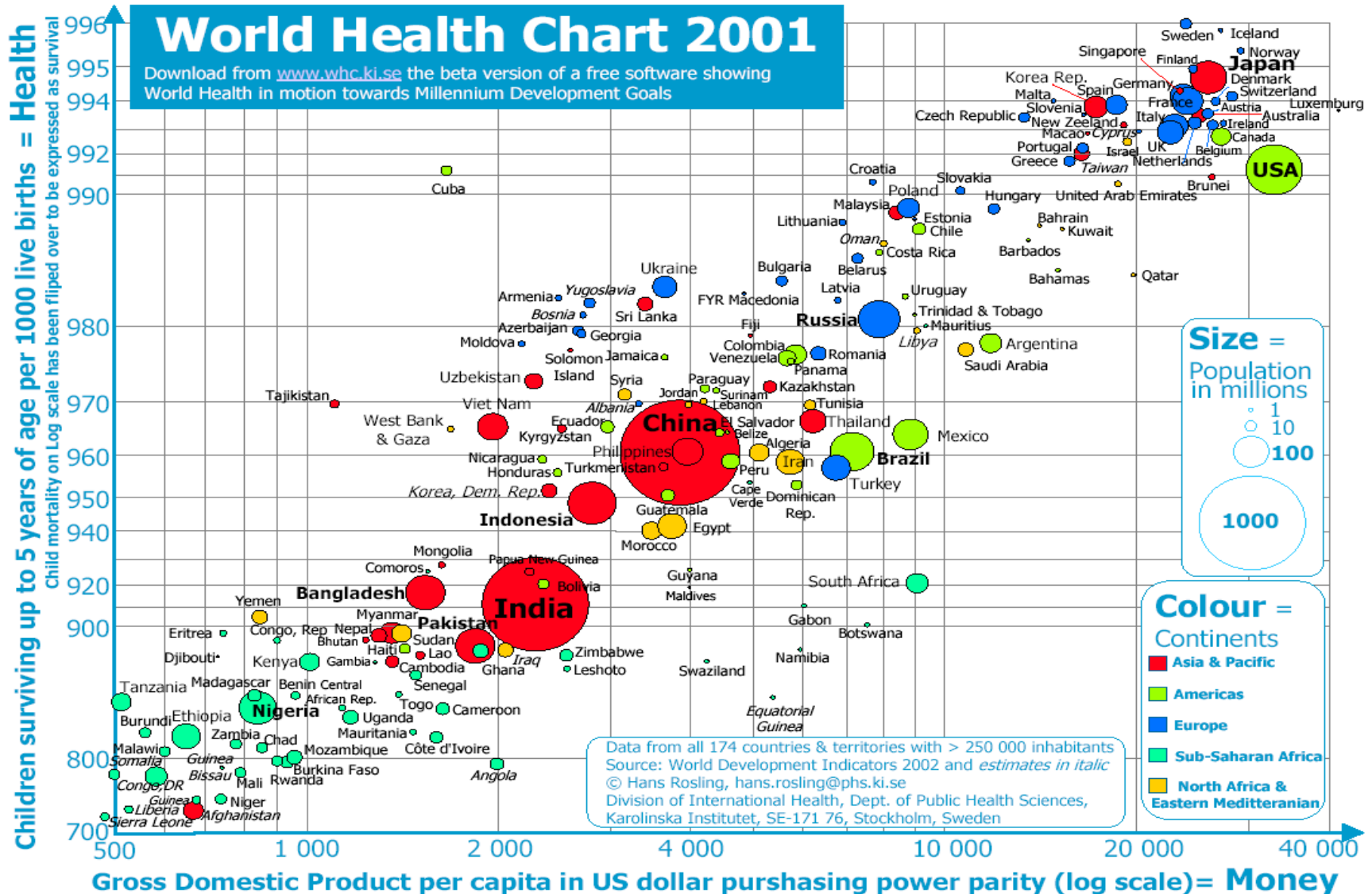
To manipulate data properly we may first **pre-process** it:

- **Data cleaning**: a process that removes or transforms noise and inconsistent data
- **Data integration**: where multiple data sources may be combined (also known as Data Fusion)
- **Data selection**: where data relevant to the analysis task are retrieved, filtered, extracted

Then we are ready for data representation:

- **Data transformation**: where data are transformed, reduced or consolidated into forms appropriate for alternative representation and/or further analysis.

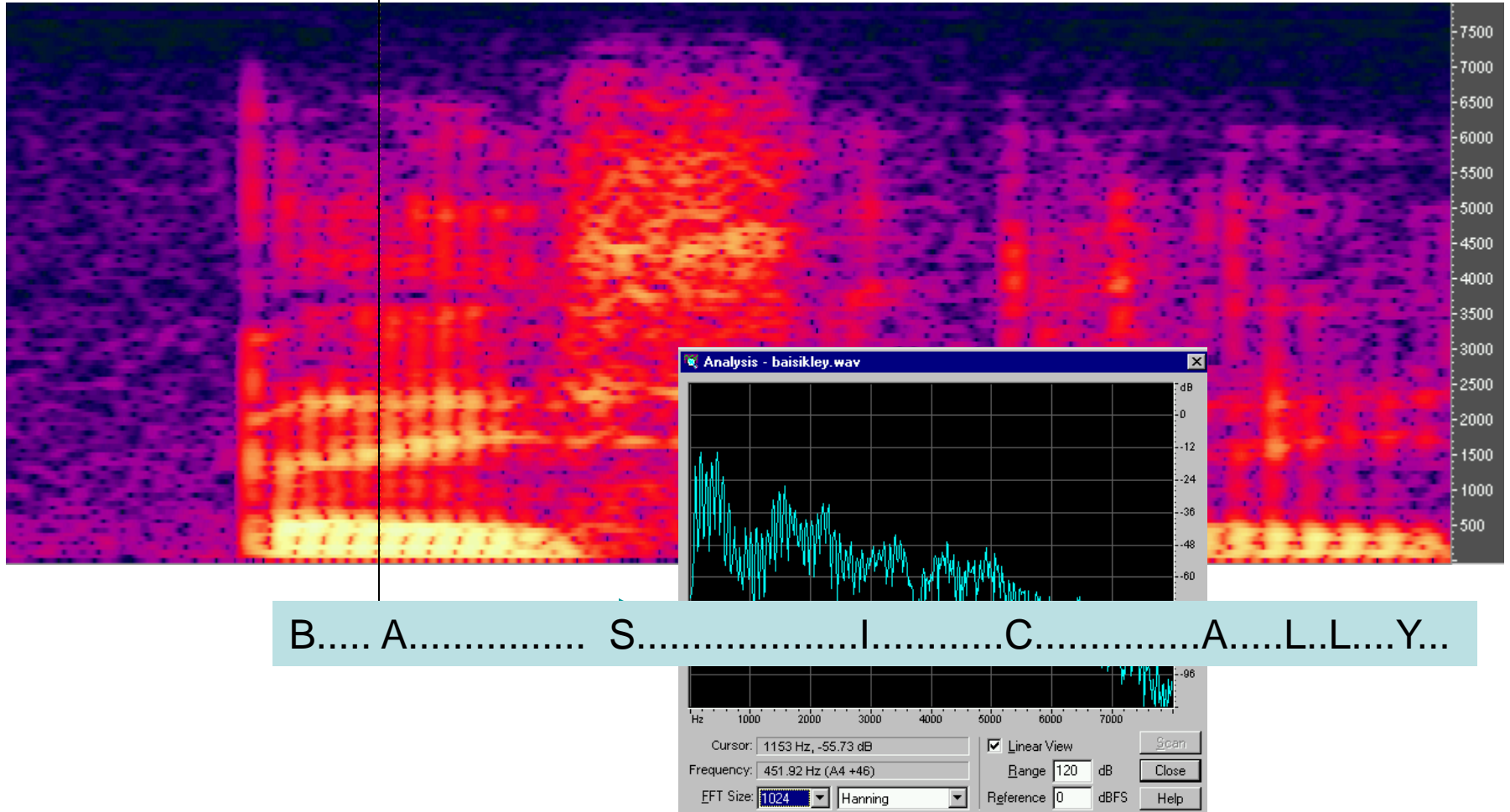
# Visualizing Data





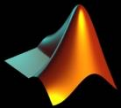
# Frequency Domain Data Analysis

**Spectrogram:** Representation of time, frequency and amplitude





# Spatial Domain Data Analysis: Cleaning/Clearing up Data



Sometimes we may manipulate data just so we (humans) can see the data better.



*Noisy Gene Sequence:*

GGATACAWCTTTAGAG



*Cleaned Gene Sequence:*

GGATACAACTTTAGAG



# Spatial Domain Data Analysis: Feature Detection



Edge Detection

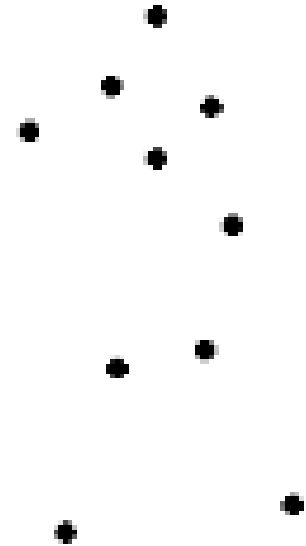


Corner Detection



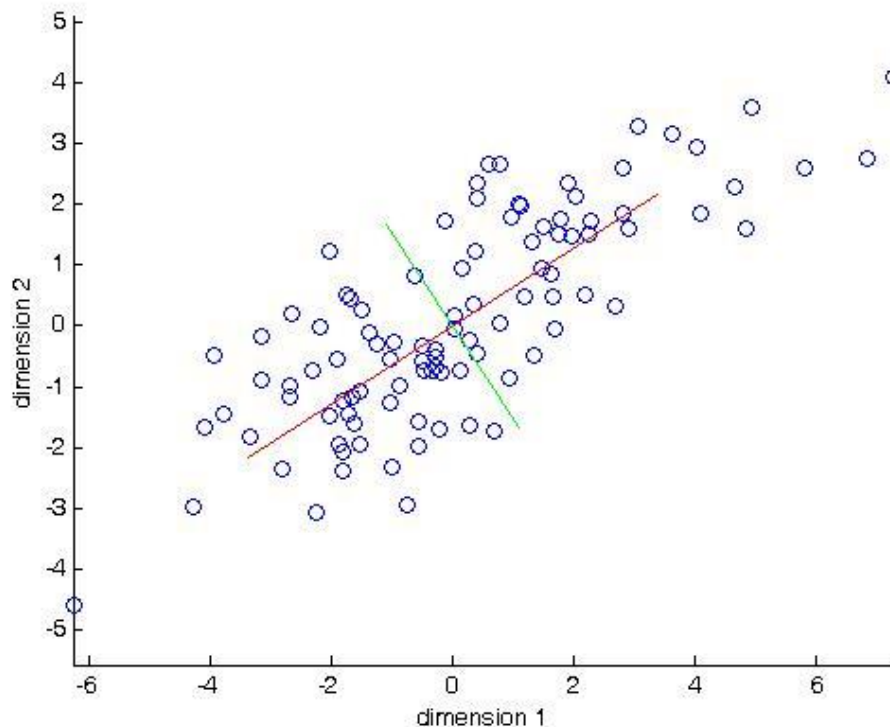
# Features help simplify the problems

- Even “impoverished” motion data can evoke a strong percept
- Some tracking examples



# Principal Component Analysis

The two principal eigenvectors demonstrate the orthogonal directions of maximum variation in the data.



Before :

$$\mathbf{C} = \begin{pmatrix} 0.258 & 0.314 \\ 0.314 & 0.403 \end{pmatrix}$$

After :

$$\mathbf{C} = \begin{pmatrix} 0.518 & 0 \\ 0 & 0.174 \end{pmatrix}$$



# Coordinate Transformations

Transforming data from one coordinate system to another for representation



*Orthographic Projection*



*Perspective Projection*

# Signals and Functions

A signal is a physical quantity that is a function of one or more independent variable(s), such as space and/or time.

Data from a Gene pool

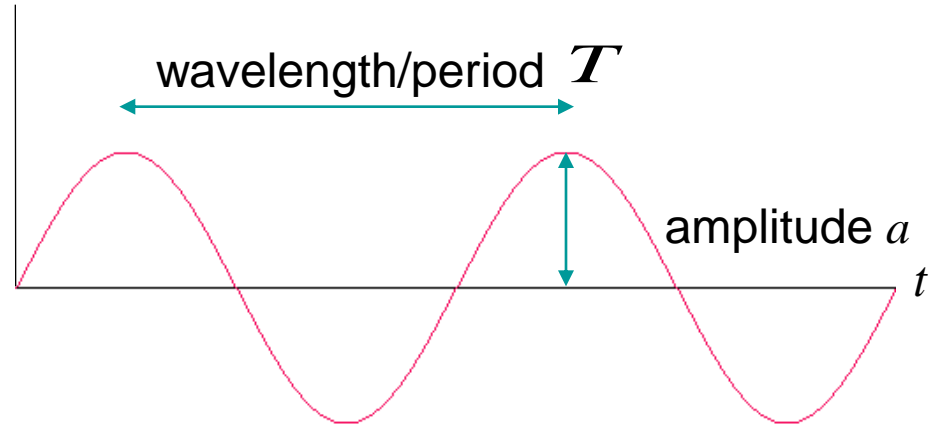
Position of a car in a video sequence

Example signals:

1D signal:  $f(t)$

2D signal:  $f(x,y)$

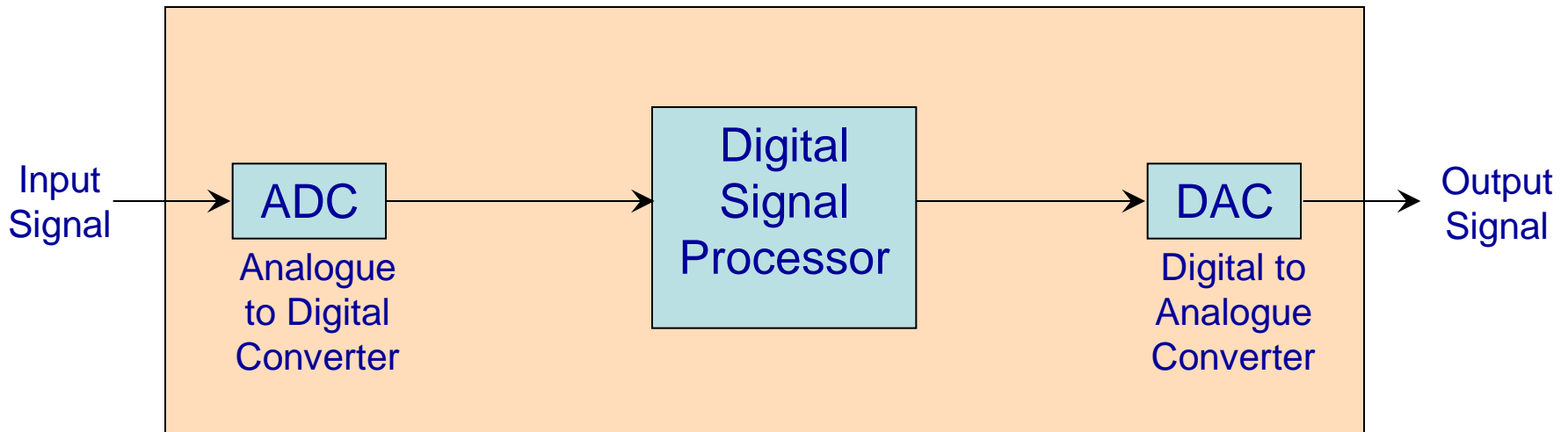
3D signal:  $f(x,y,t)$  etc.





# What is DSP?

- **Digital Signal Processing** – the processing or manipulation of signals using digital techniques

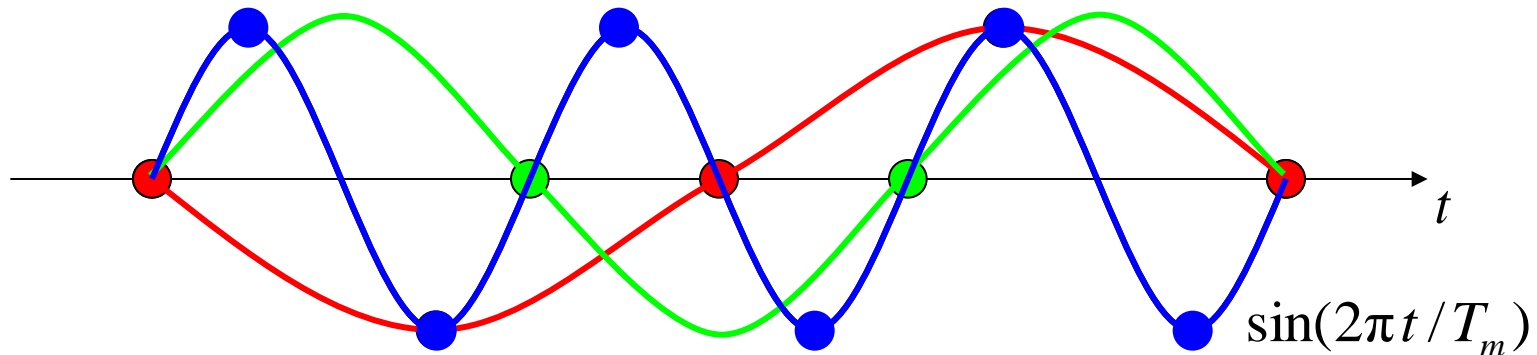




# Shannon's Sampling Theorem

*“An analogue signal containing components up to some maximum frequency  $u$  (Hz) may be completely reconstructed by regularly spread samples, provided the sampling rate is at least  $2u$  samples per second”*

Also referred to as the Nyquist criterion: sampling frequency should be at least twice the highest spatial frequency.



# Sampling

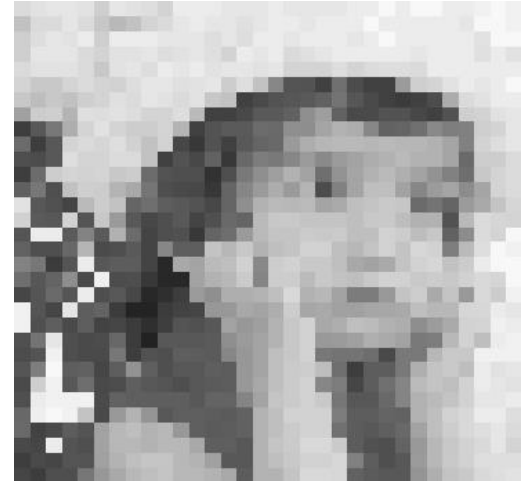
The effect of sparser sampling...is **ALIASING**



256 x256



64x64



32x32

Anti-aliasing achieved by filtering to remove frequencies above Nyquist limit.

# Quantization

This results from representing a continuously varying function  $f(x)$  with a discrete one using quantization levels



16 levels



6 levels



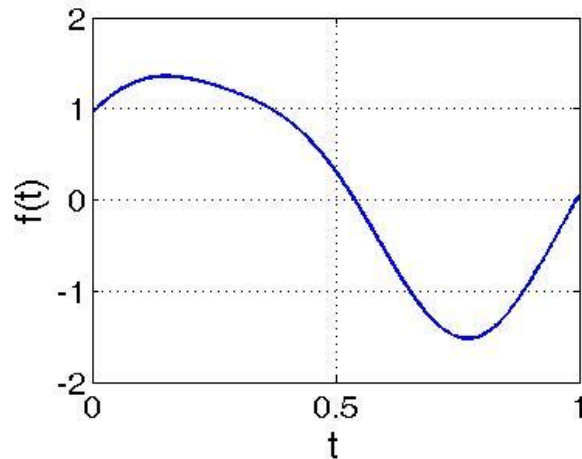
2 levels

- Matlab code: 

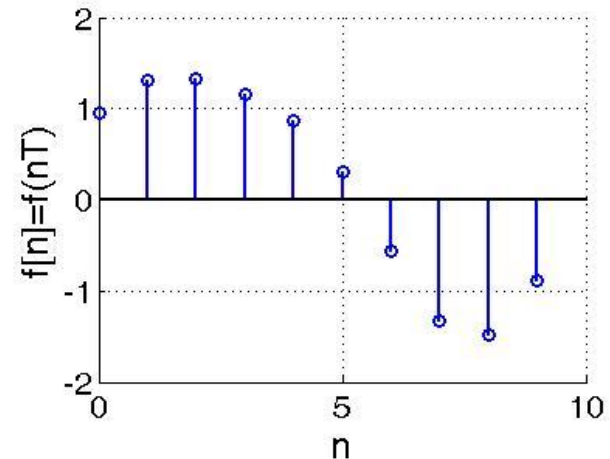
```
F = imread('romina.gif');  
[X, map] = gray2ind(F, 16); // 2, 6, or 16  
imview(X, map);
```

# Signal Processing

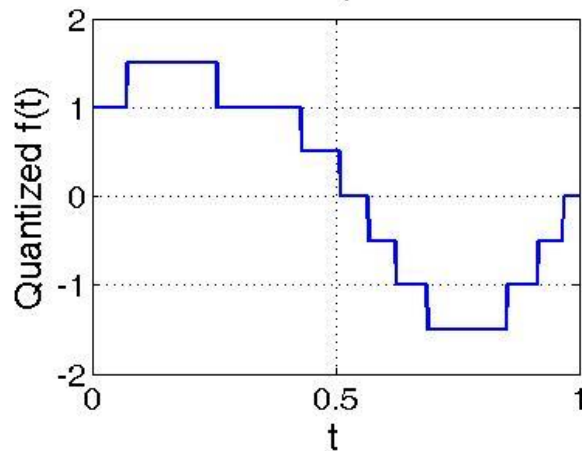
Continuous Time, Continuous Value



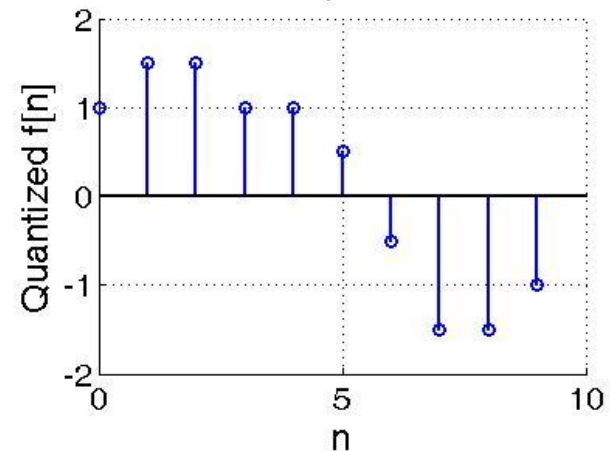
Discrete Time, Continuous Value



Continuous Time, Discrete Value



Discrete Time, Discrete Value



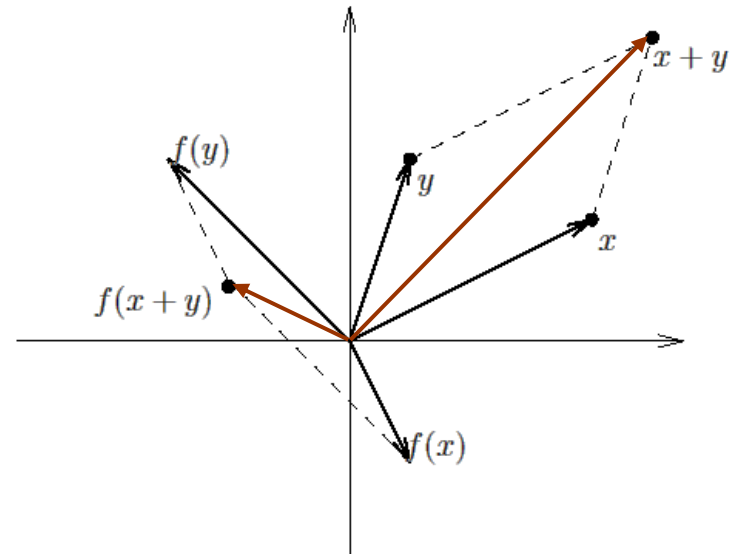
# Linear Systems

- For a linear system: output of the linear combination of many input signals is the same linear combination of the outputs → *superposition*

A function  $f$  is linear if

- $f(x + y) = f(x) + f(y)$
- $f(\alpha x) = \alpha f(x)$

i.e., superposition holds.

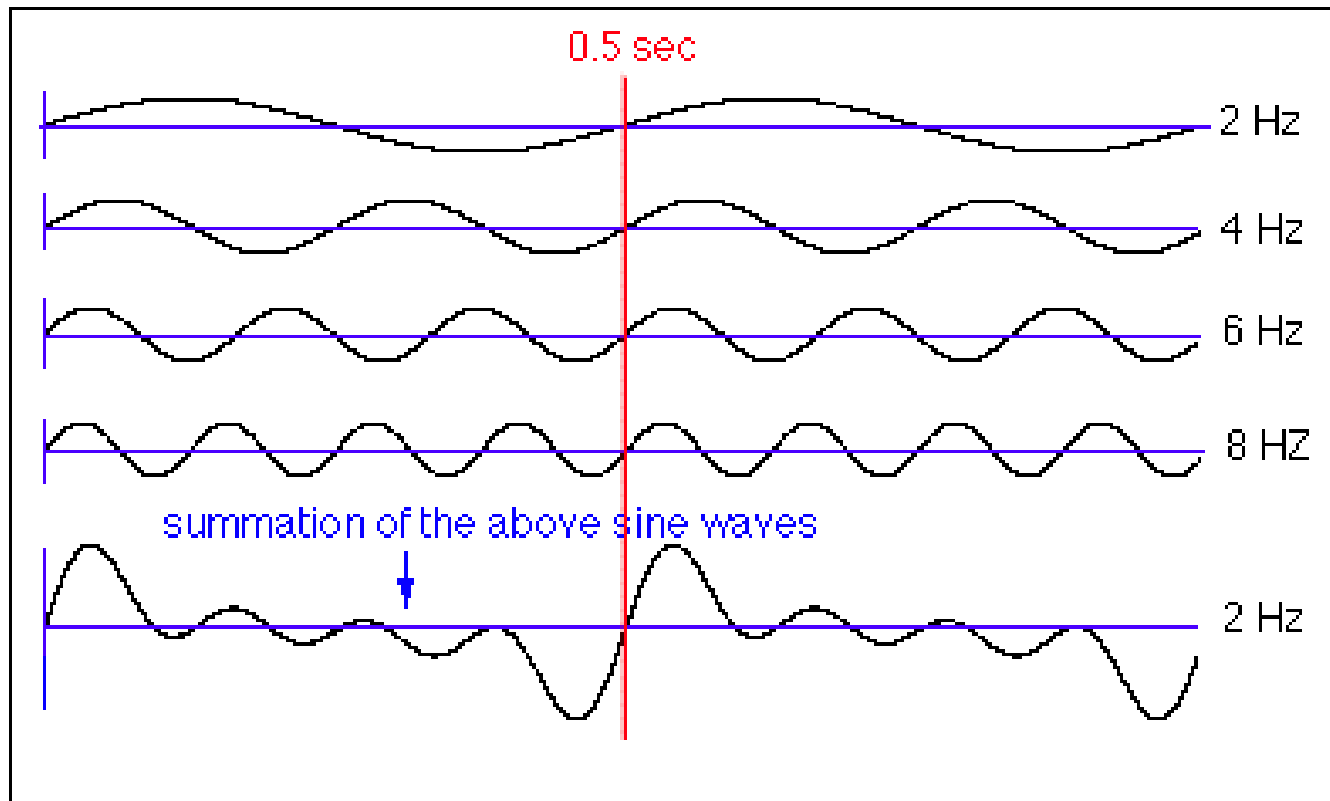


Linearity allows us to decompose our input into smaller, elementary objects. Output is the sum of the system's response to these basic objects.



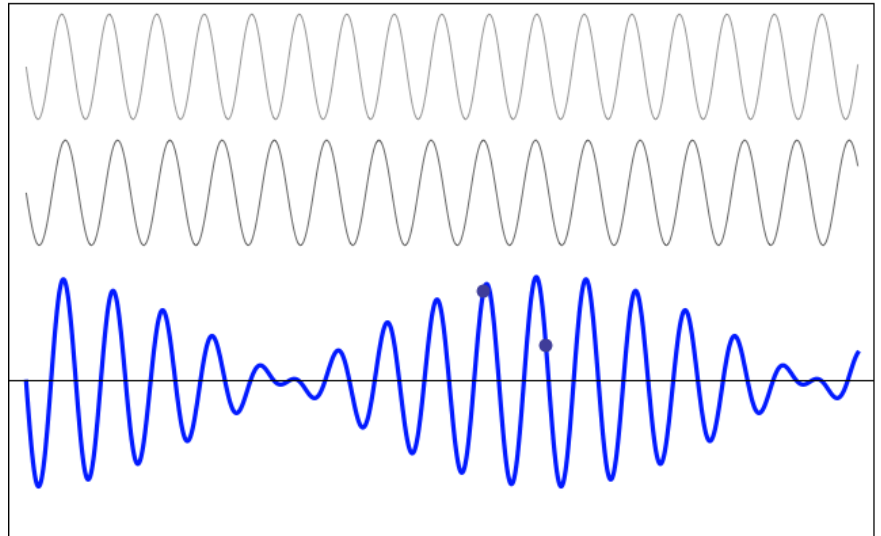
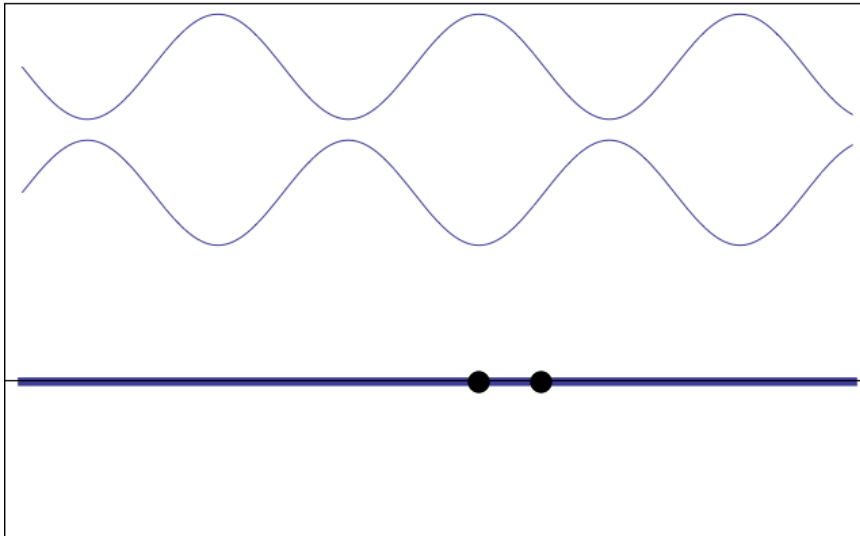
# Linear Systems

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# Linear Systems

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# Example: White Light?

White light is made up of variable wavelengths of each component color.

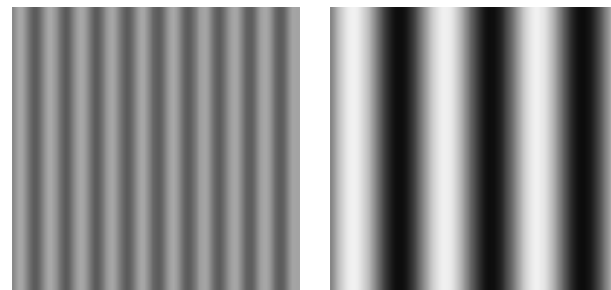


AAAAATAAAAA  
0000001000000

$$\delta[n] = \begin{cases} 0, n \neq 0 \\ 1, n = 0 \end{cases}$$

# Basic signals...

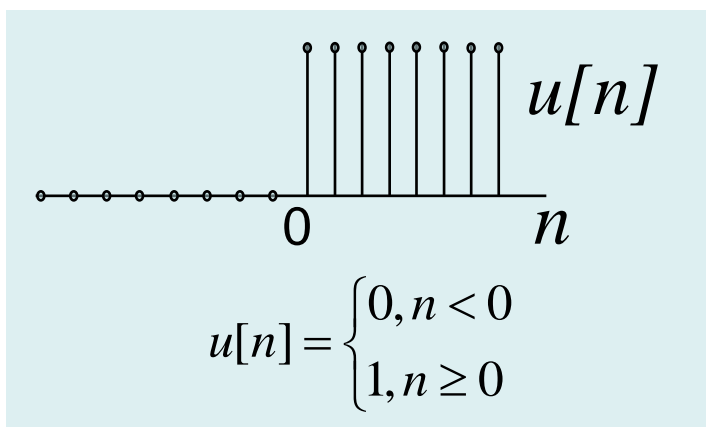
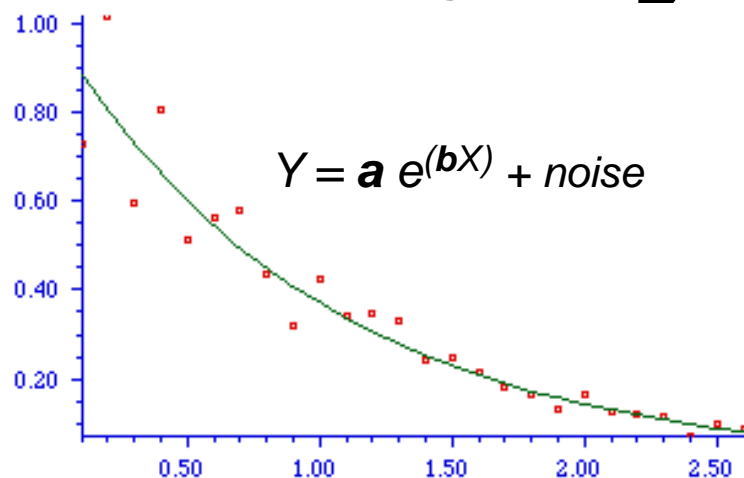
$$x = \sin(t) = \sin(t+2\pi)$$



Some basic signals:

- Unit impulse signal
- Unit step signal
- Exponential signal
- Periodic signal

All signals can be represented by these basic signals!



# Overview of next few lectures

- Fourier Series
- 1D and 2D Fourier Transform
- Convolution
- Feature Selection and Extraction
- PCA
- Coordinate transformations