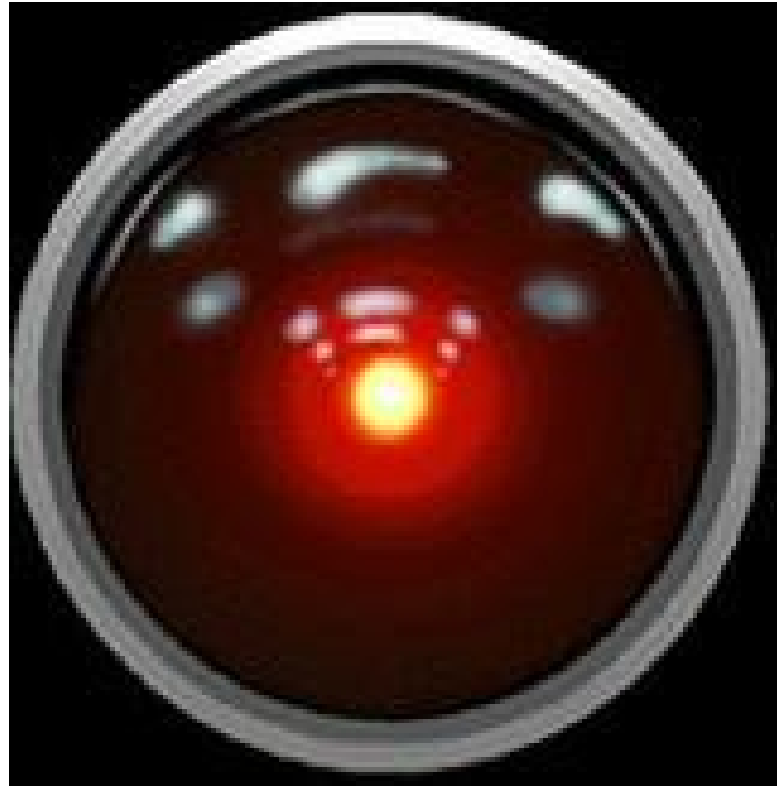




CS 378: Autonomous Intelligent Robotics

Instructor: Jivko Sinapov

<http://www.cs.utexas.edu/~jsinapov/teaching/cs378/>

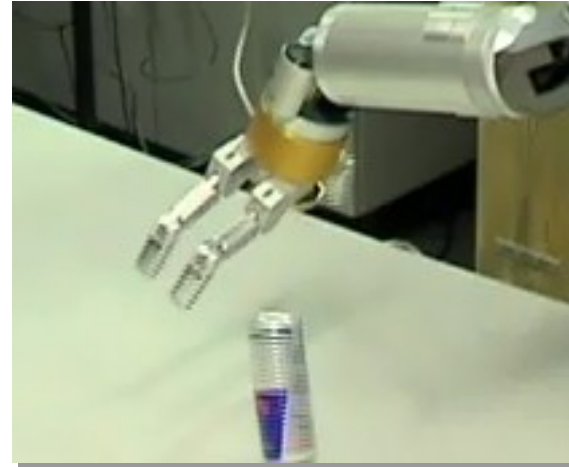
Audio Processing and Computational Perception of Natural Sound

Why Sound?

Why Sound?

What actually happened:

The robot dropped a soda-can



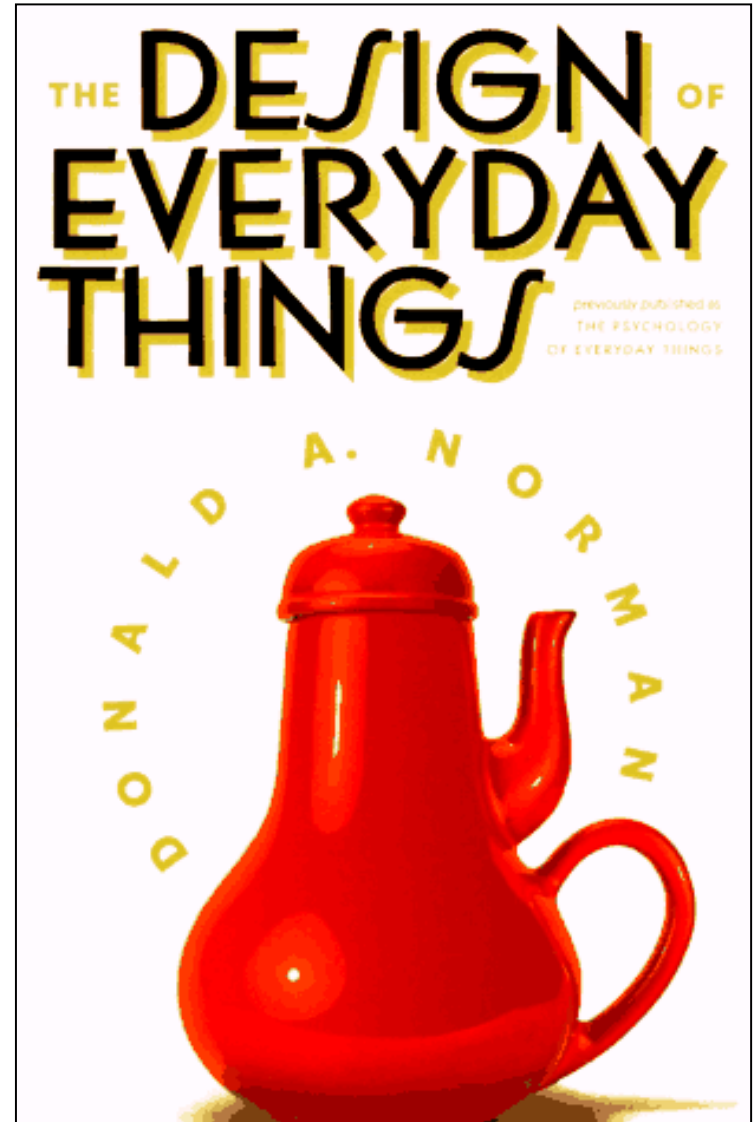
Why Natural Sound is Important

“...natural sound is as essential as visual information because **sound tells us about things that we can't see**, and it does so while our eyes are occupied elsewhere. “

“Sounds are generated when materials interact, and the sounds tell us whether they are **hitting, sliding, breaking, tearing, crumbling, or bouncing**. “

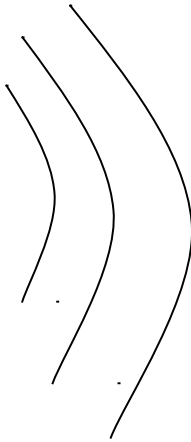
“Moreover, **sounds differ according to the characteristics of the objects**, according to their size, solidity, mass, tension, and material. “

Don Norman, *“The Design of Everyday Things”*, p.103



Why Natural Sound is Important

Sound Producing Event



[Gaver, 1993]

Why should a robot use acoustic information?

Human environments are cluttered with objects that generate sounds

Help a robot perceive events and objects outside of field of view

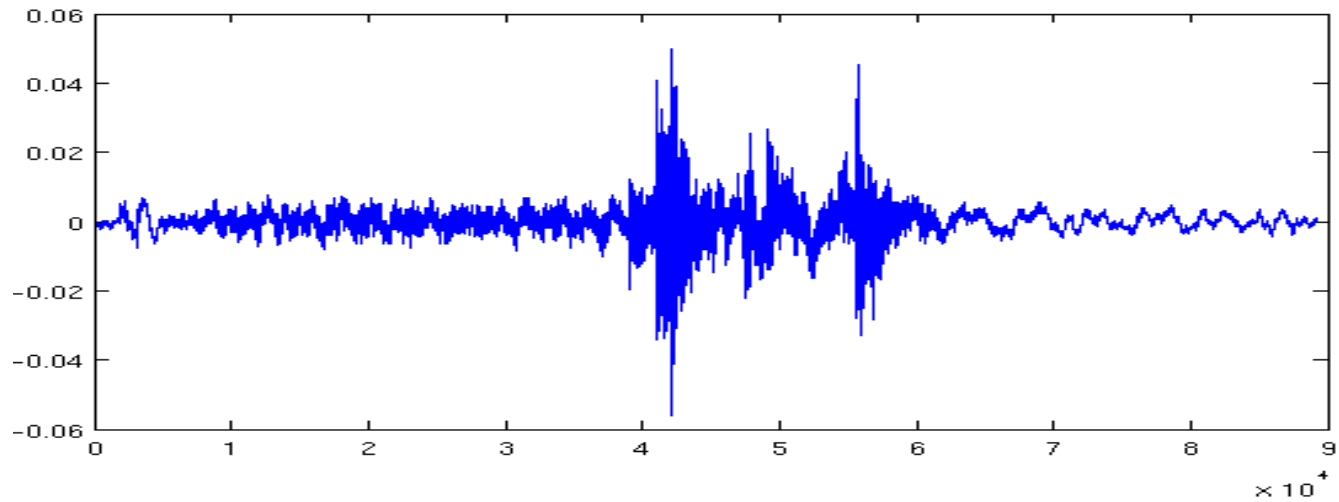
Help a robot perceive material properties of objects, and form natural object categories

What is Sound?

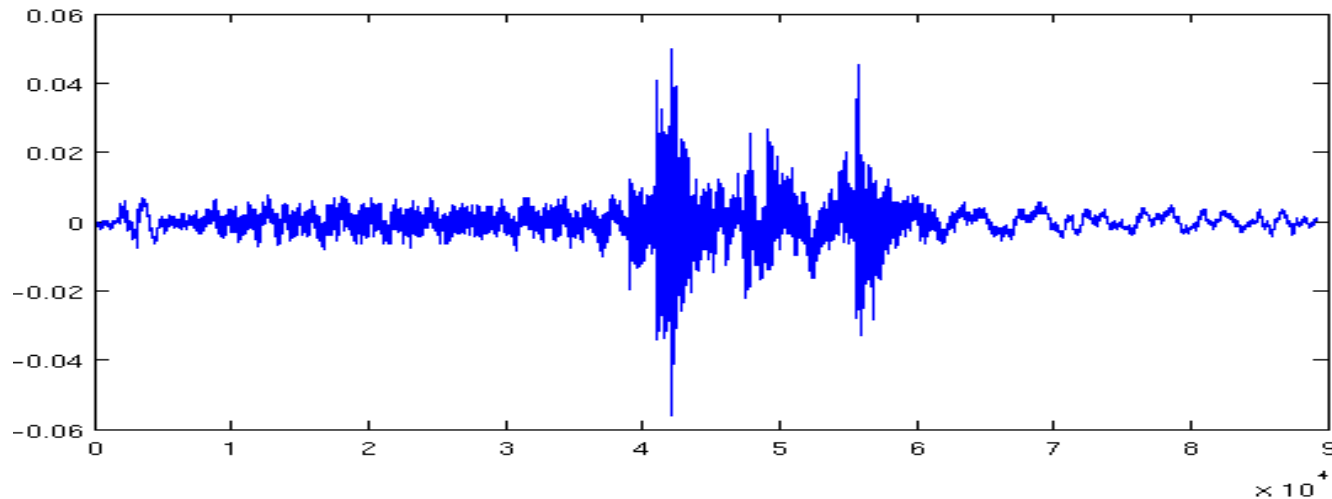
What is Sound?



What is Sound?

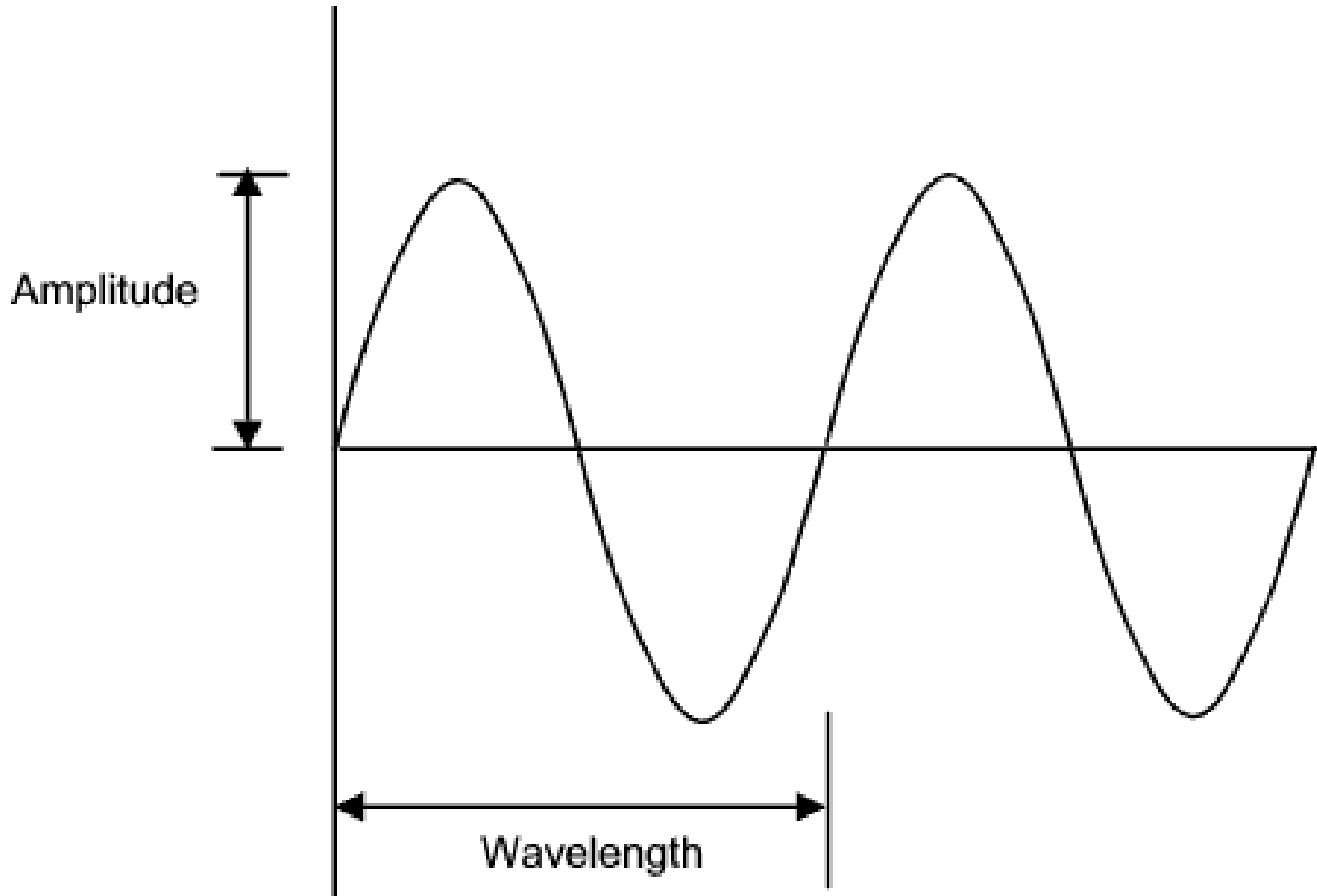


What is Sound?

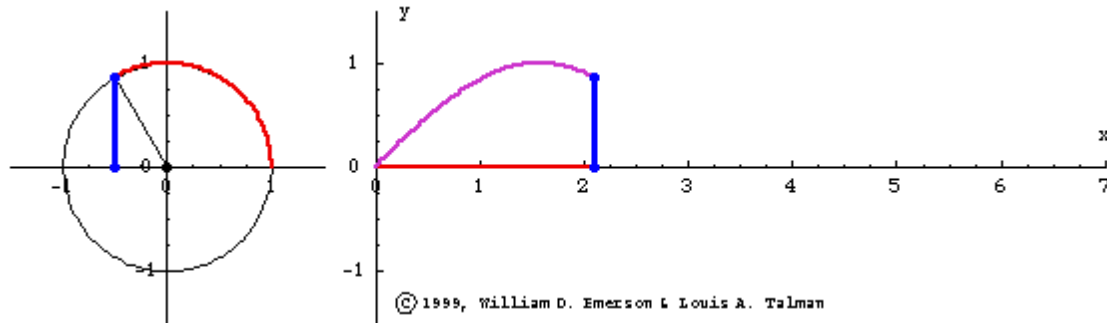


....from a computer's point of view, raw audio is a sequence of 44.1K floating point numbers arriving each second

Sine Wave



Sine Curve



Frequency

- Measured in Hertz (Hz)
- Named after Heinrich Hertz
- 1 Hertz = 1 repetition per second
- Typically denoted with the letter f

Period

- How long does one cycle take?
- It is the reciprocal of the frequency
- Measured in seconds
- Typically denoted with the letter T

Frequency vs Period Animation



$$f = 0.5 \text{ Hz}$$
$$T = 2.0 \text{ s}$$



$$f = 1.0 \text{ Hz}$$
$$T = 1.0 \text{ s}$$



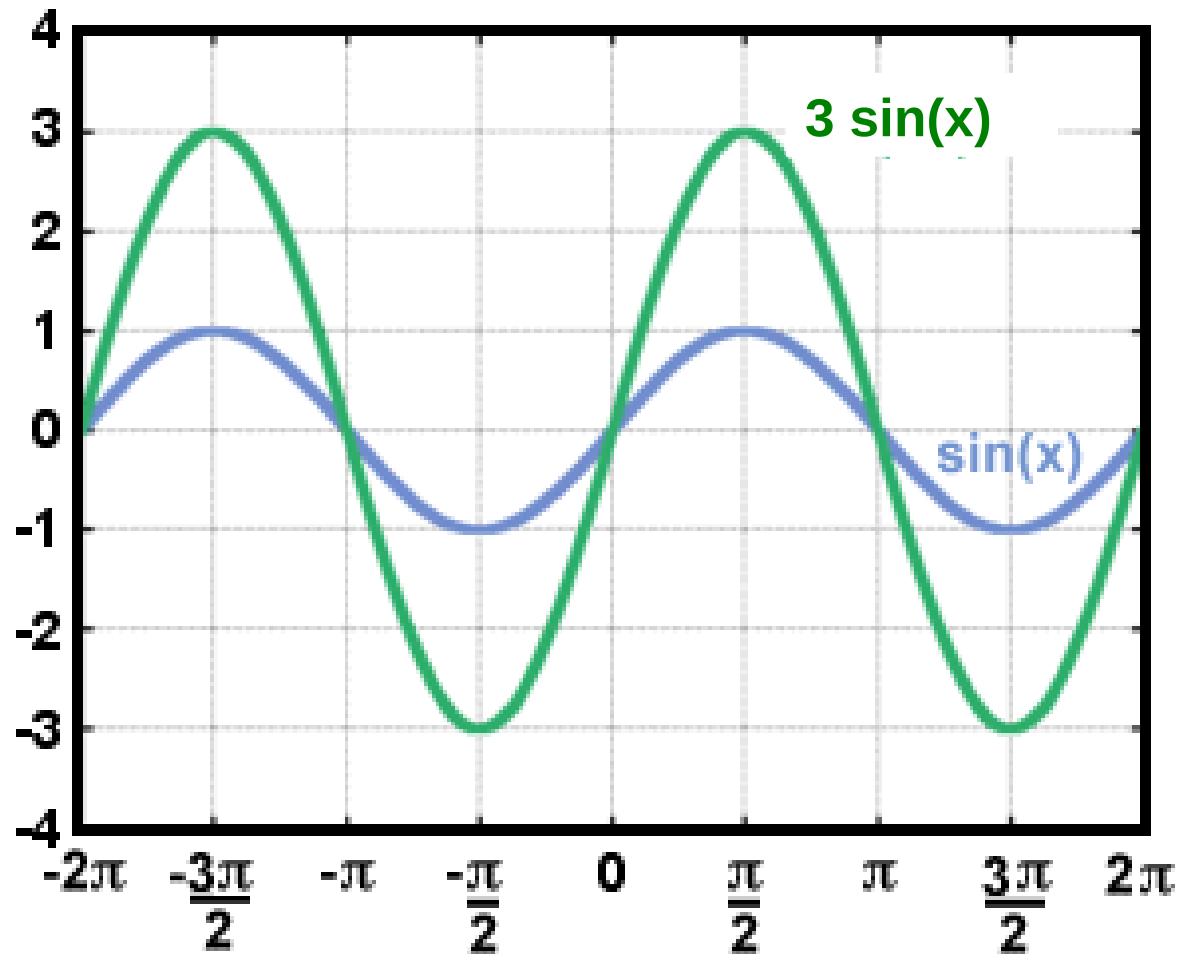
$$f = 2.0 \text{ Hz}$$
$$T = 0.5 \text{ s}$$

Frequency vs Period

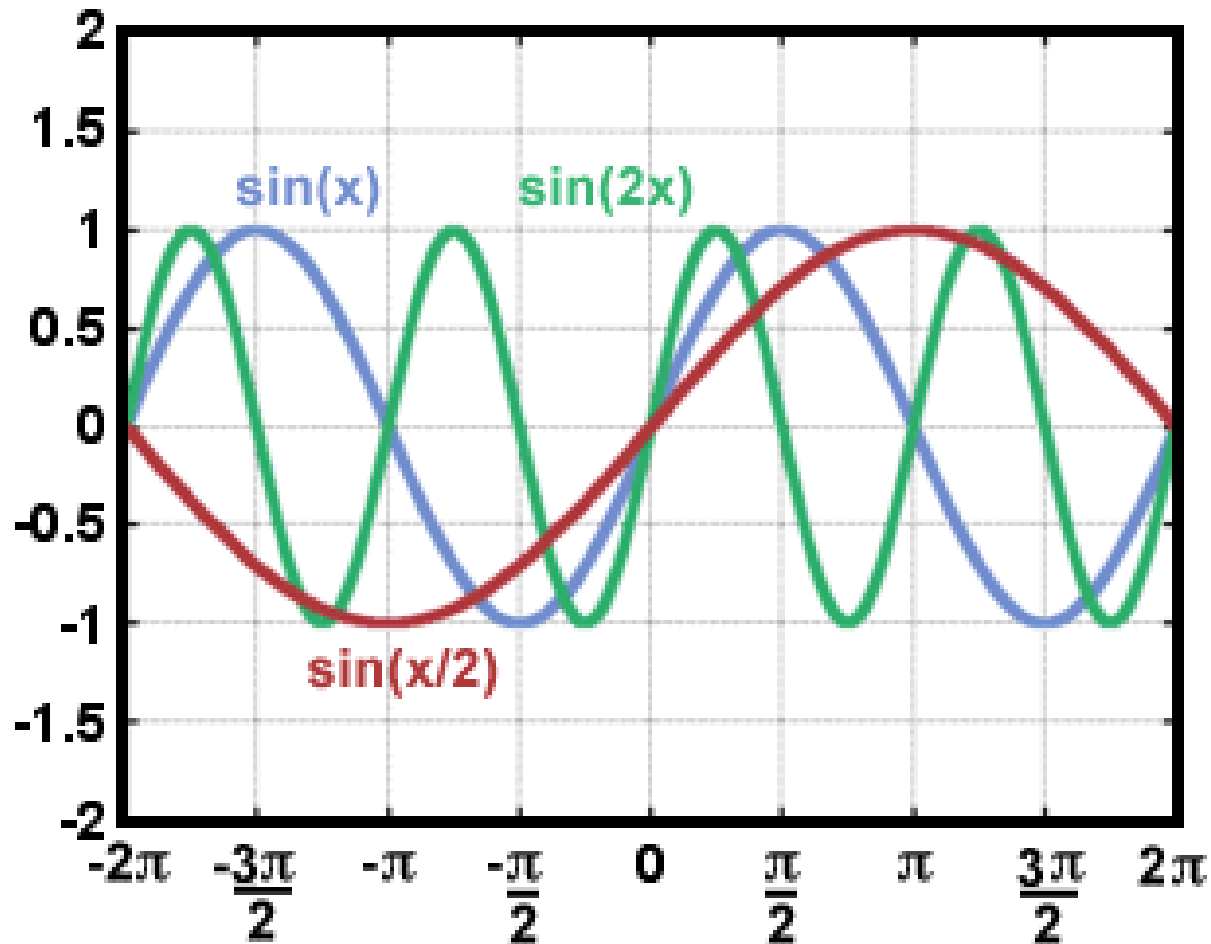
Frequency	1 mHz (10 ⁻³)	1 Hz (10 ⁰)	1 kHz (10 ³)	1 MHz (10 ⁶)	1 GHz (10 ⁹)	1 THz (10 ¹²)
Period (time)	1 ks (10 ³)	1 s (10 ⁰)	1 ms (10 ⁻³)	1 μs (10 ⁻⁶)	1 ns (10 ⁻⁹)	1 ps (10 ⁻¹²)

$$T = \frac{1}{f}$$

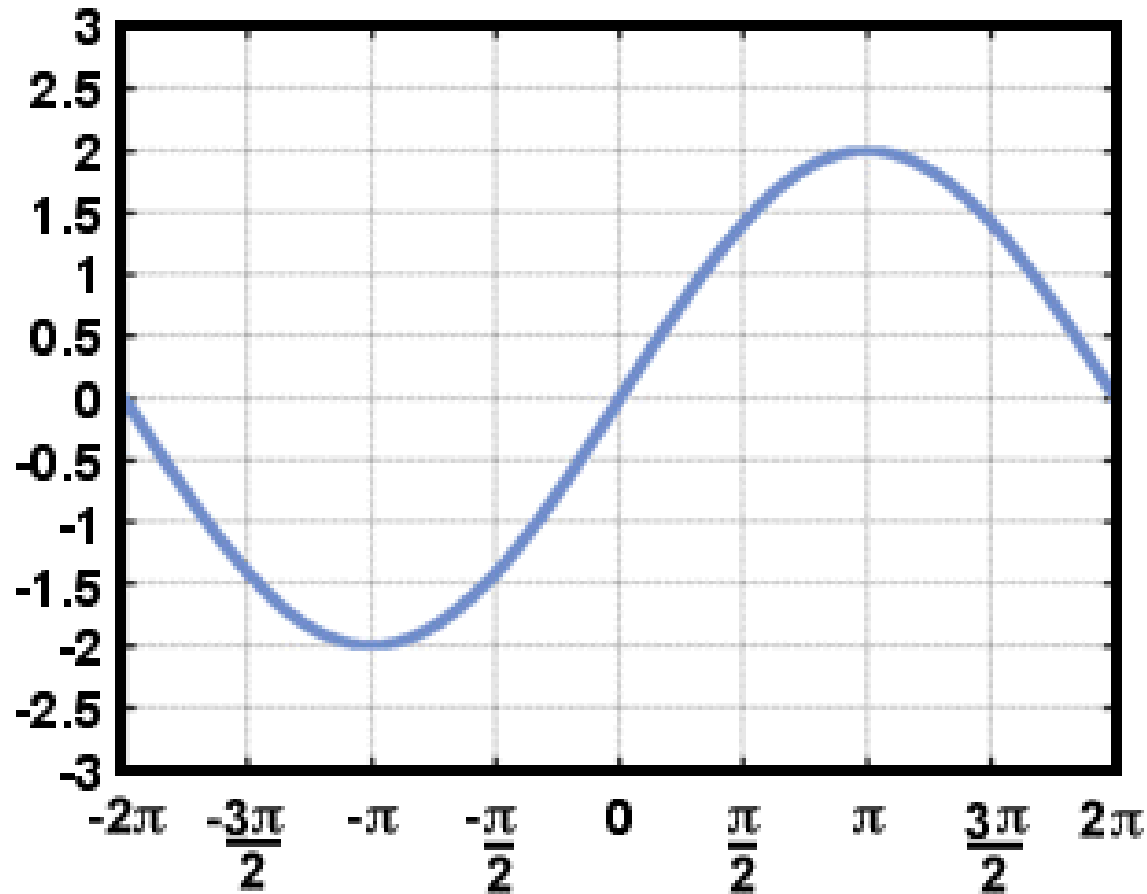
Amplitude (vertical stretch)



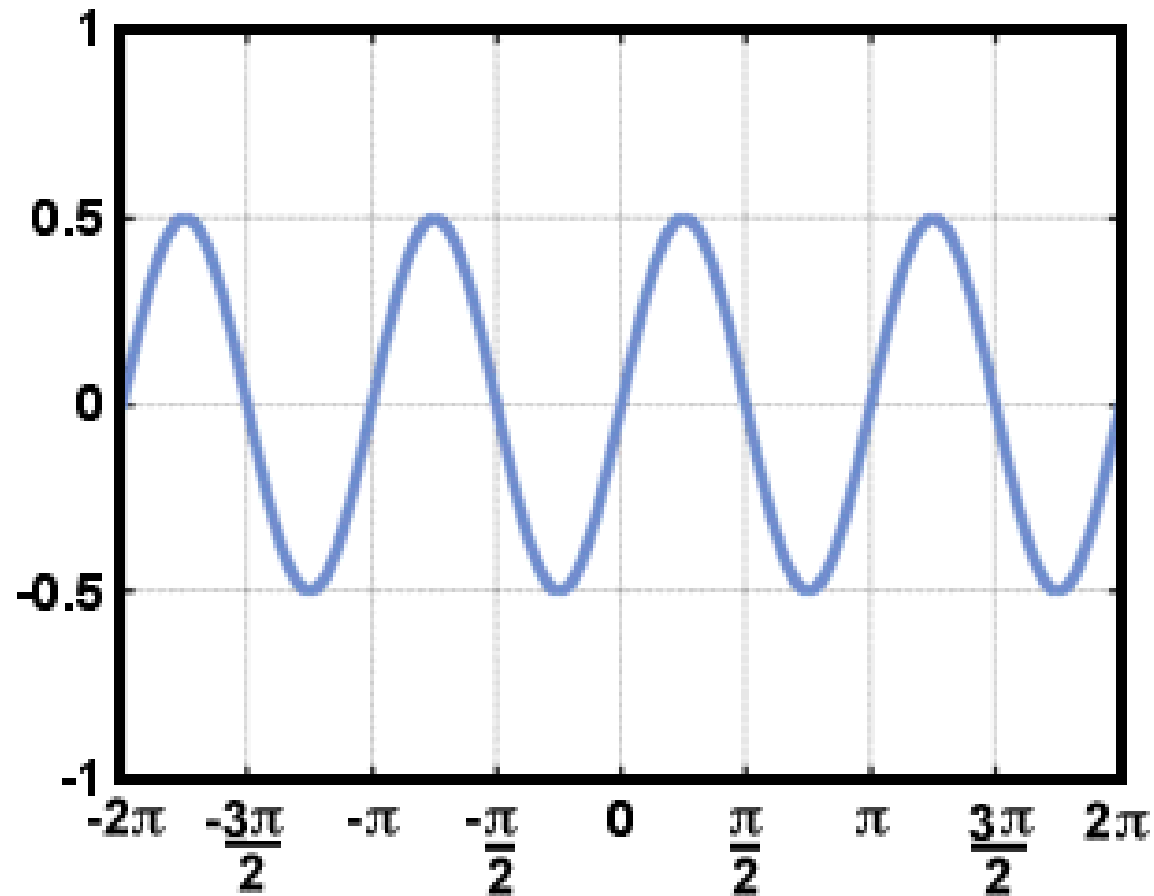
Frequency (horizontal stretch)



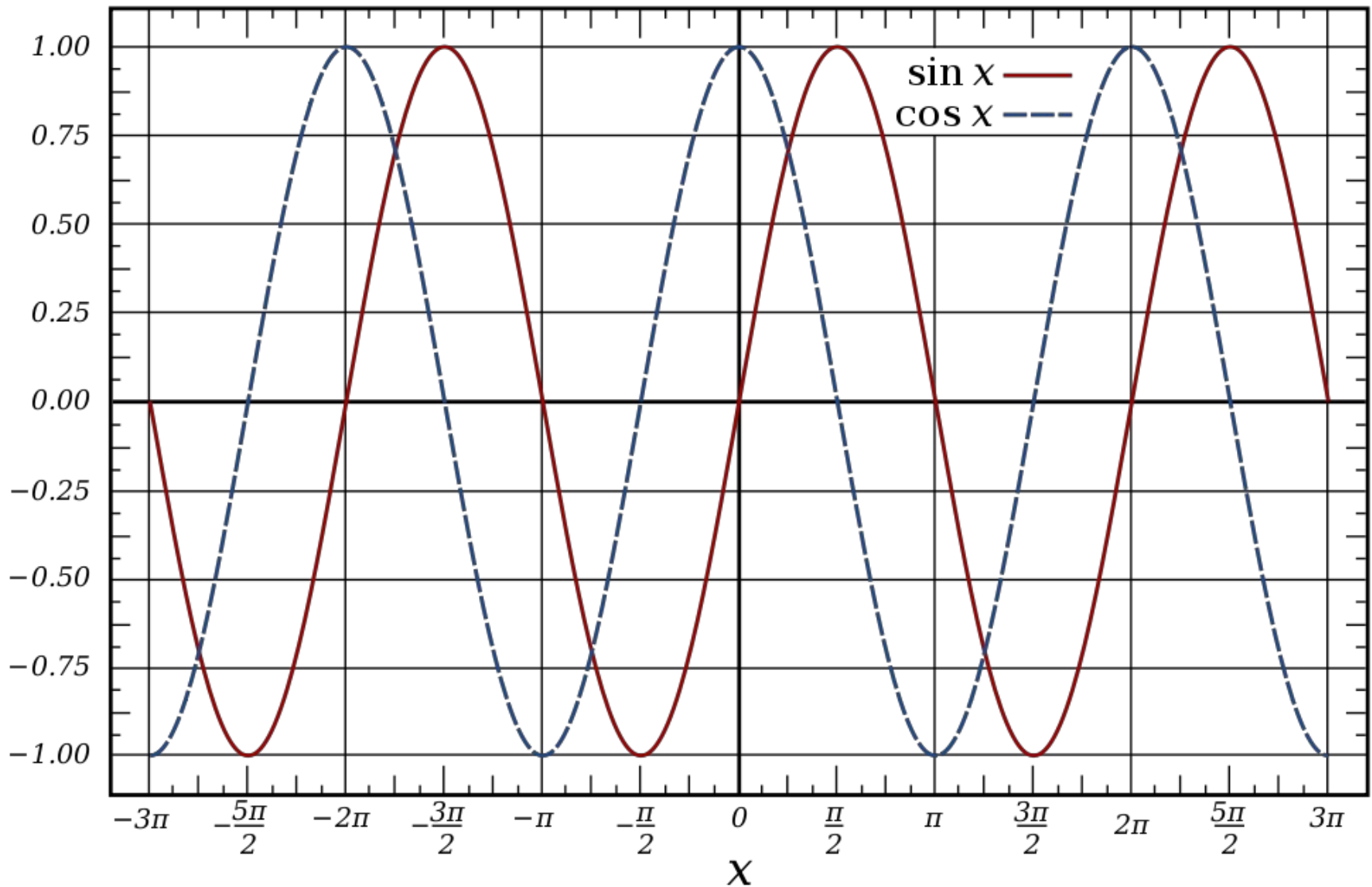
What is the Period and the Amplitude?



What is the Period and the Amplitude?



Sines vs Cosines



Formula for the Sine Wave

$$y(t) = A \cdot \sin(\omega t + \phi)$$

Formula for the Sine Wave

$$y(t) = A \cdot \sin(\omega t + \phi)$$

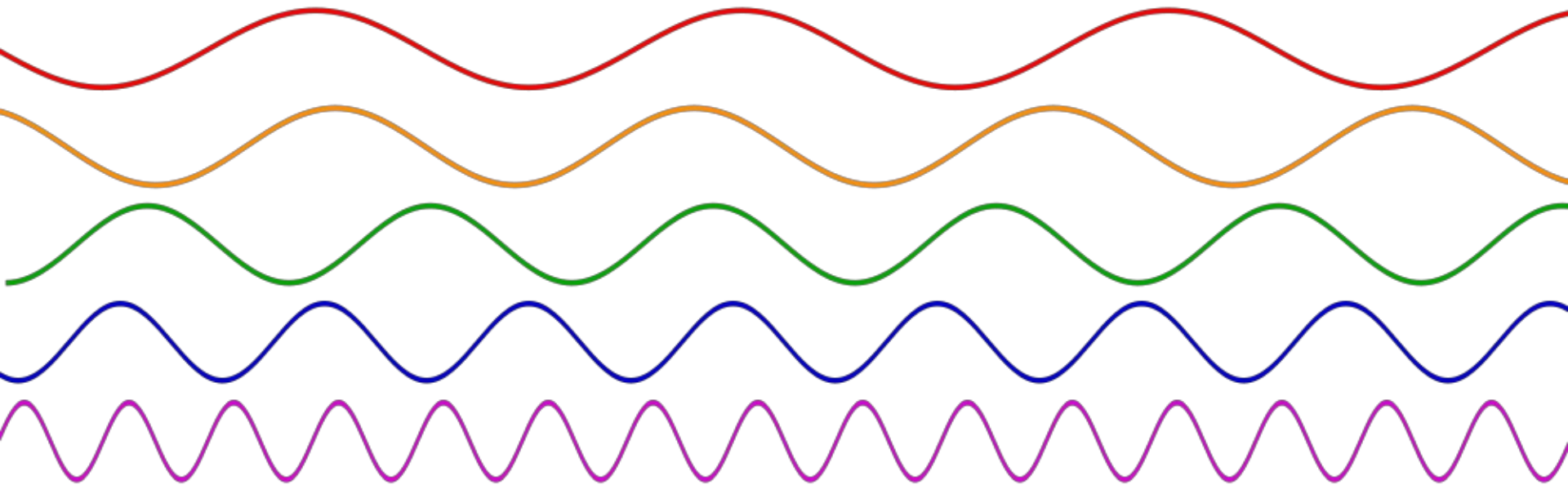
- A , the amplitude, is the peak deviation of the function from its center position.
- ω , the angular frequency, specifies how many oscillations occur in a unit time interval, in radians per second
- ϕ , the phase, specifies where in its cycle the oscillation begins at $t = 0$.

A function $x(t)$ is periodic if we can find a T for which the following hold

$$x(t) = x(t + T) = x(t + 2T) = x(t + 3T) = \dots$$

Sinusoidal waves of various frequencies

Low Frequency

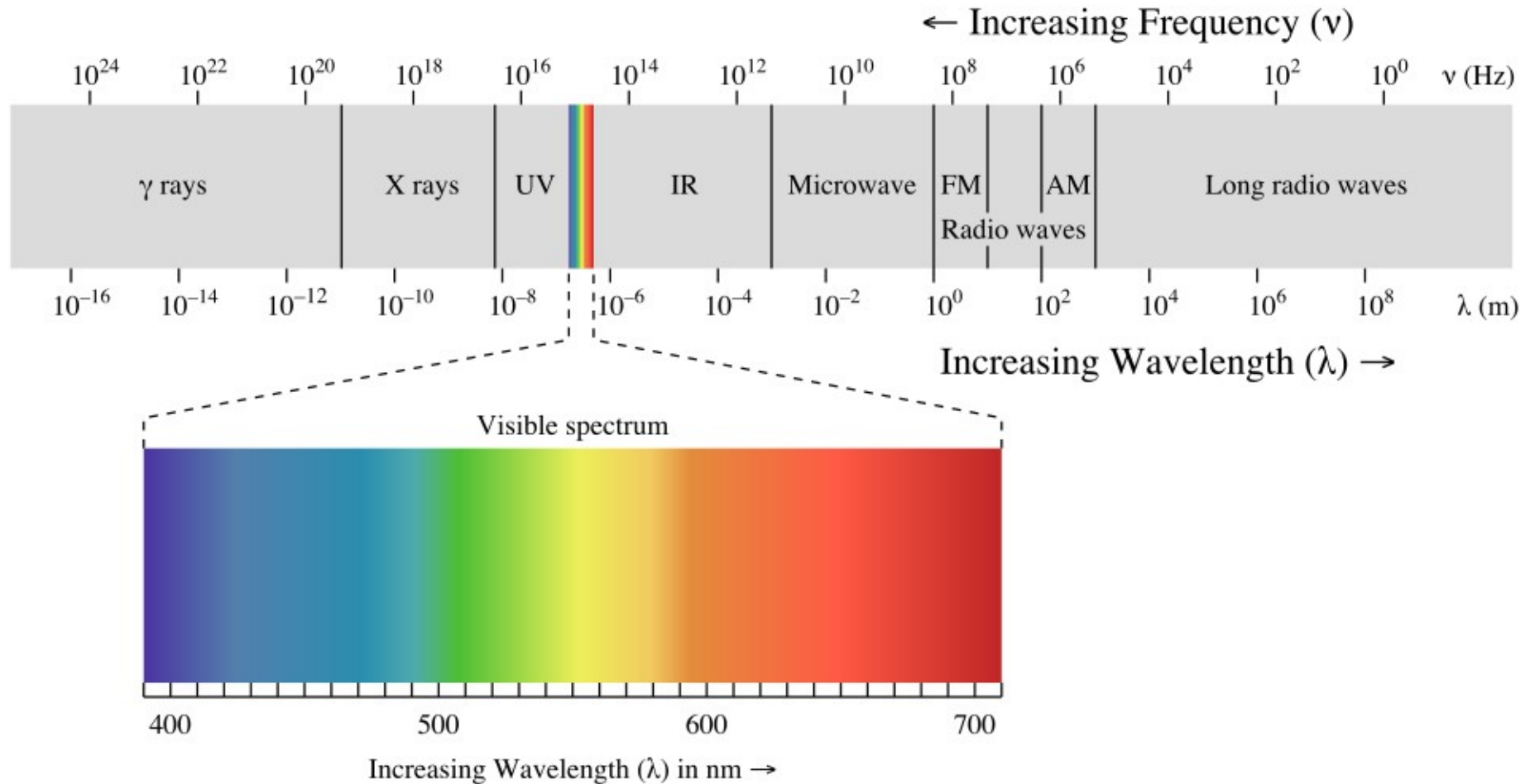


High Frequency

Spectrum

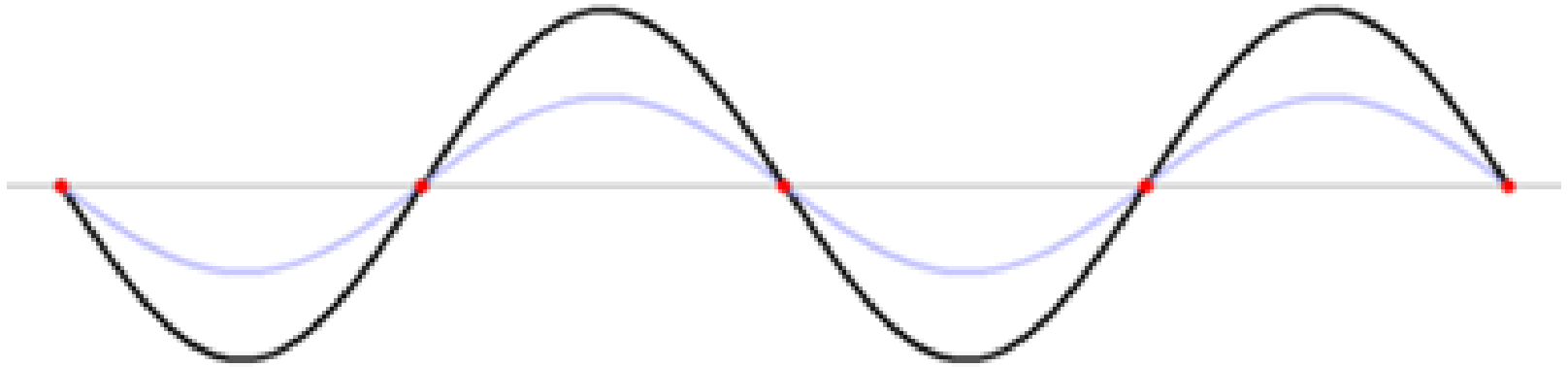


Light Spectrum



Standing Wave

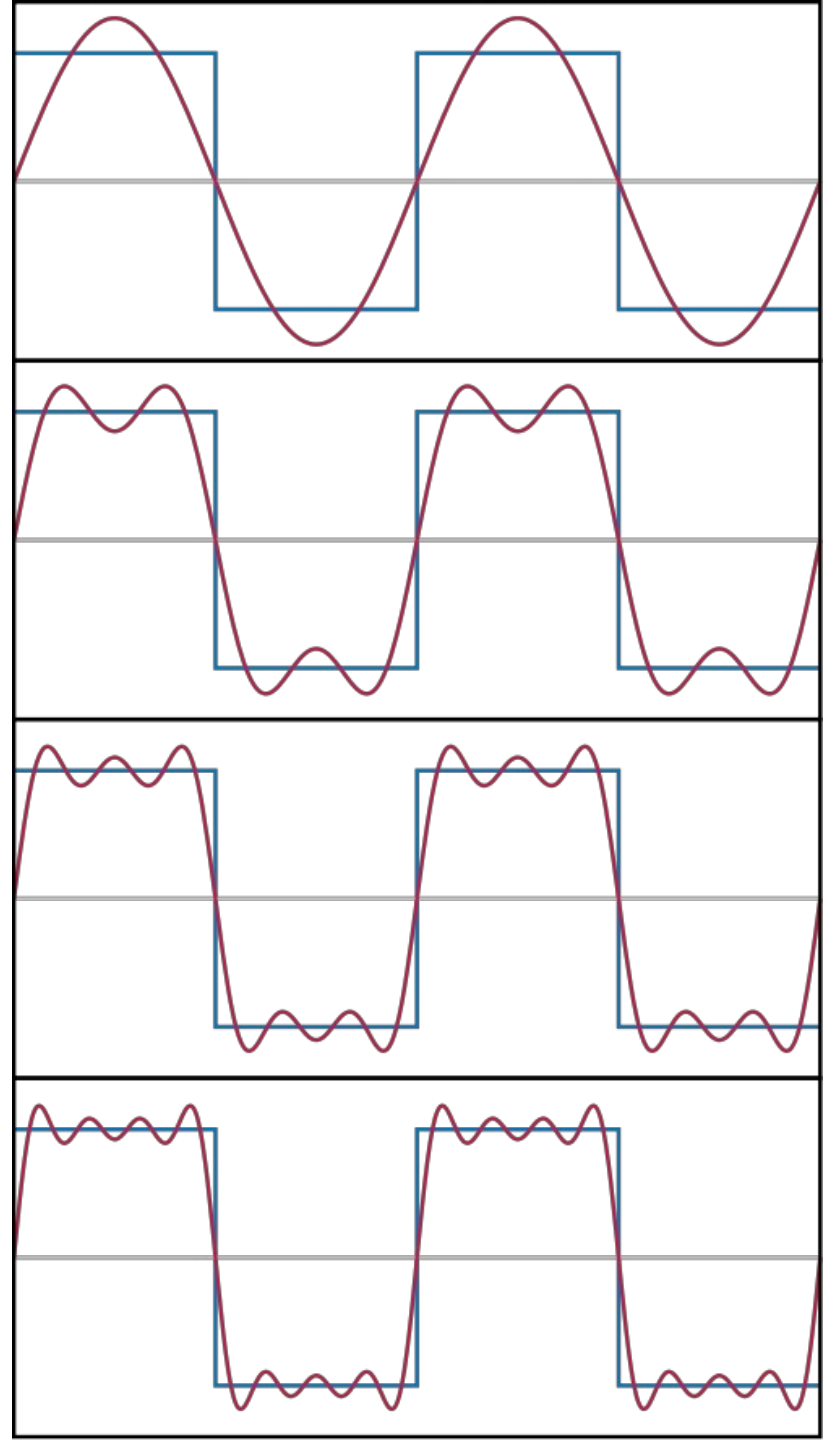
(shown in black, equal to the sum of the red and the blue waves traveling in opposite directions)



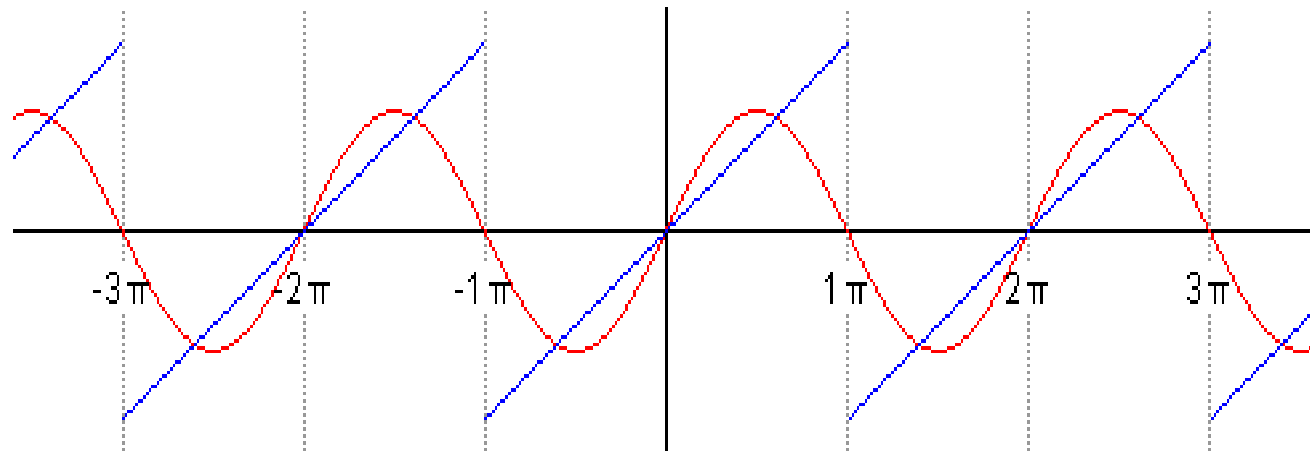
Fourier Series

A Fourier series decomposes periodic functions or periodic signals into the sum of a (possibly infinite) set of simple oscillating functions, namely sines and cosines

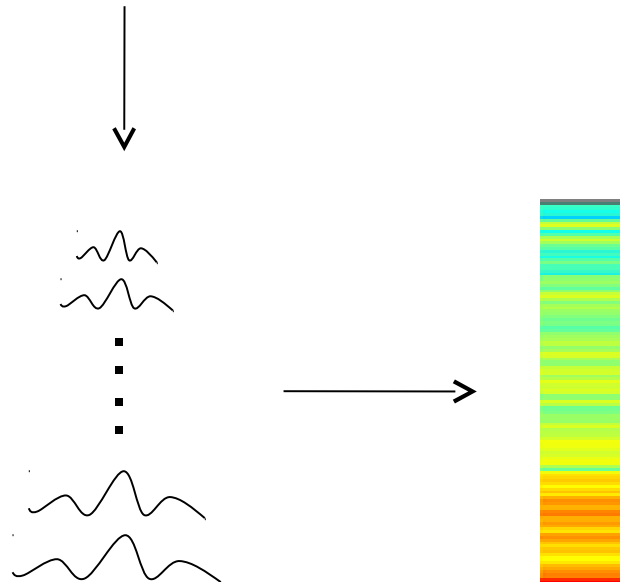
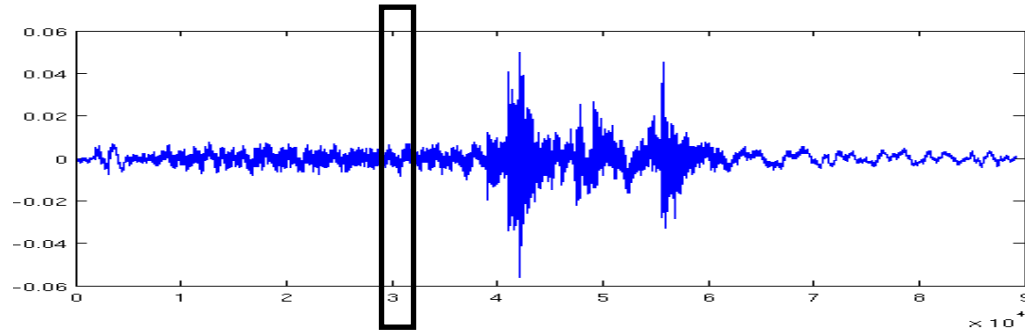
Approximation



Approximation

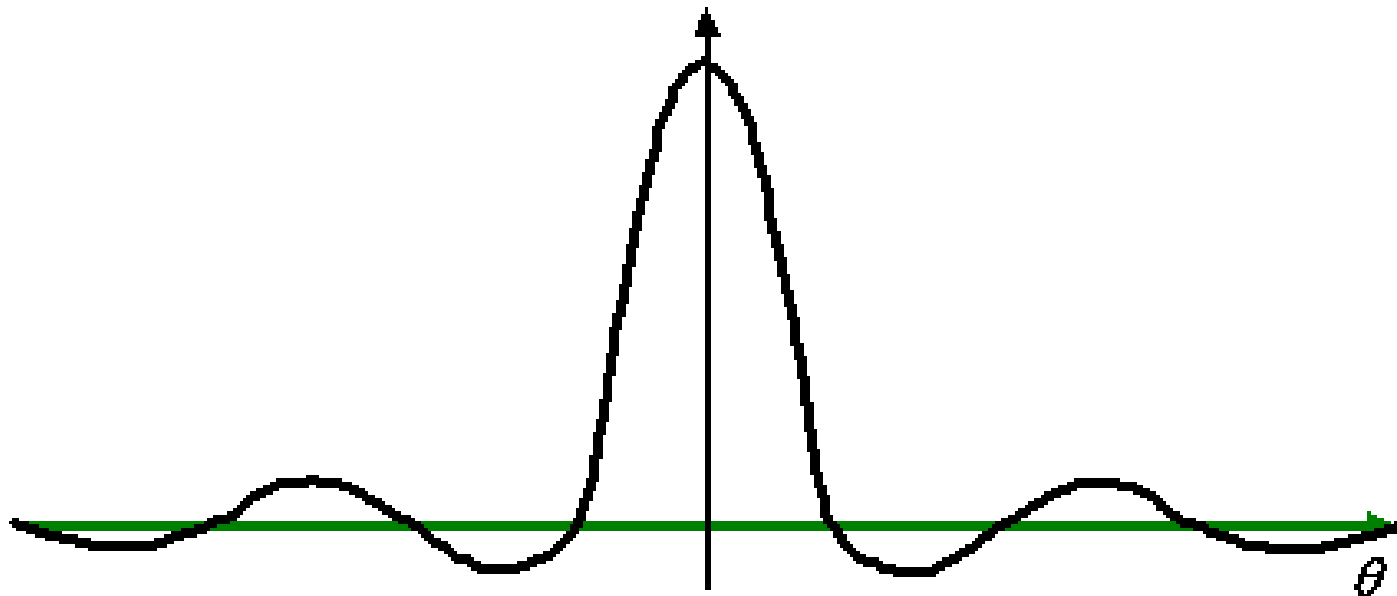


Discrete Fourier Transform

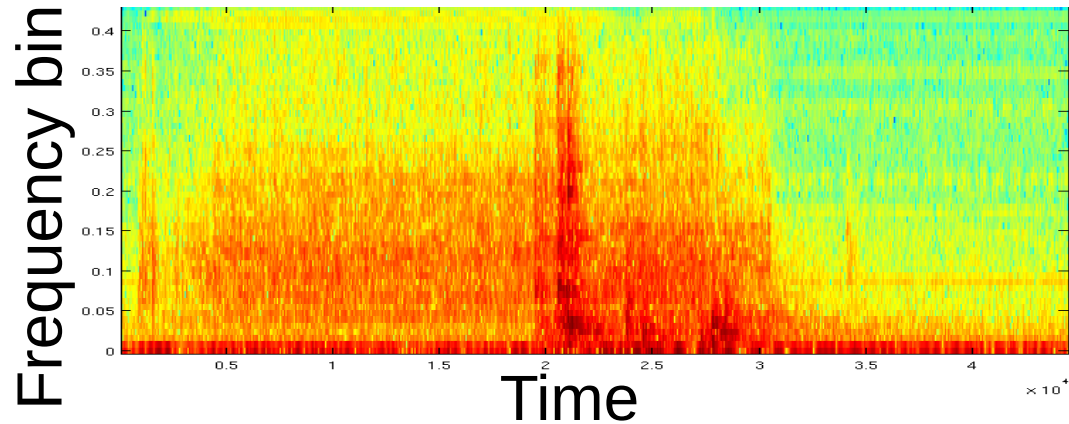
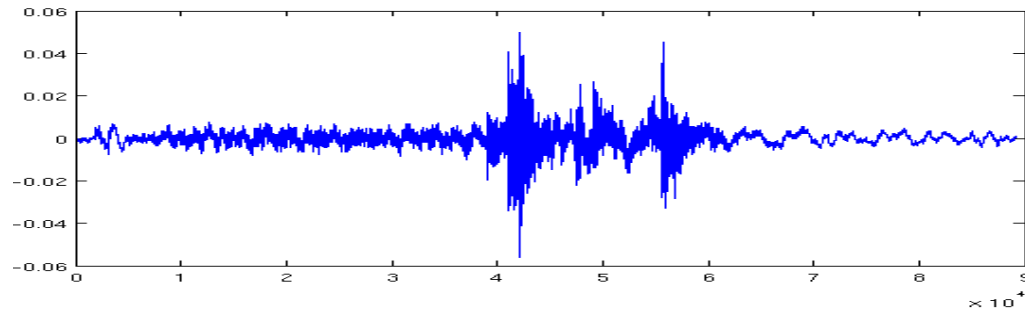


Discrete Fourier Transform

*Wave amplitude in Fraunhofer plane (the
Fourier transform of a top-hat function)*



Discrete Fourier Transform



Research Question

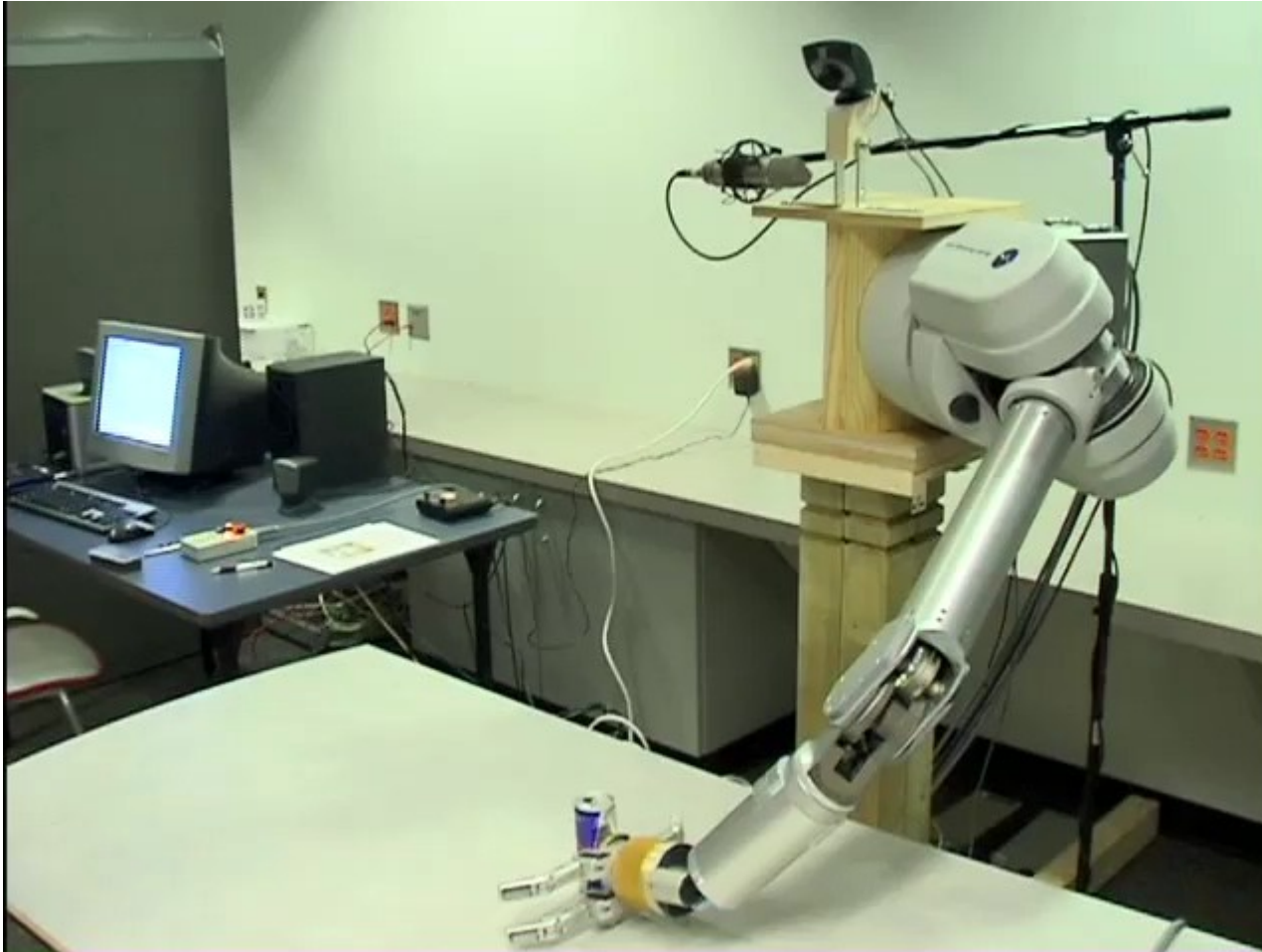
Can the DFT be used by a robot to perceive objects and their properties using sound?

Research Question

Can the DFT be used by a robot to perceive objects and their properties using sound?

How should the robot associate a particular sound with an object?

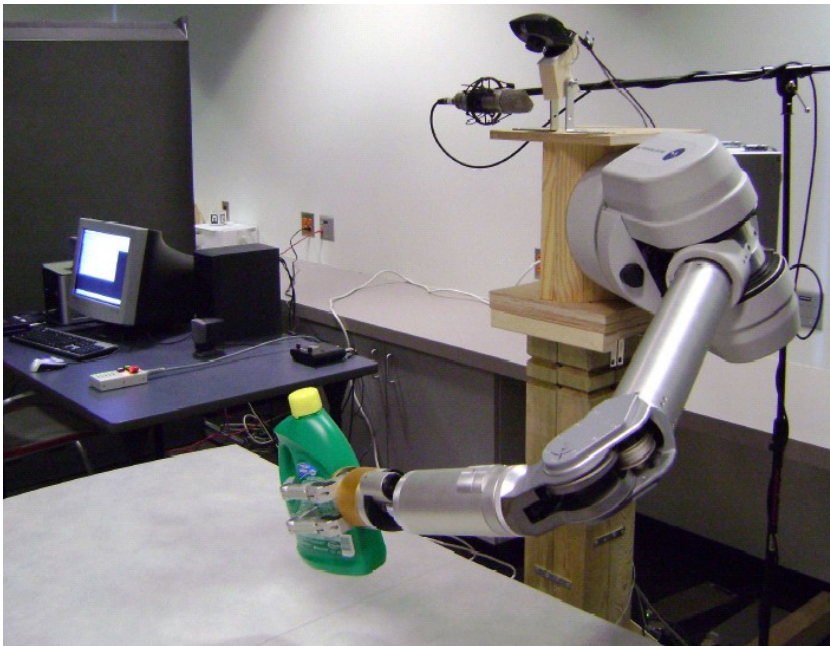
Object Exploration by a Robot



Object Exploration by a Robot



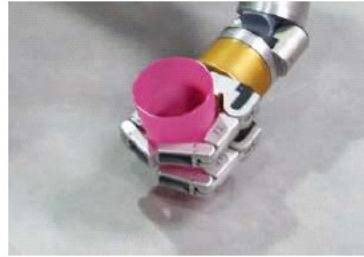
Objects



[Sinapov, Weimer, and Stoytchev, ICRA 2009]

Behaviors

Grasp:



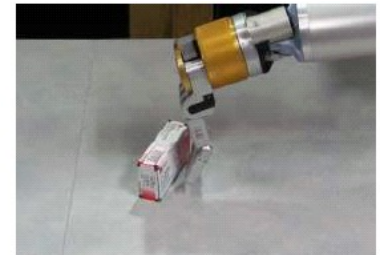
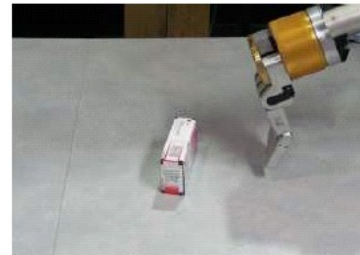
Push:



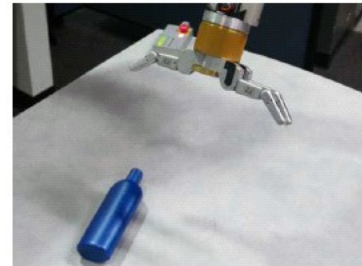
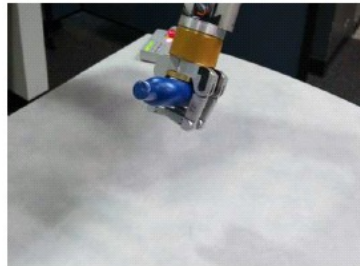
Shake:



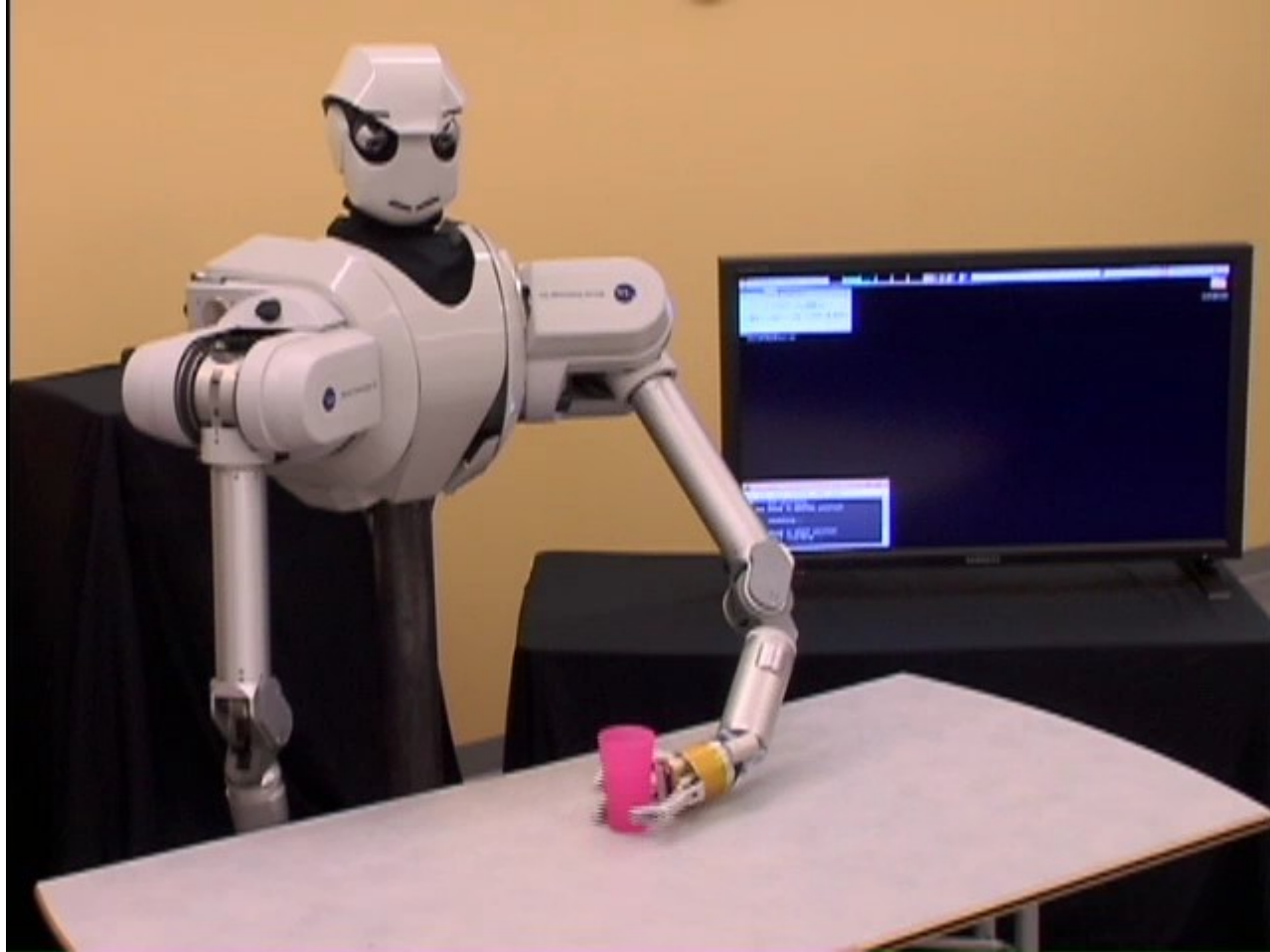
Tap:



Drop:

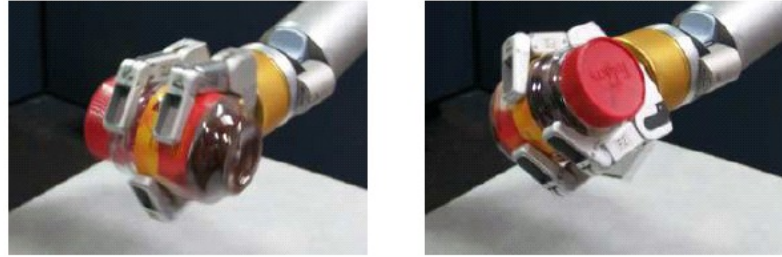


Recognition Video

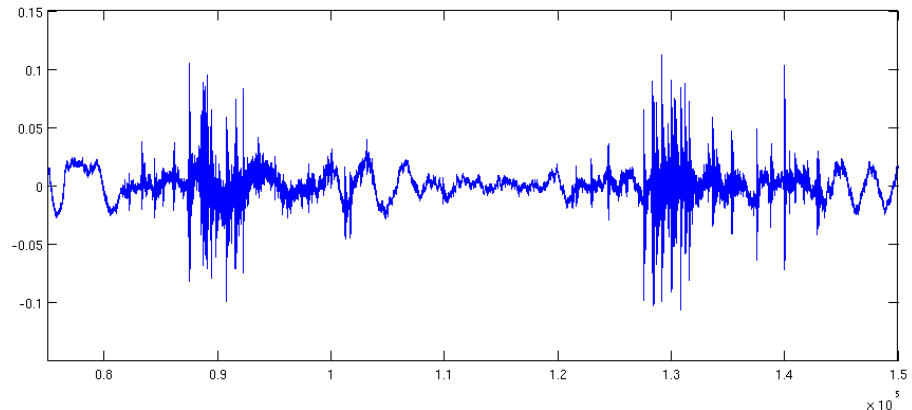


Audio Feature Extraction

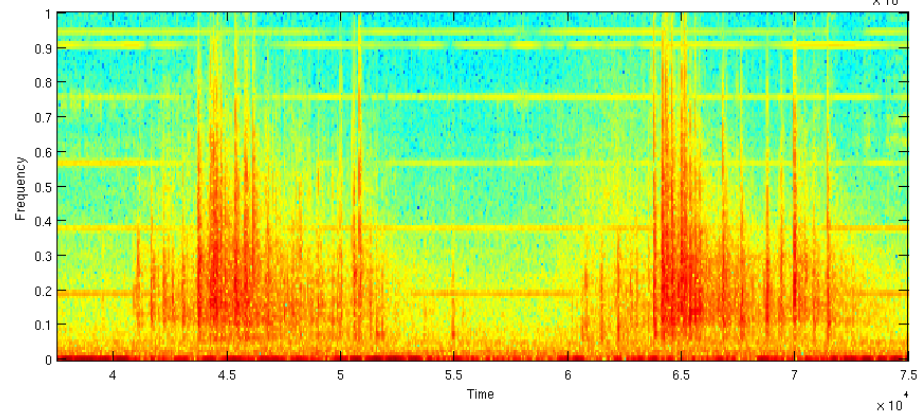
Behavior Execution:



WAV file recorded:

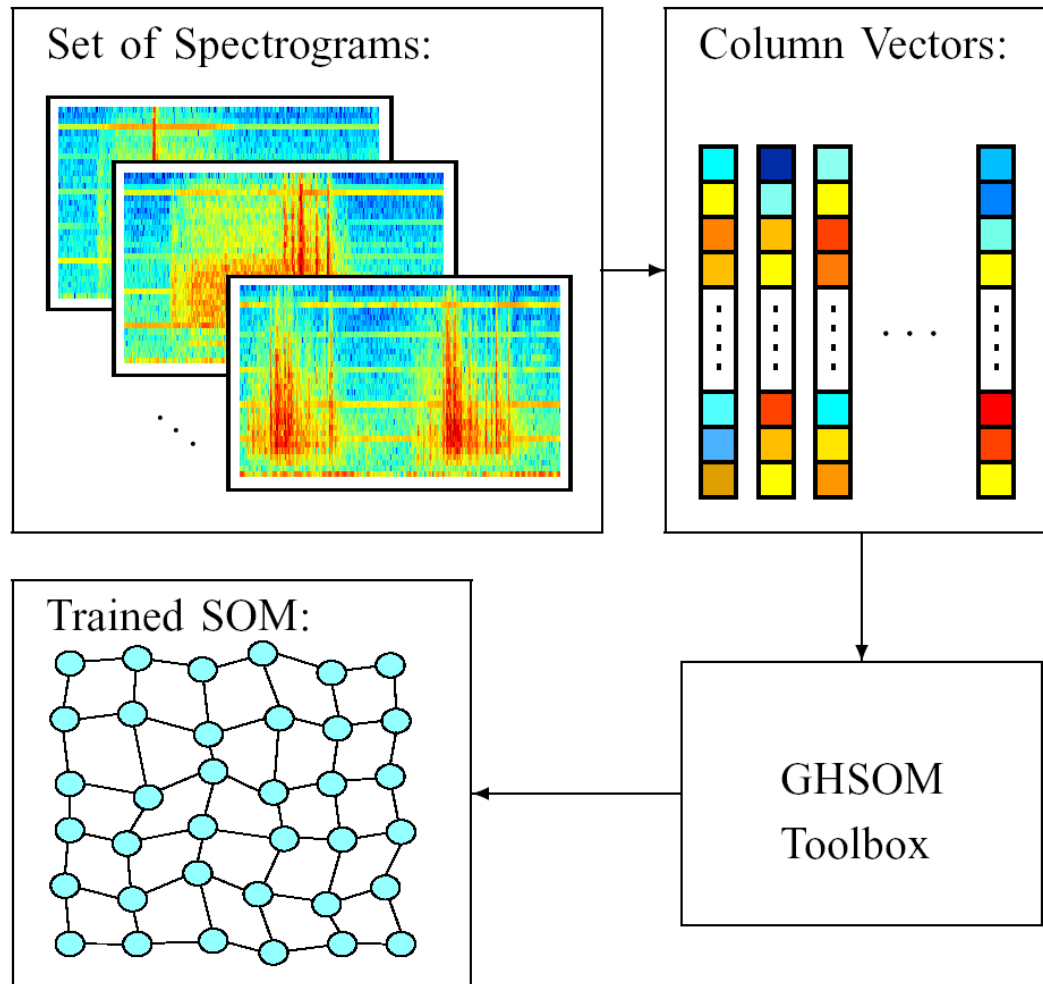


Discrete Fourier Transform:



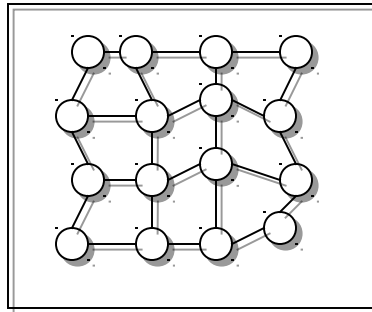
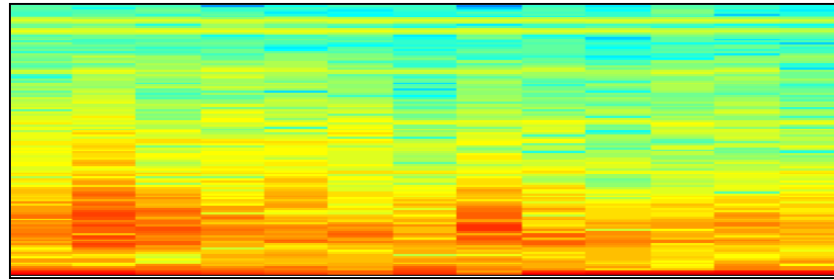
Audio Feature Extraction

1. Training a self-organizing map (SOM) using DFT column vectors:



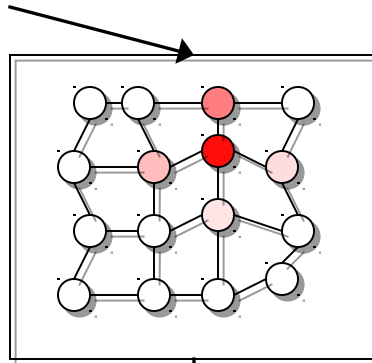
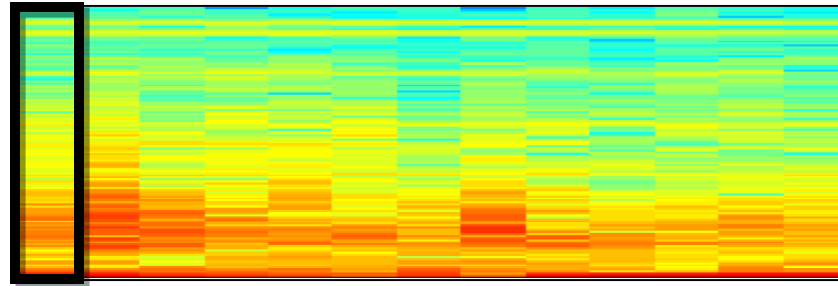
Audio Feature Extraction

2. Use SOM to convert DFT spectrogram to a sequence:



Audio Feature Extraction

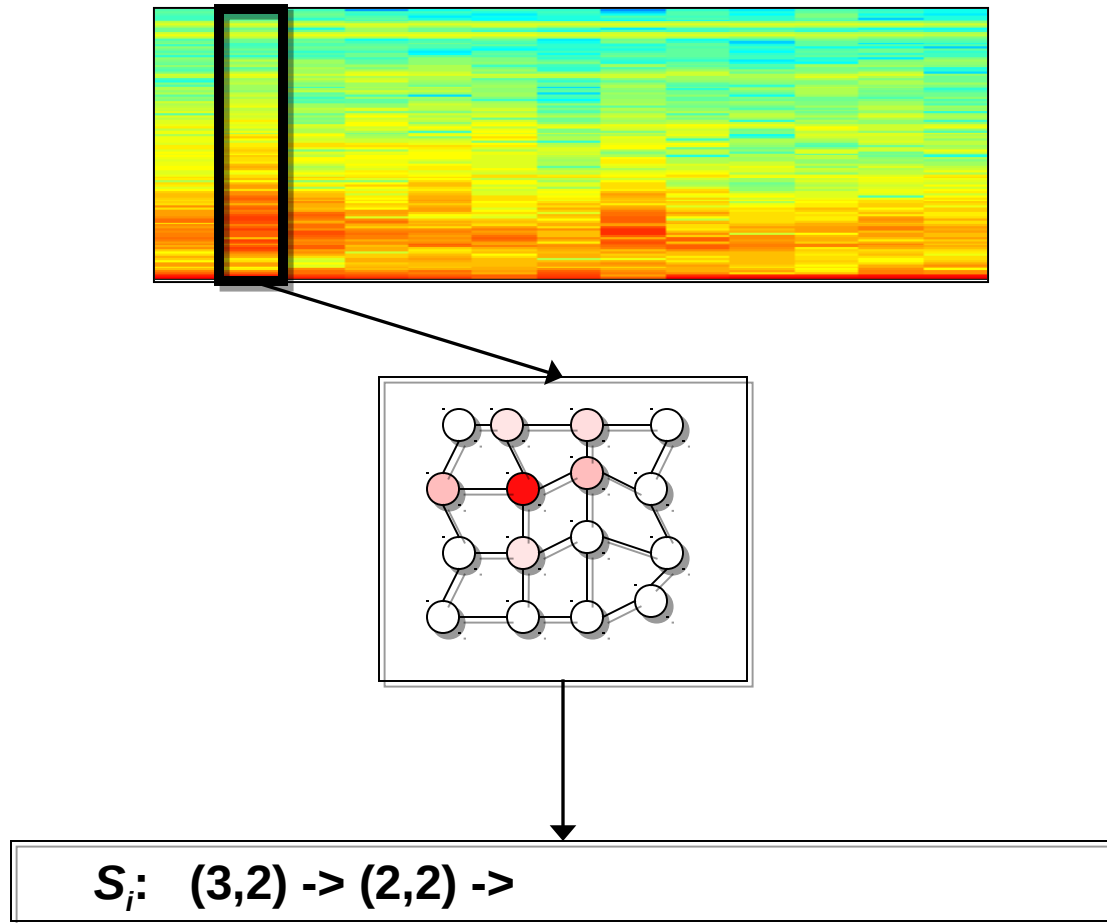
2. Use SOM to convert DFT spectrogram to a sequence:



$S_i: (3,2) \rightarrow$

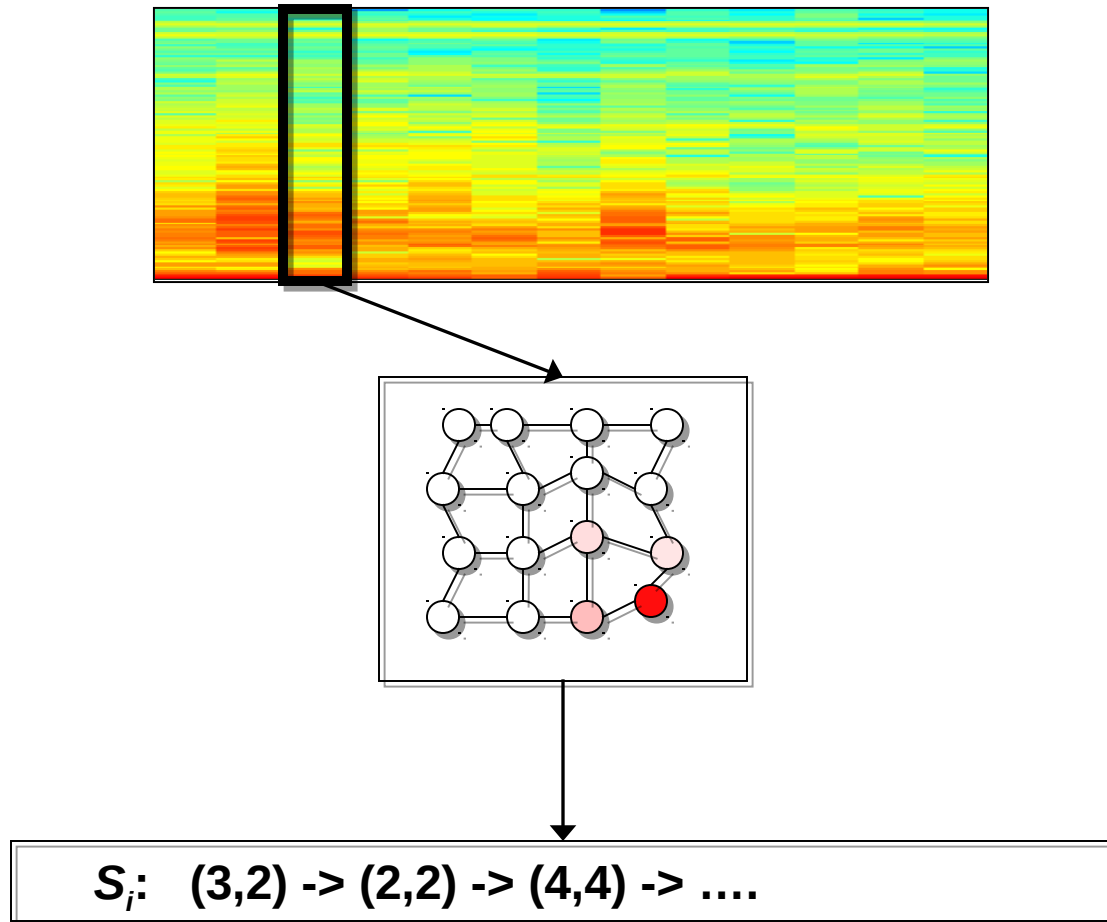
Audio Feature Extraction

2. Use SOM to convert DFT spectrogram to a sequence:



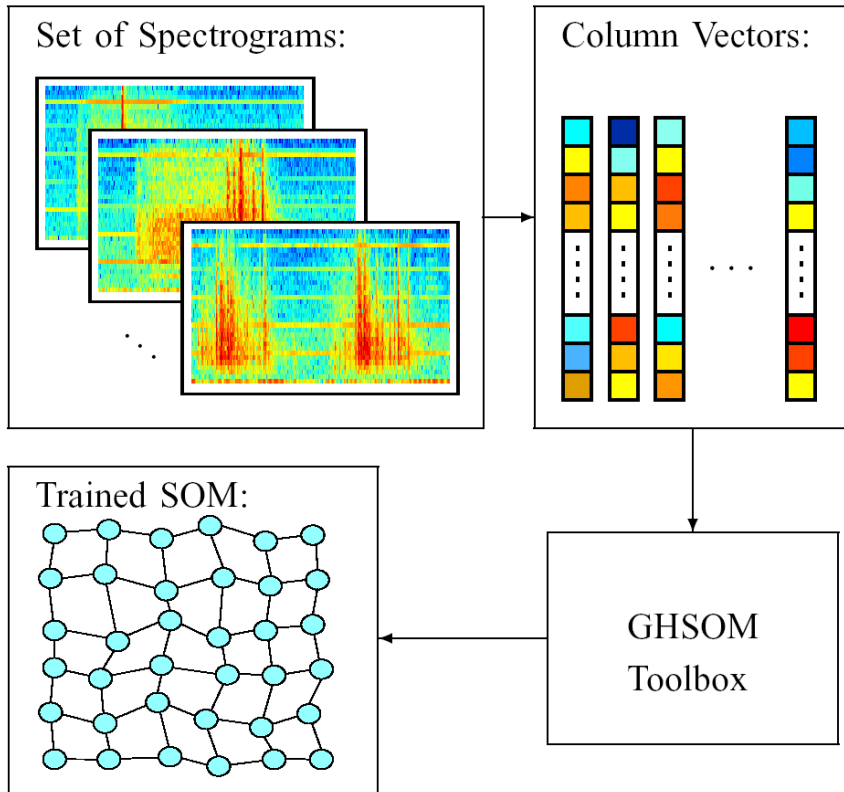
Audio Feature Extraction

2. Use SOM to convert DFT spectrogram to a sequence:

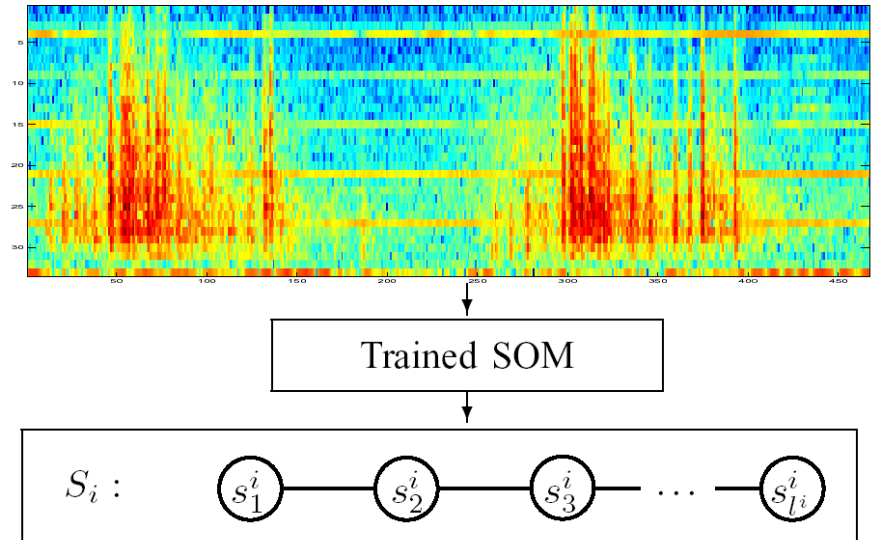


Audio Feature Extraction

1. Training a self-organizing map (SOM) using column vectors:

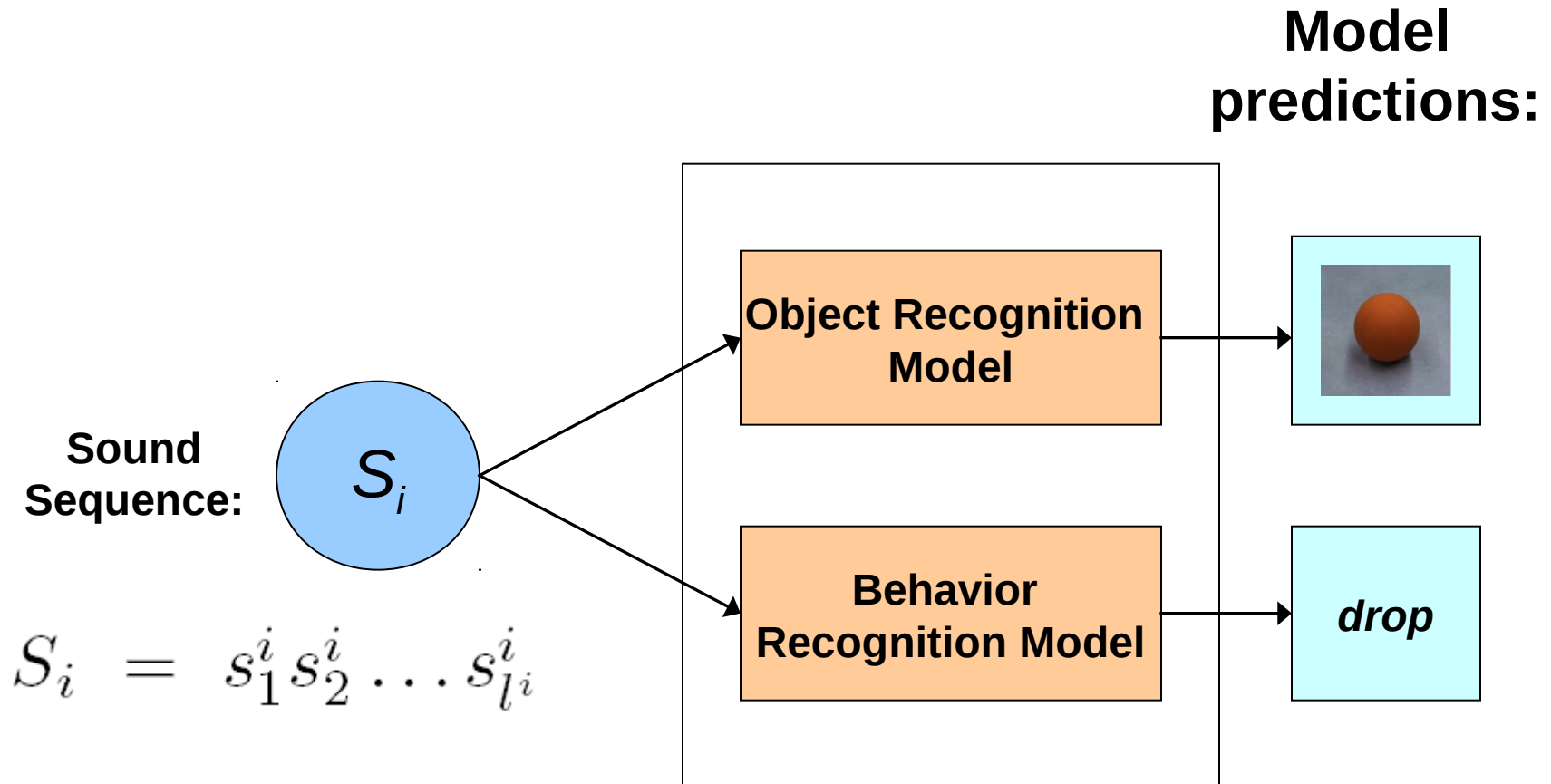


2. Discretization of a DFT of a sound using a trained SOM



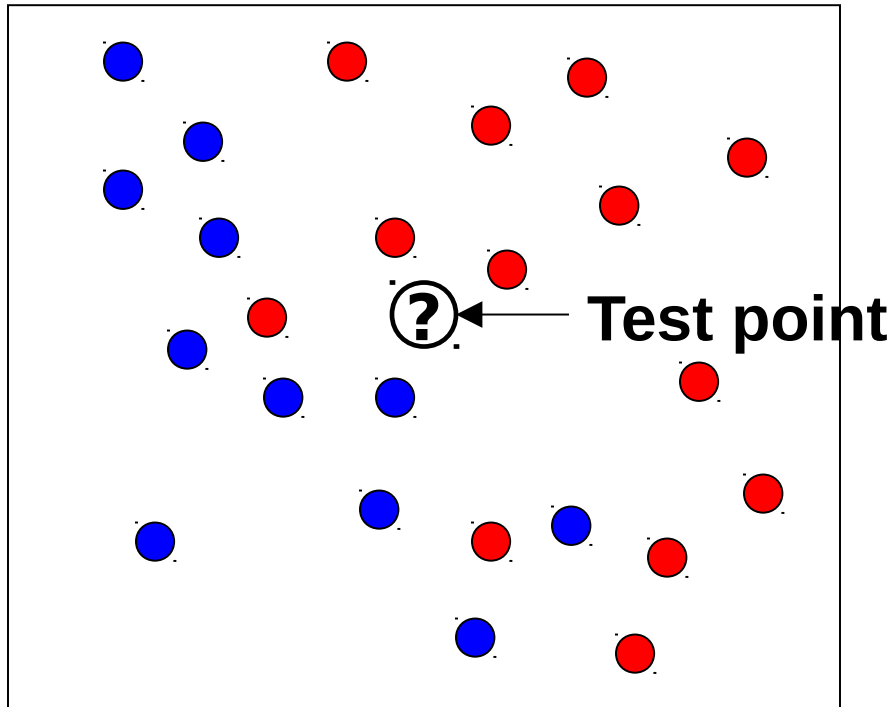
$S_i = s_1^i s_2^i \dots s_{l_i}^i$ is the sequence of activated SOM nodes over the duration of the sound

Problem Formulation



Recognition Model

- k-NN: *memory-based* learning algorithm



With $k = 3$:

2  neighbors

1  neighbors

Therefore,

$\text{Pr}(\text{red}) = 0.66$

$\text{Pr}(\text{blue}) = 0.33$

Off-Line Evaluation

- 10 trials performed with each of the 36 objects with each of the 5 behaviors
- A total of 1800 interactions, about 12 hours
- 10 fold cross-validation
- Performance Measure for object and behavior recognition:t

$$\% \text{ Accuracy} = \frac{\# \text{ correct predictions}}{\# \text{ total predictions}} \times 100$$

Evaluation Results

Behavior	k-Nearest Neighbor	Support Vector Machine
Grasp	67.89 %	75.27 %
Shake	49.47 %	50.56 %
Drop	85.79 %	80.56 %
Push	82.89 %	84.44 %
Tap	78.15 %	75.84 %
Average	72.84 %	73.33 %

Chance accuracy = 2.7 %

Evaluation Results

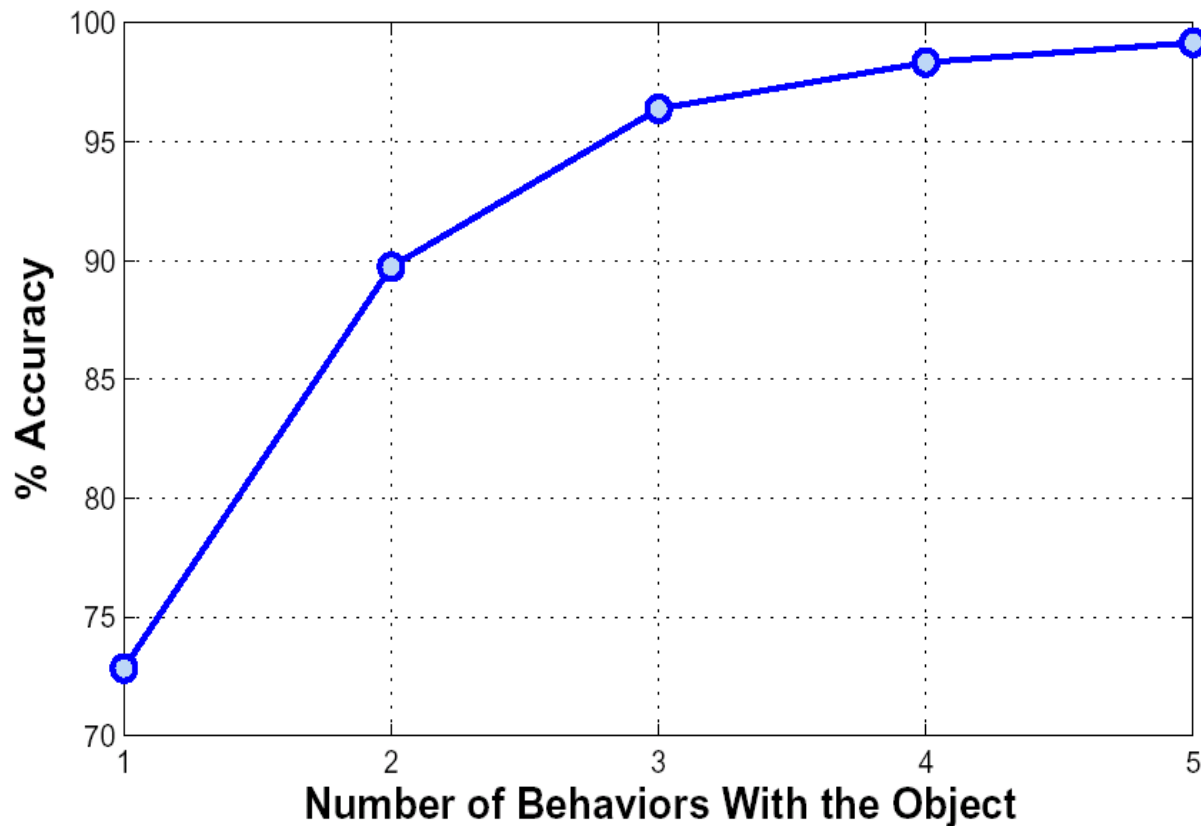
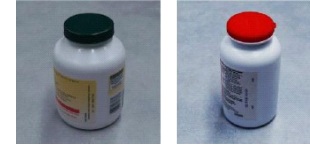
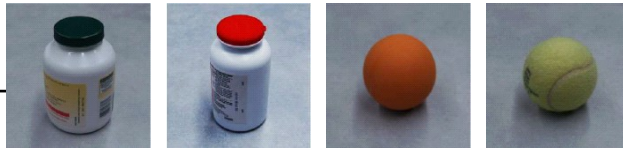


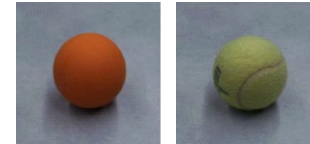
Fig. 6. Object recognition performance with k-Nearest Neighbor as the number of interactions with the object is varied from 1 (the default, used to generate Table I) to 5 (applying all five behaviors on the object).

Estimating Acoustic Object Similarity using Confusion Matrix

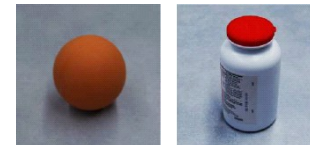
Predicted



: similar



: similar

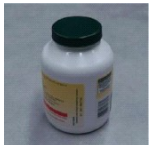


: different

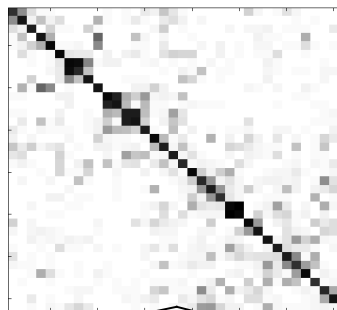


: different

Actual

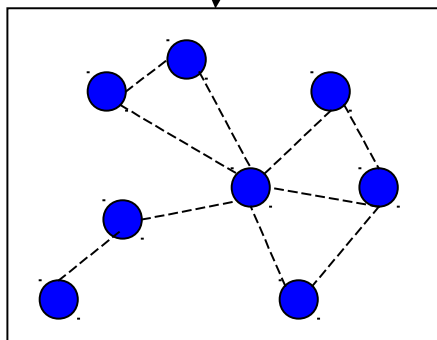
	40	4	0	0
	6	42	0	0
	0	0	21	6
	0	0	8	35

**Full Confusion Matrix
for all 36 objects:**

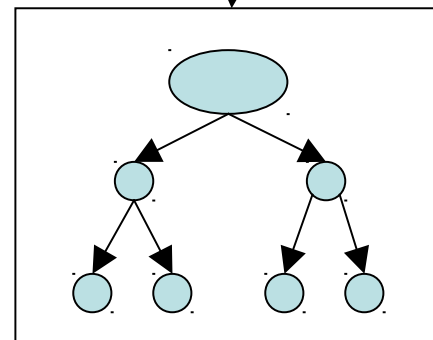


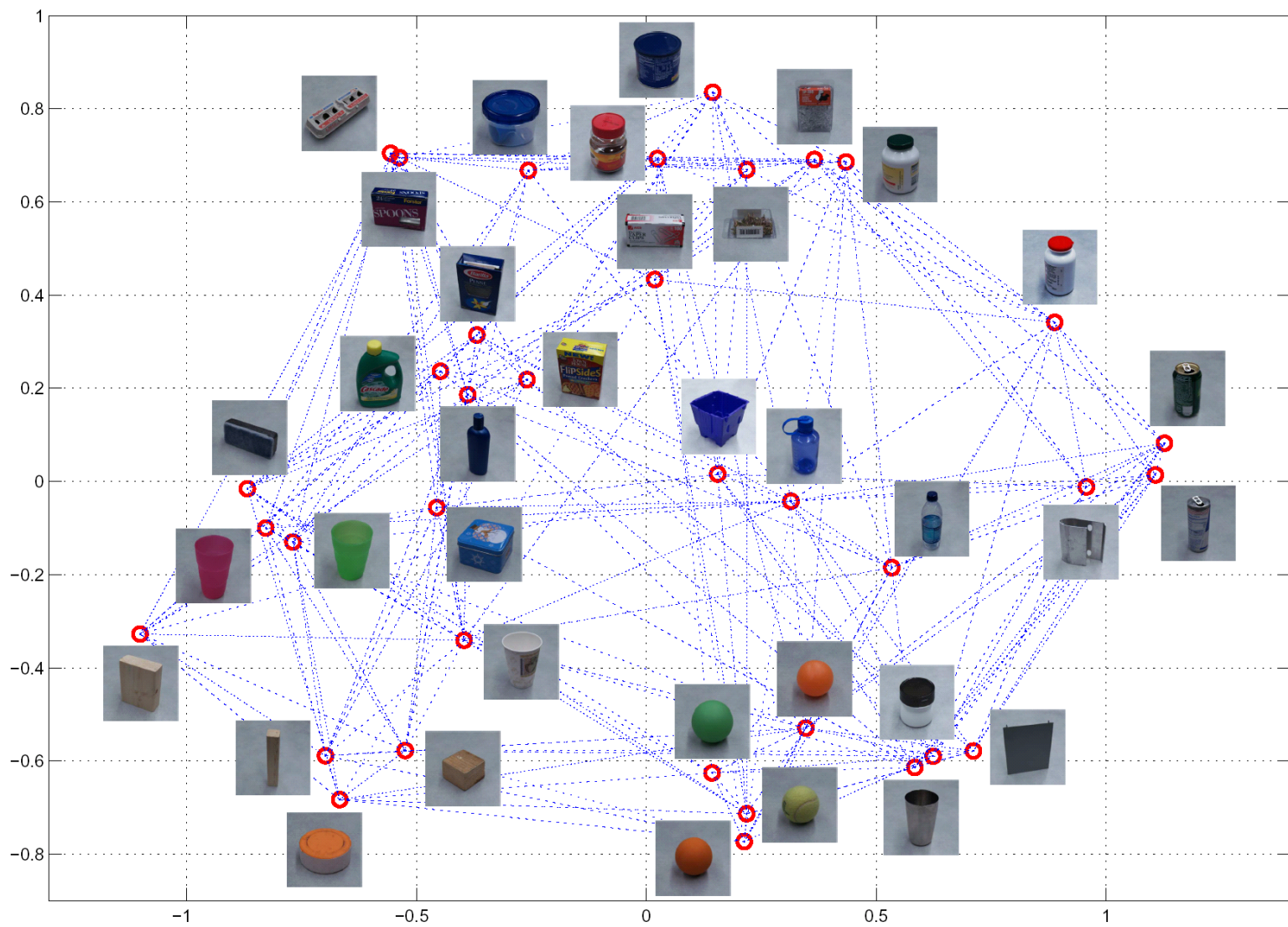
invert

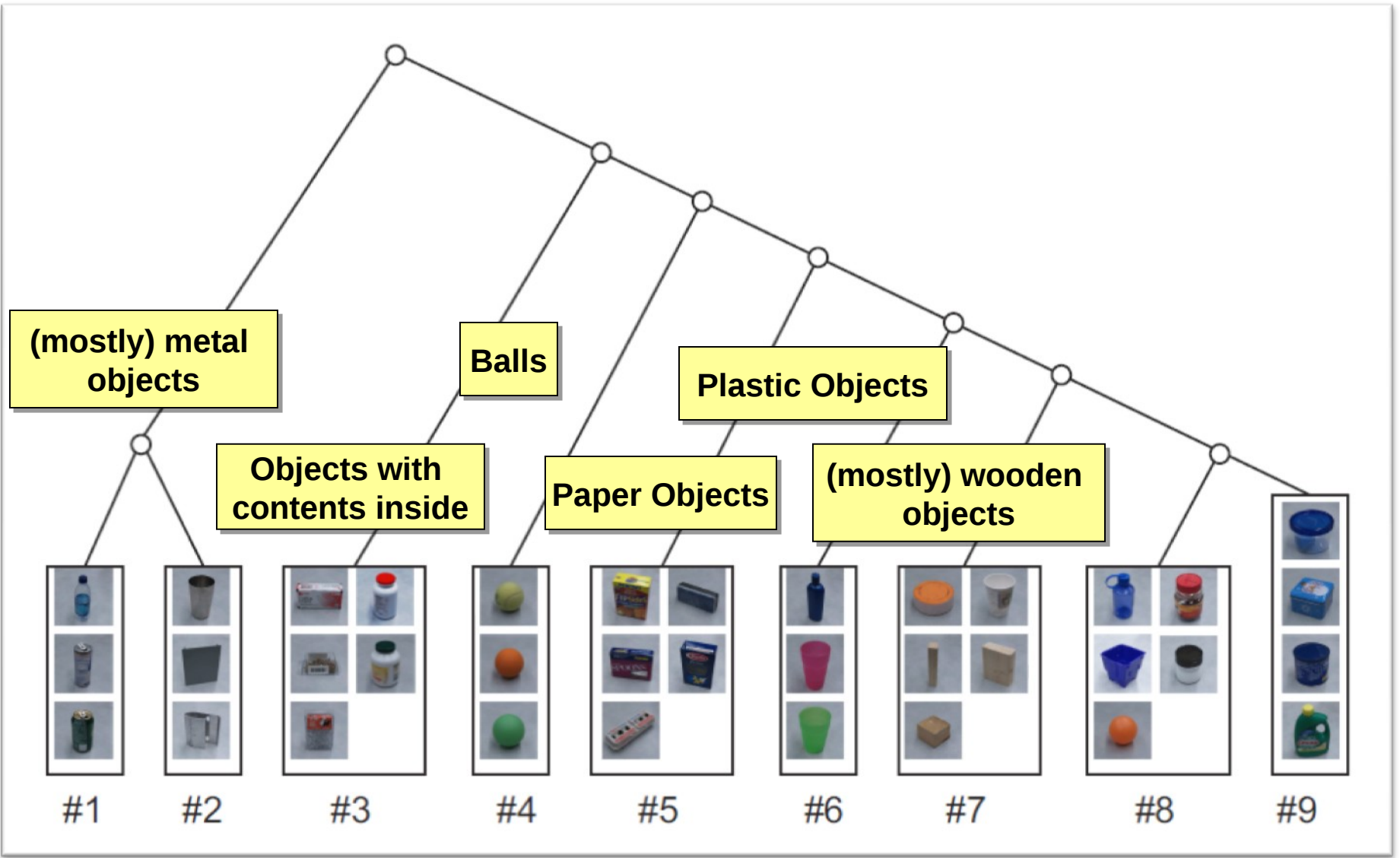
ISOMAP



Hierarchical Clustering

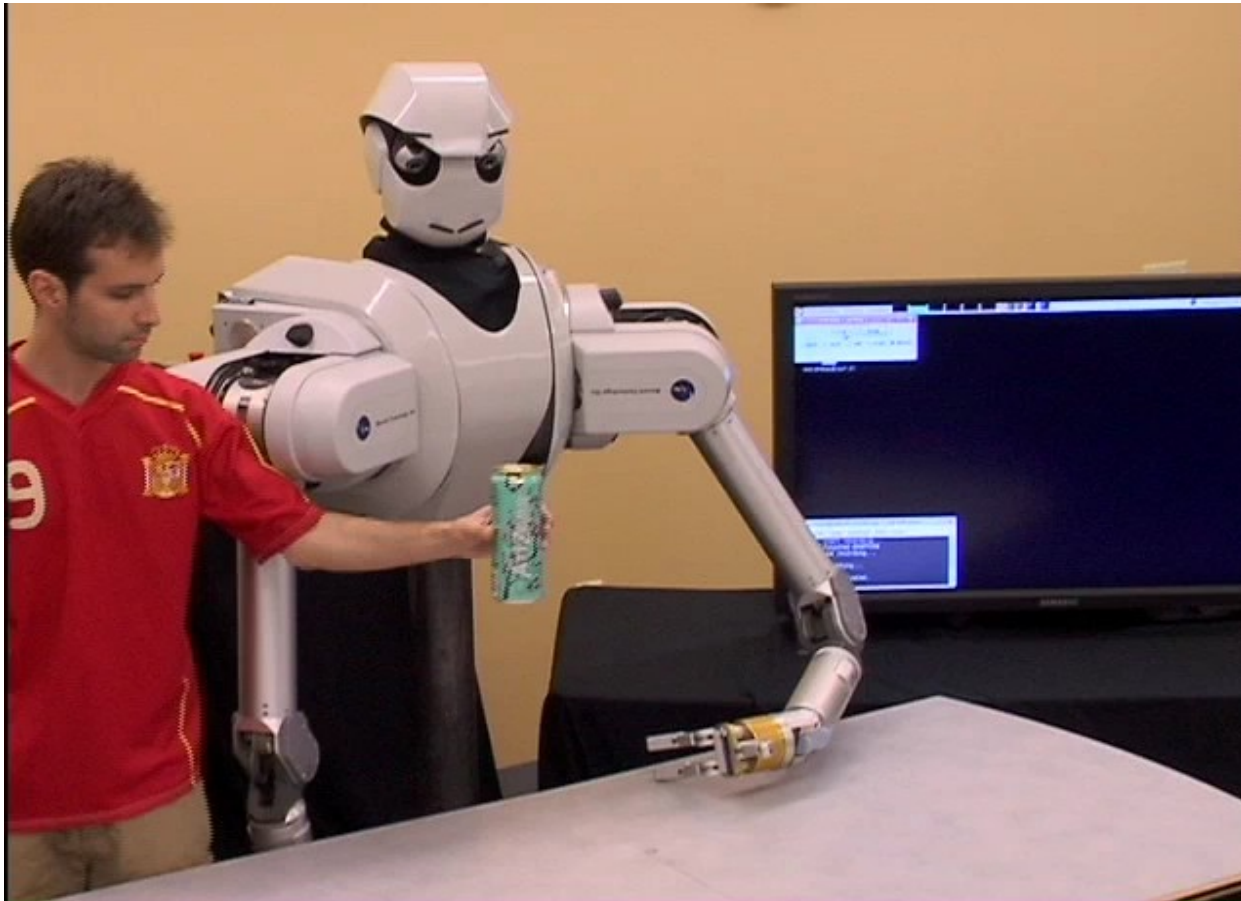






Recognizing the sounds of objects
manipulated by other agents

Recognizing the sounds of objects manipulated by other agents



Further Reading

- Sinapov, J., Wiemer, M., and Stoytchev, A. (2008). **Interactive Learning of the Acoustic Properties of Objects by a Robot.** *In proceedings of the "Robot Manipulation: Intelligence in Human Environments" workshop held at the Robotics Science and System Conference, 2008.*
- Sinapov, J., Wiemer, M., and Stoytchev, A. (2009). **Interactive Learning of the Acoustic Properties of Household Objects.** *In proceedings of the 2009 IEEE International Conference on Robotics and Automation (ICRA).*

Discussion

- What kind of sounds should our mobile robots pay attention to?
- What would auditory perception allow them to do that they currently cannot?

THE END