

# *E9 205 Machine Learning for Signal Processing*

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## Introduction to Machine Learning of Sensory Signals

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# Feature Extraction

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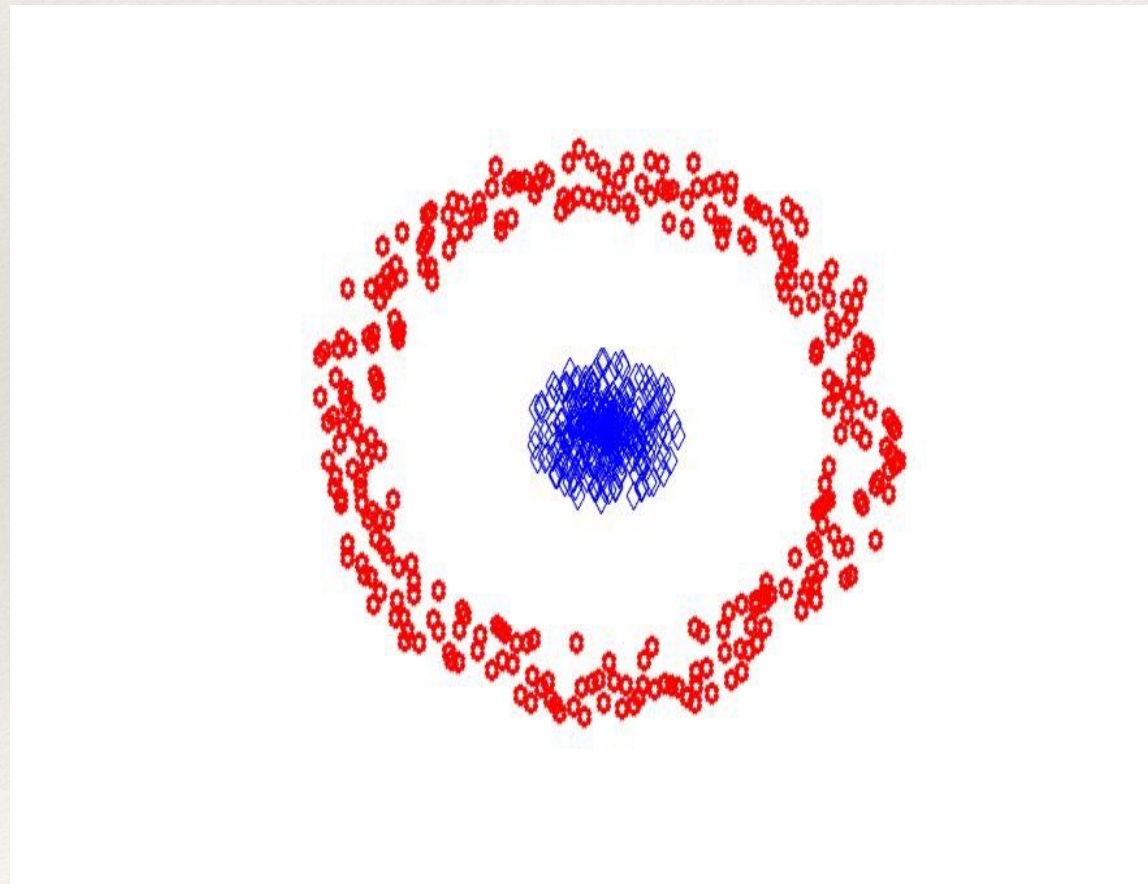
- ❖ Feature Extraction
  - ❖ Using measured data to build desirable values.
  - ❖ Attributes of the data that are informative and non-redundant.
  - ❖ Resilience to noise / artifacts.
  - ❖ Facilitating subsequent learning algorithm.



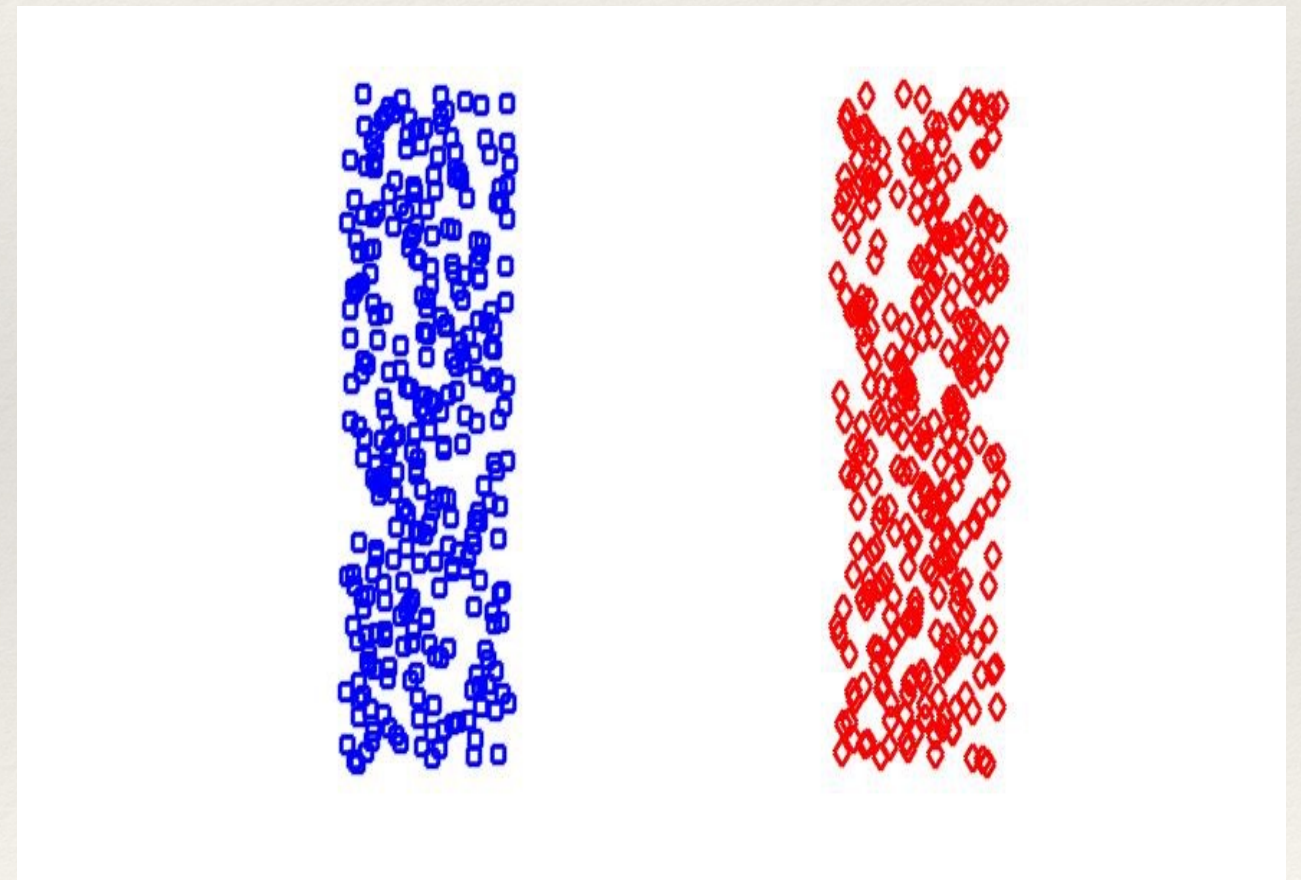
# Feature Extraction

## ❖ Representation Problem

Cartesian Coordinates



Polar Coordinates





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# Feature Extraction

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## Scope for this course

I. Feature Extraction in Text.

II. Feature Extraction in Speech and Audio signals.

III. Feature Extraction for Images.



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# Text Modeling - Introduction to NLP

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- ❖ Definitions
  - ❖ Documents, Corpora, Tokens (Terms)
- ❖ Term Frequency (TF)
- ❖ Collection Frequency (CF)
- ❖ Document Frequency (DF)
- ❖ TF-IDF
- ❖ Bag of words model



# Example [Manning and Schutze, 2006]

Word	cf	df
try	10422	8760
insurance	10440	3997

► **Figure 6.7** Collection frequency (cf) and document frequency (df) behave differently, as in this example from the Reuters collection.

term	$df_t$	$idf_t$
car	18,165	1.65
auto	6723	2.08
insurance	19,241	1.62
best	25,235	1.5

► **Figure 6.8** Example of idf values. Here we give the idf's of terms with various frequencies in the Reuters collection of 806,791 documents.



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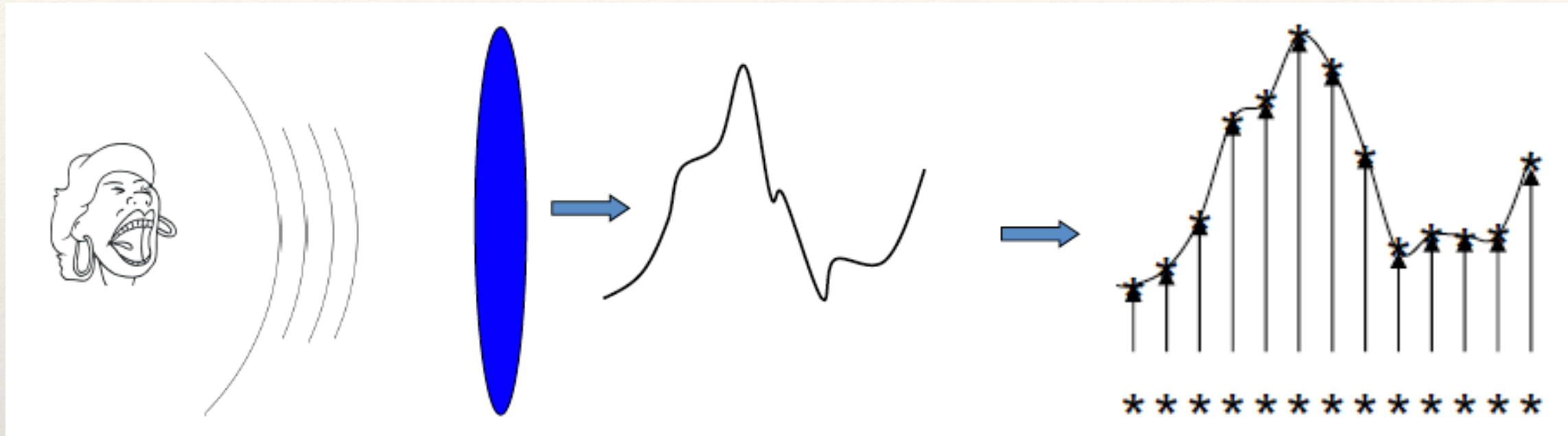
# Speech and Audio

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- ❖ Speech / Audio - 1D signals
  - ❖ Generated by pressure variations producing regions of high pressure and low pressure.
  - ❖ Travels through a medium of propagation (like air, water etc).
  - ❖ Human sensory organ - eardrum.
    - ❖ Converting pressure variations to electrical signals.
    - ❖ Action mimicked by a microphone.



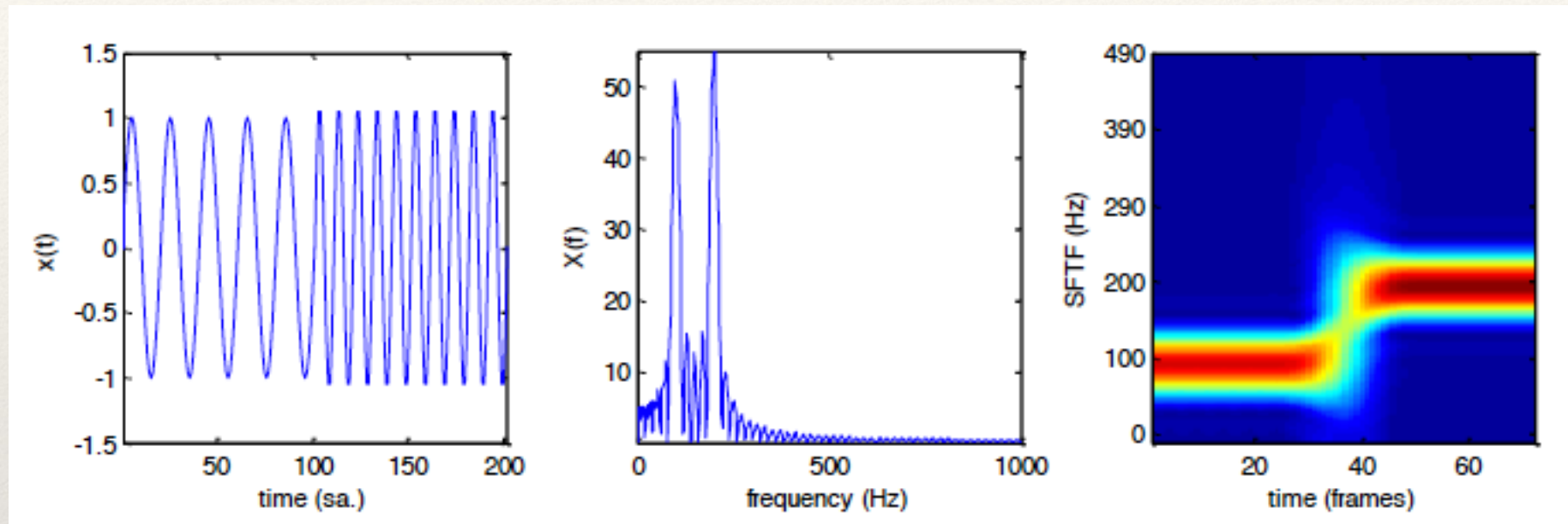
# Sound waves in a computer



- ❖ Analog continuous signal from the microphone
- ❖ Discretized in time - sampling.
- ❖ Digitized in values - quantization.



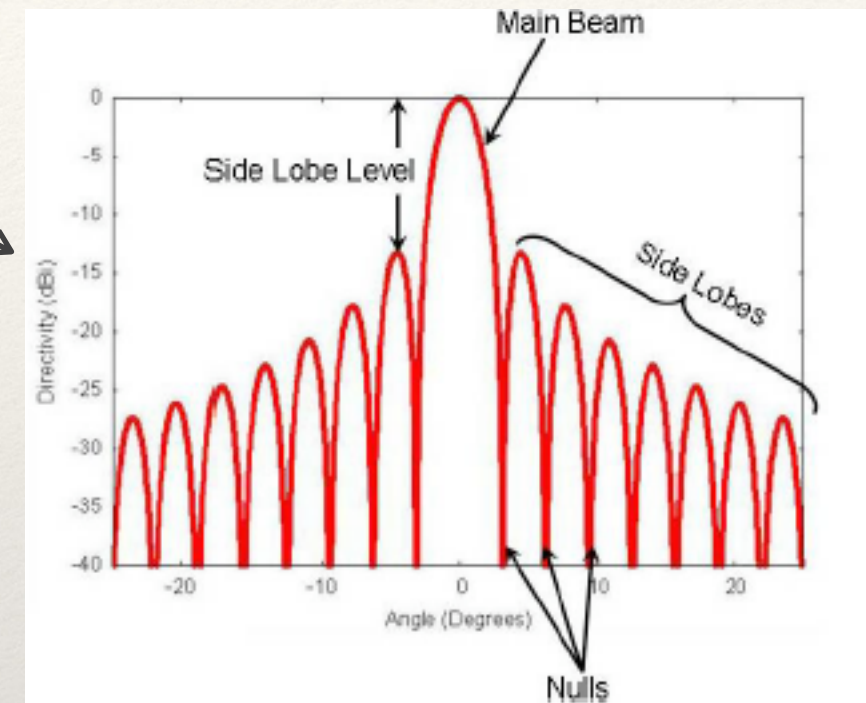
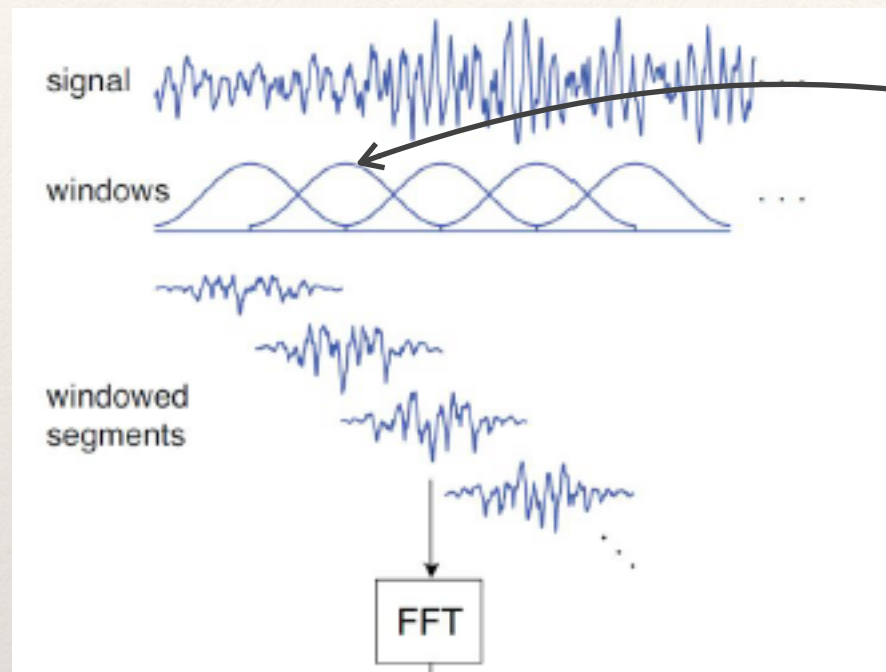
# Why do we need time varying Fourier Transform



- ❖ When the signal properties change in time
- ❖ DFT will only capture the average spectral character
- ❖ Short-window analysis can indicate the change in spectrum.

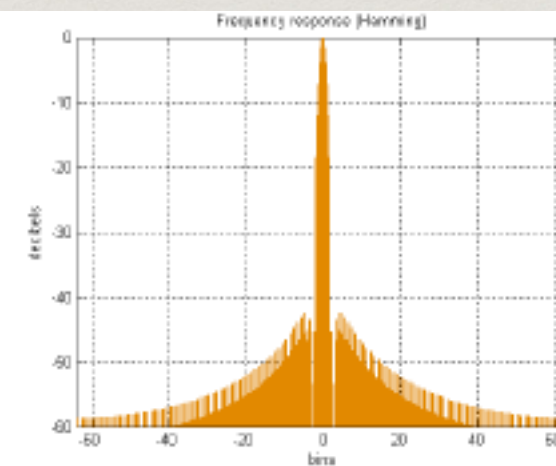
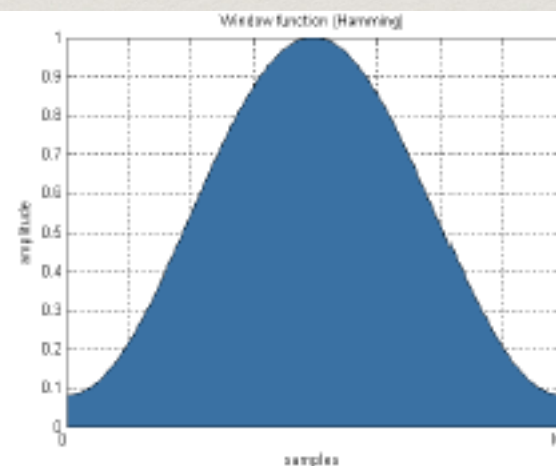


# Summary of STFT Properties



$$X[k, n_0]$$

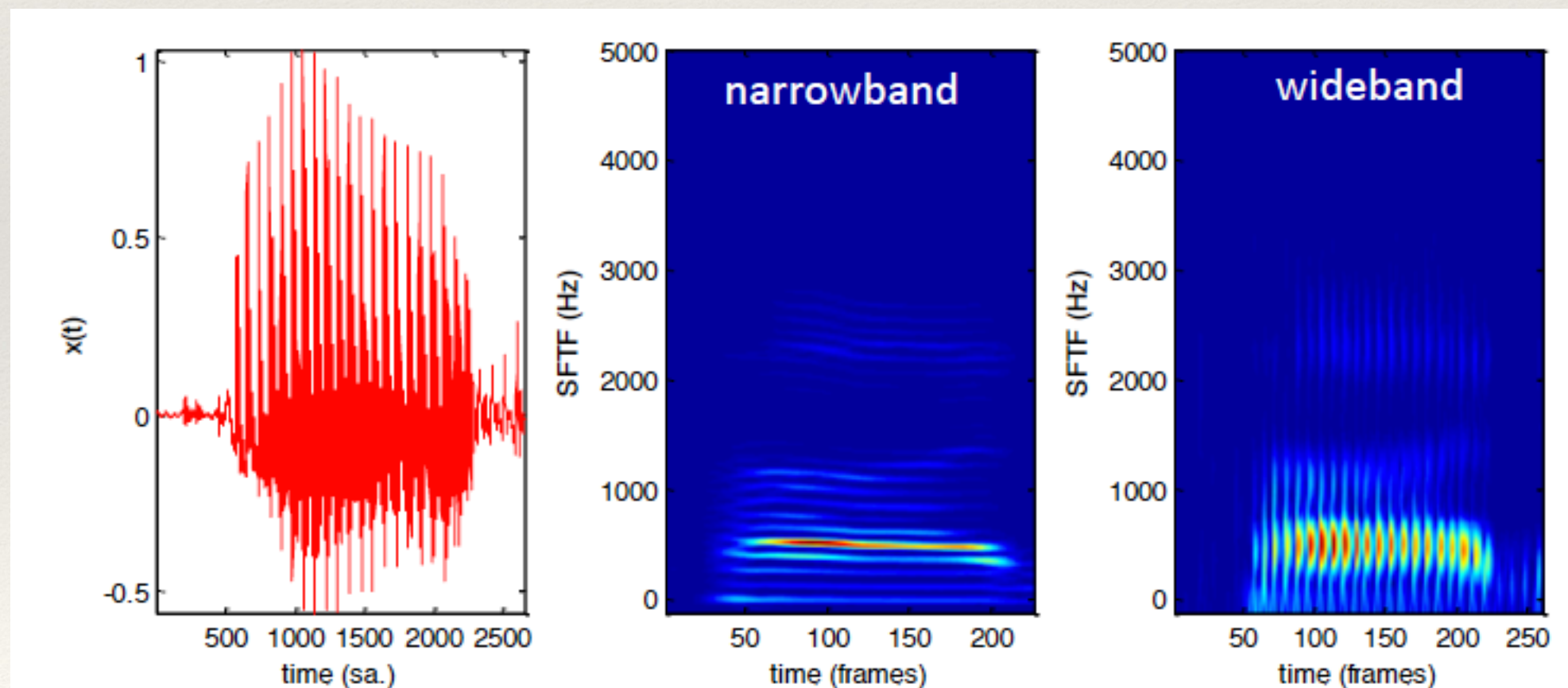
Hamming





# Narrowband versus Wideband

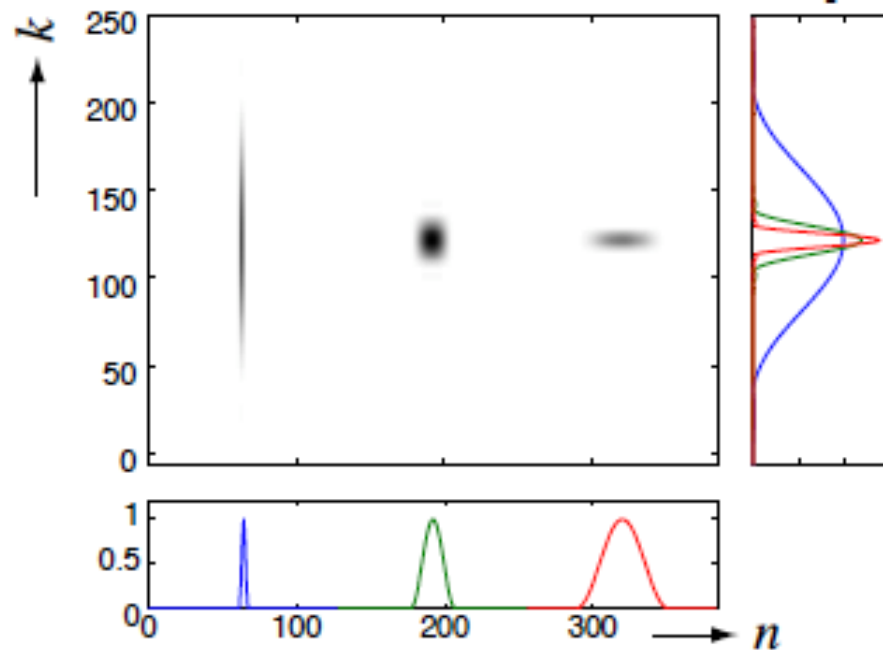
- ❖ Short windows - poor frequency resolution - wideband spectrogram
- ❖ Long windows - poor time resolution - narrowband spectrogram





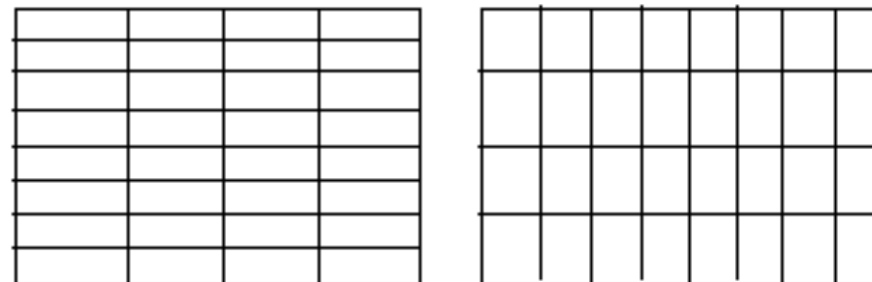
# Narrowband versus Wideband

- Can illustrate time-frequency tradeoff on the time-frequency plane:



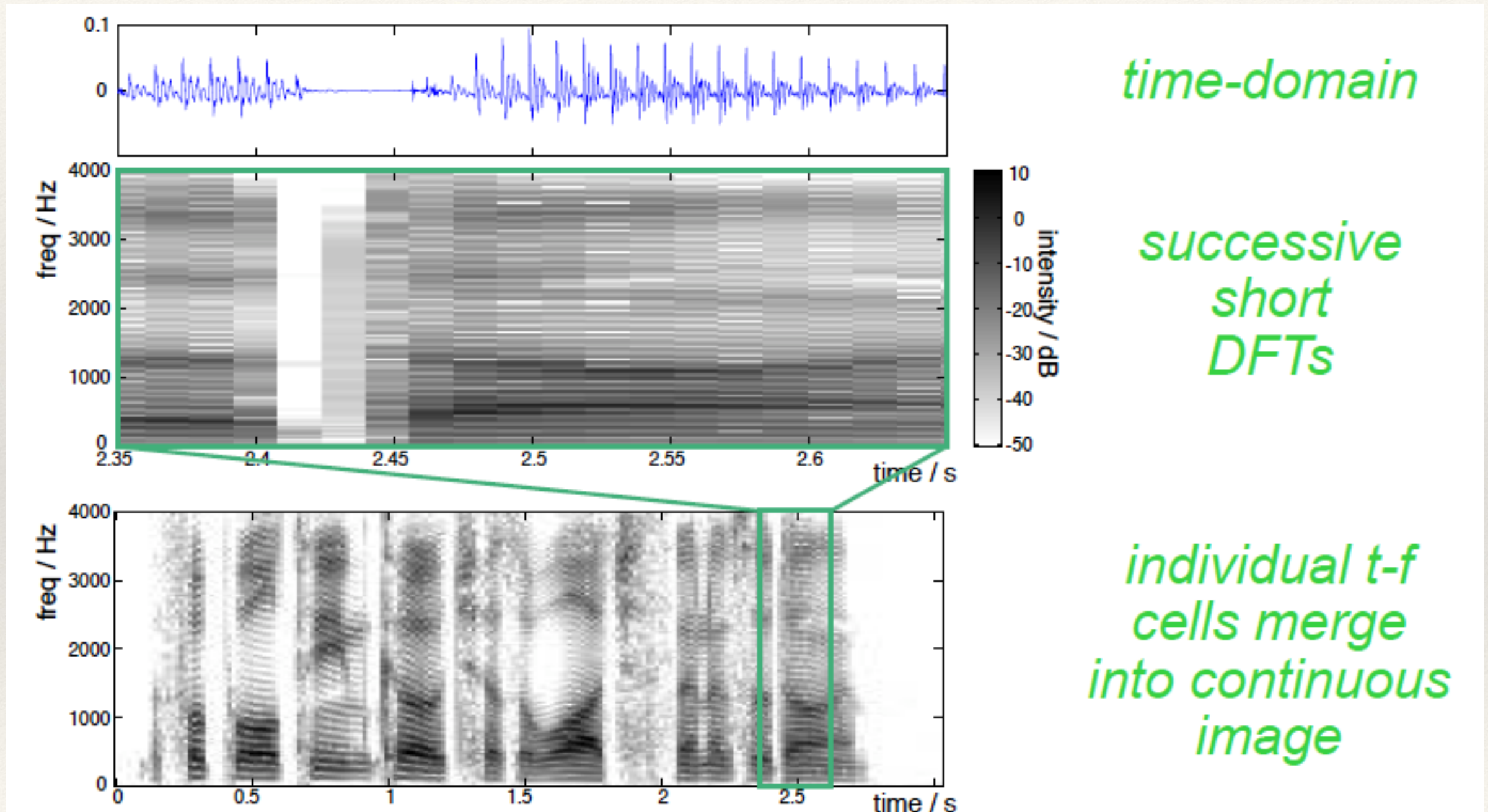
*disks show 'blurring'  
due to window length;  
area of disk is constant  
→ Uncertainty principle:  
 $\delta f \cdot \delta t \geq k$*

- Alternate tilings of time-freq:





# Spectrogram of Real Sounds





# Narrowband versus Wideband

