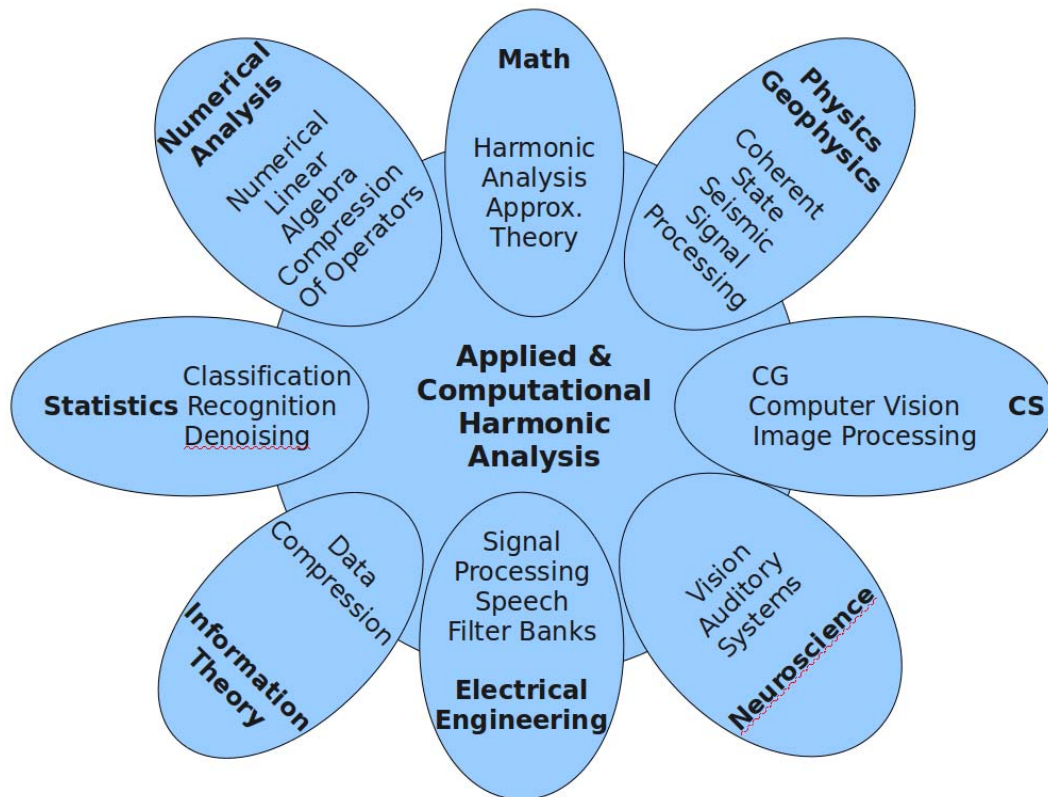


MAT 271: Applied & Computational Harmonic Analysis

Note Title

Lecture 1: Overture

Applied & Computational Harmonic Analysis (ACHA)
is an extremely **interdisciplinary** field!



This is not surprising considering the ubiquity of the celebrated **Fourier transform**!

The key role of ACHA
→ provides great tools for representing your data/functions/operators.

* Representation of data / functions using ACHA-based tools is:

(1) Efficient → good for compression / approximation

$$f(x) \approx \sum_{k=1}^n \alpha_k \psi_k(x)$$

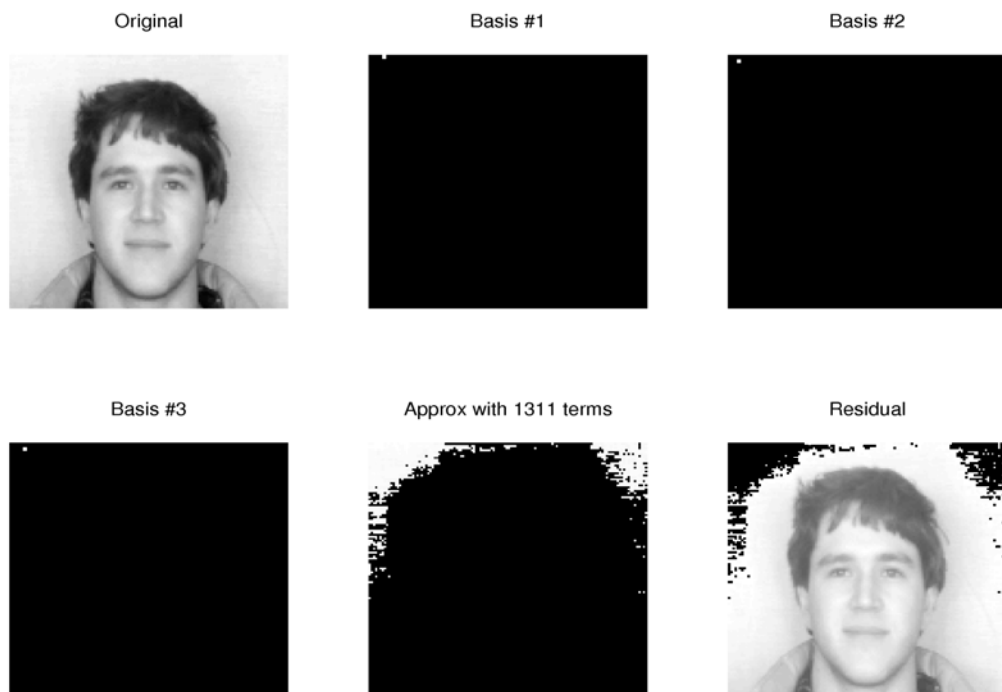
where n : small, $\{\psi_k\}$: an ONB
or more precisely,

$$\|f - \sum_{k=1}^n \alpha_k \psi_k\| = O(n^{-\beta})$$

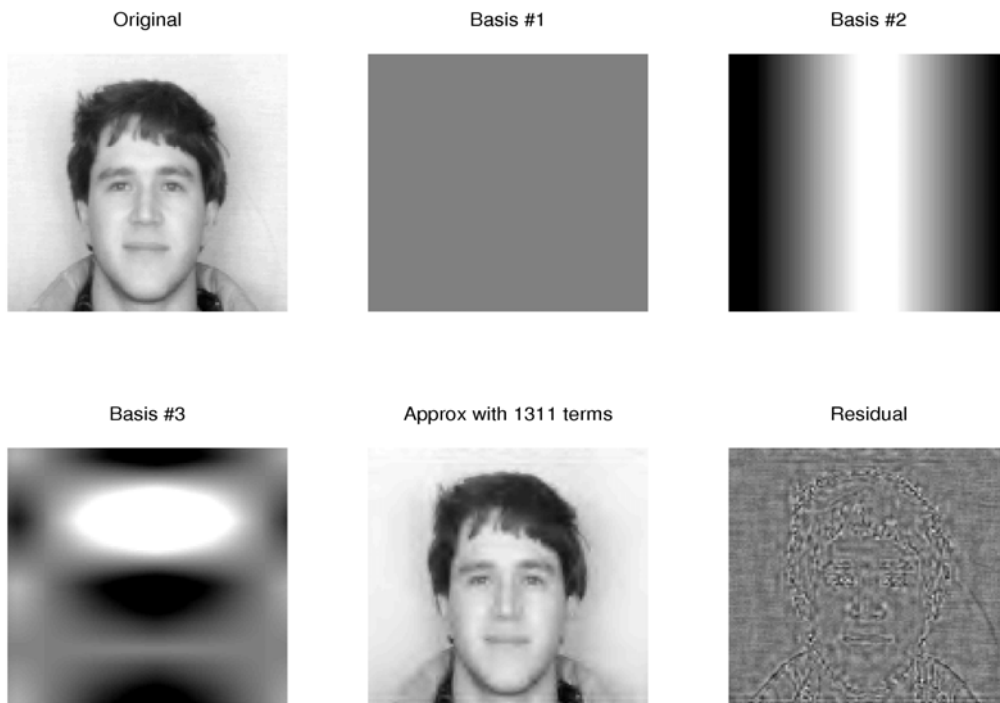
$\exists \beta > 0$

What classes of fcns possess such rate of approximation for a given set of $\{\psi_k\}$?

A bad example:

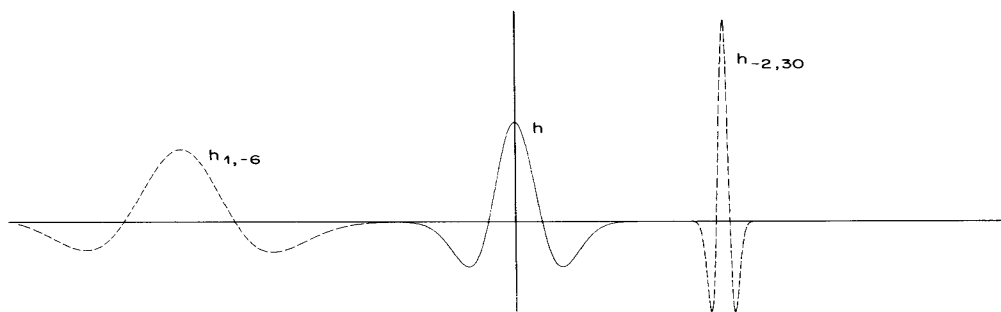


A good example:



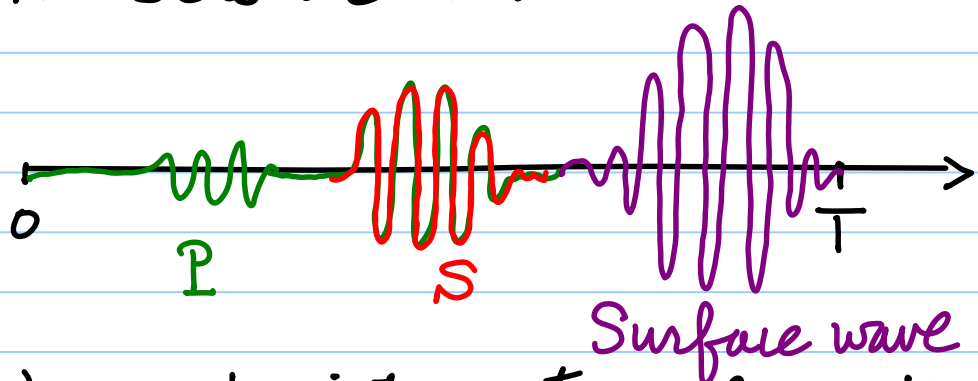
(2) Meaningful/useful

- $\{\alpha_k\}$ are used for interpretation, classification, discrimination, ...
- $\{\varphi_k\}$ are easily interpretable.

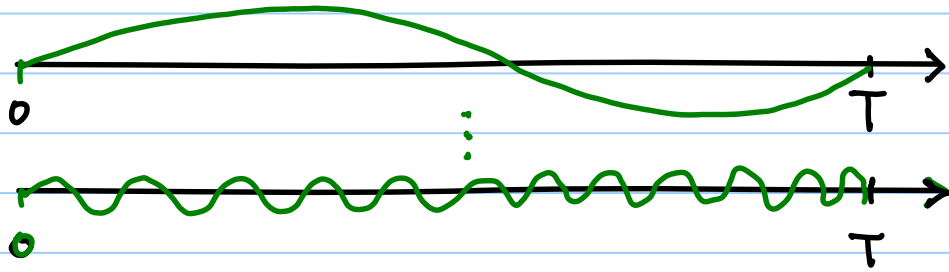


specific location, frequency, scale, duration, ...

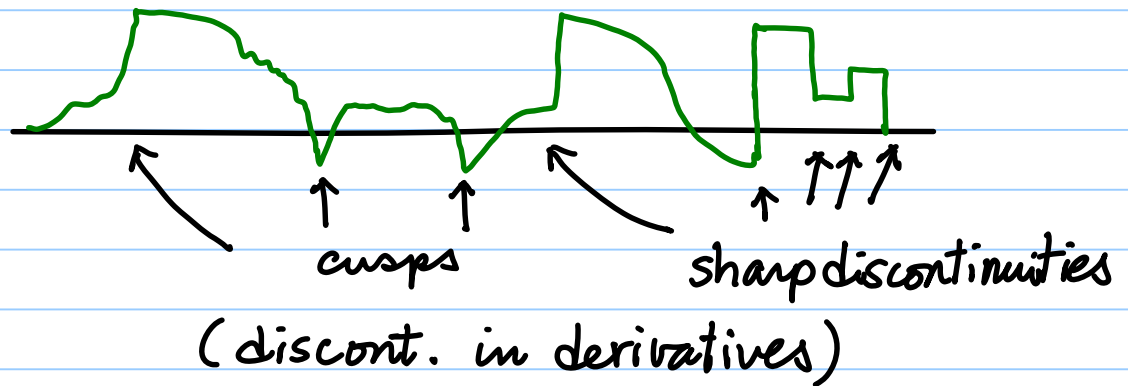
Ex. 1. Seismic waves



→ Wrong to interpret such a signal with global sinusoids



Ex. 2 images (profiles of images)



→ Wrong to use global sinusoids
→ less efficient
coeff's do not decay fast ...

(3) Representation should be **adapted** to your task!

Representation (or feature extraction) for compression may be different from that for classification
⇒ portraits of twins

(4) Computationally **fast & stable**

• If $\{\varphi_k\}$ are orthonormal, then

$$\alpha_k = \langle f, \varphi_k \rangle = \int f(x) \overline{\varphi_k(x)} dx$$

⇒ Matrix-vector multiplication in general.

$$\alpha = \Phi^* f$$

which requires $O(n^2)$ op's : slow!
Want to have $O(n \log n)$ or faster!

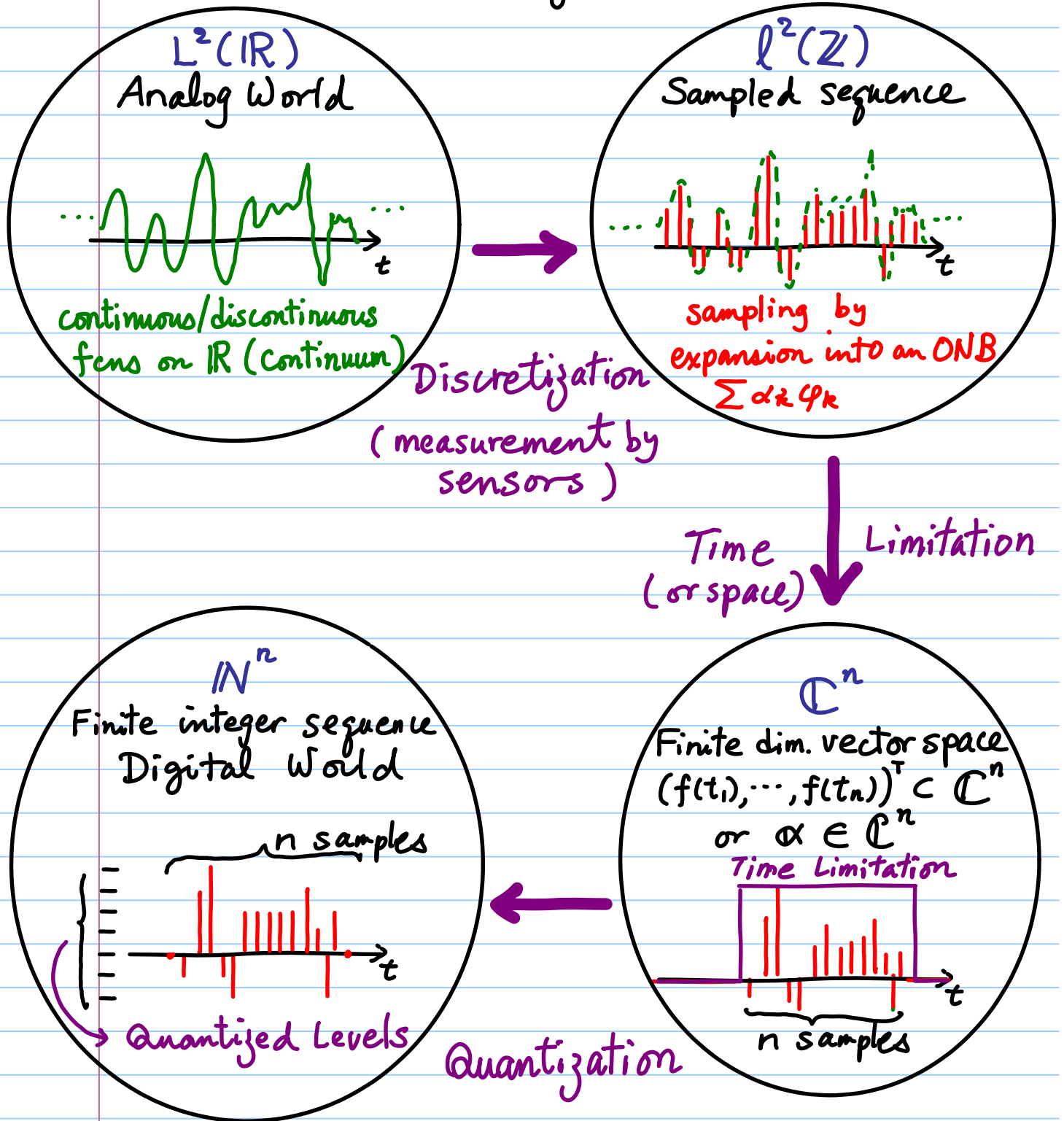
(5) Representation allows us to follow a very important strategy for any data / signal / image analysis tasks

It's a philosophy called

Ulf Grenander "Analysis by Synthesis"
Pattern Theory

We can truly "analyze" or "understand" the data / signal / images by looking at how they are composed of elementary "molecules" or "atoms".

* What is a signal?



* We won't discuss the quantization procedure in this
 * Sampling can be viewed as an expansion course.
 of a signal into a special ONB \Rightarrow The Shannon Sampling Thm

* Until we discuss Harmonic Analysis on graphs & networks, we assume that

our signals are supported on

\mathbb{R}^d , $d \geq 1$ (unbdd, continuum);

$\prod_{j=1}^d (a_j, b_j) \subset \mathbb{R}^d$ (a rectangle);

\mathbb{Z}^d , $d \geq 1$ (unbdd, discrete); or

$\prod_{j=1}^d \mathbb{Z}_{n_j}$, $n_j \in \mathbb{N}$ (regular lattice)

↑

$(0, 1, \dots, n_1-1) \times (0, 1, \dots, n_2-1) \times \dots \times (0, 1, \dots, n_d-1)$

