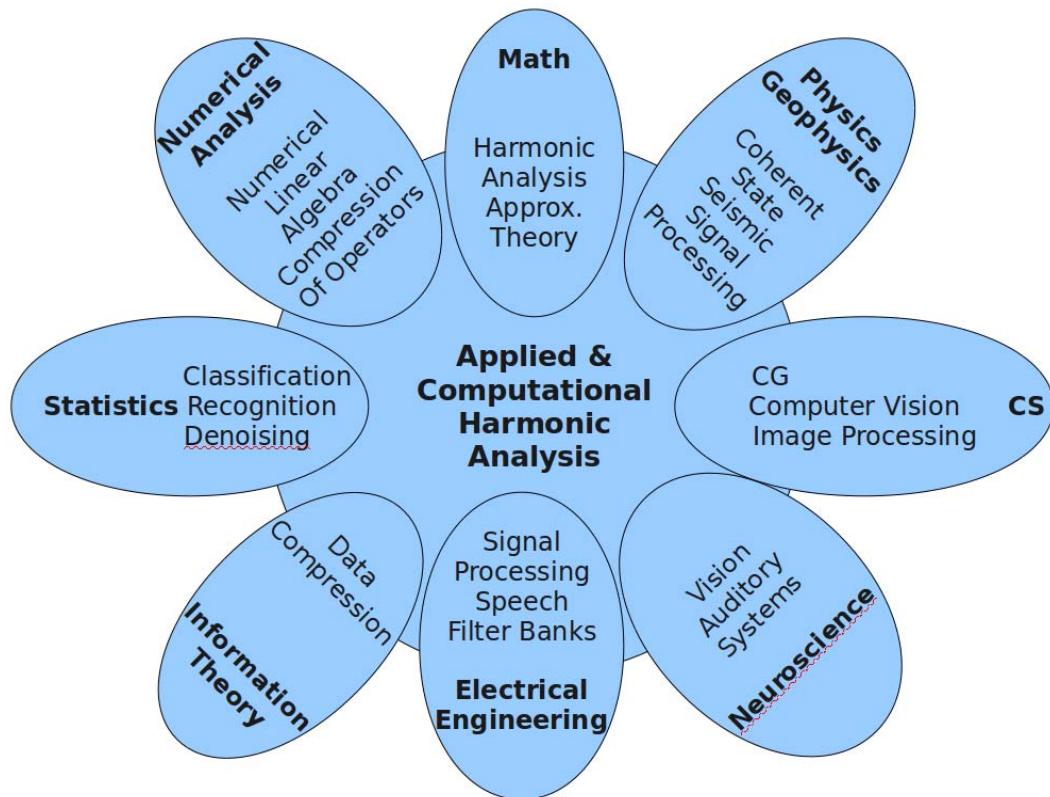


## Lecture 1 : Overture

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

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

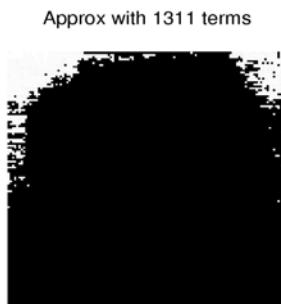
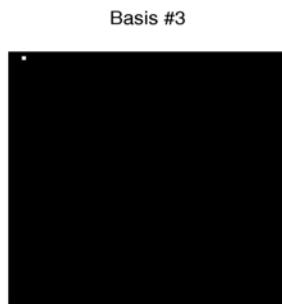
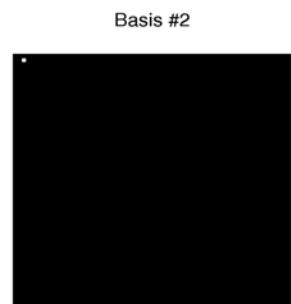
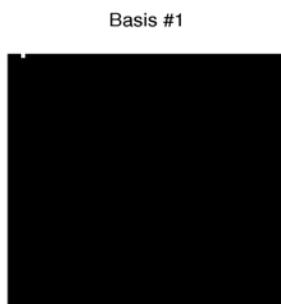
where  $n$ : small,  $\{\varphi_k\}$ : an ONB  
or more precisely,

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

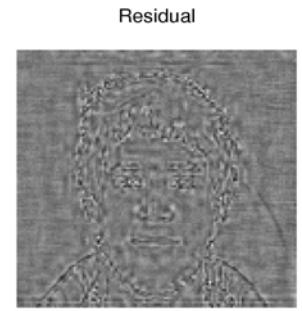
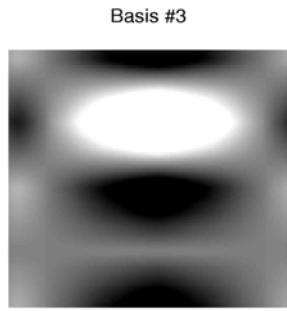
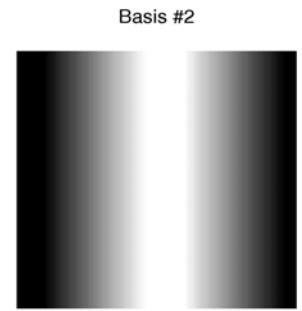
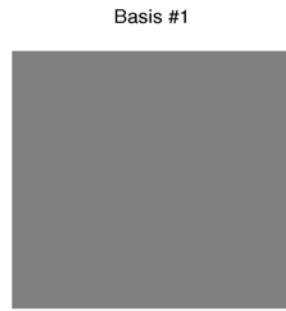
$\exists \beta > 0$

what class of fcns possess such rate of approximation?

A bad example :

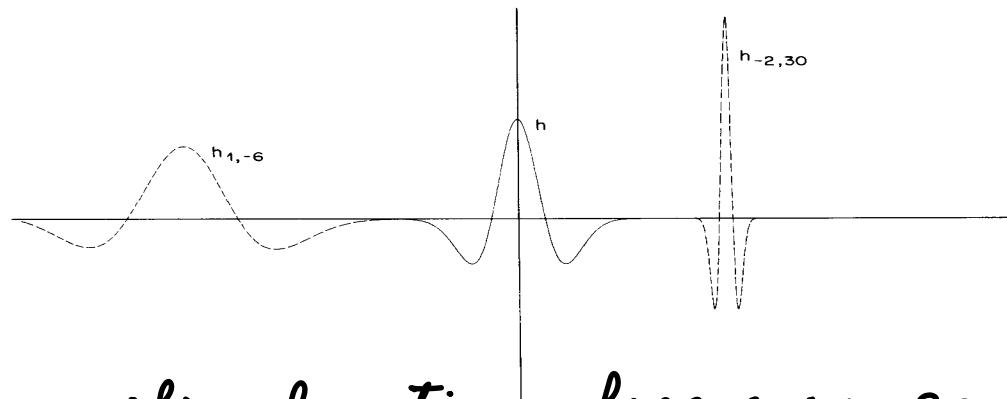


A good example:



## (2) Meaningful/useful

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

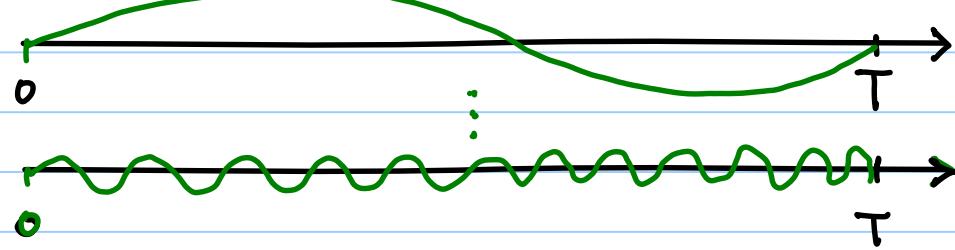


specific location, frequency, scale,  
duration, ...

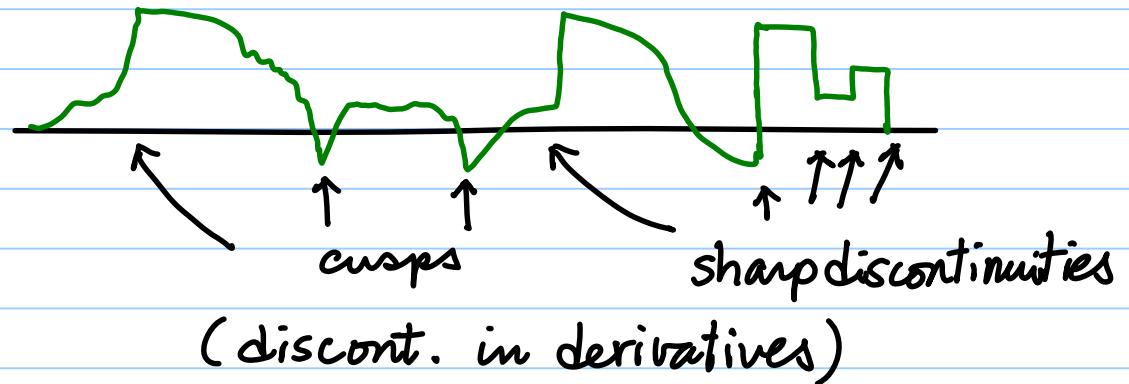
## Ex. 1. Seismic waves



→ 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.

$$x = \Phi^* f$$

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

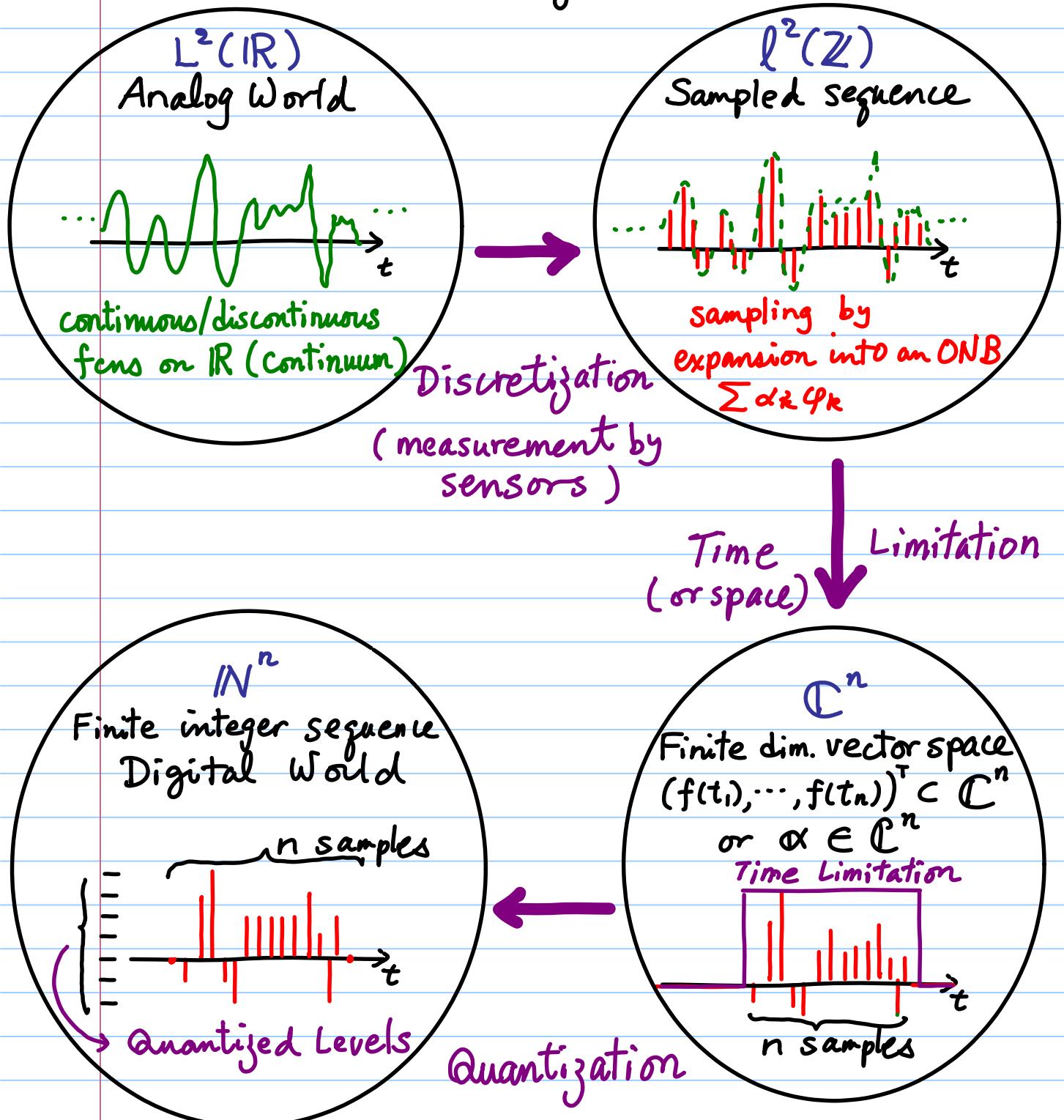
It's a philosophy called

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 course.
- \* Sampling can be viewed as an expansion of a signal into a special ONB  $\Rightarrow$  **The Shannon Sampling Theorem**

\* Until we cover 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)



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

