

# Introduction to Hilbert Space Frames

Robert Crandall

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## Introduction

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# What is a frame?

A frame is a generalization of a Hilbert space basis that is not necessarily linearly independent. A frame allows us to represent any vector as a set of "frame coefficients," and to reconstruct a vector from its coefficients in a numerically stable way.

# Why frames?

Often want to decompose a function in terms of functions that are not linearly independent.

Examples:

- ▶ Windowed Fourier transforms
- ▶ Wavelet transforms
- ▶ Non-uniform sampling

# Coefficient Representations of Hilbert Space Vectors

Often useful to represent a vector by a sequence of coefficients.  
Familiar in  $\mathbb{R}^n$ ; consider

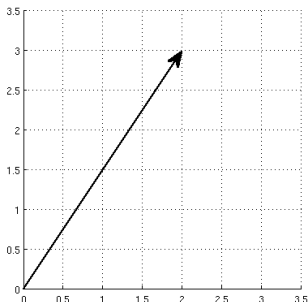


Figure: The vector  $(2, 3)$  in  $\mathbb{R}^2$  equal to  $2e_1 + 3e_2$

## Another Example: Fourier coefficients

In  $L^2(0,1)$ , the functions  $\phi_n = e^{2\pi in}$  ( $n \in \mathbb{Z}$ ) form an orthonormal basis. We can represent functions in terms of their Fourier coefficients:

$$c_n = \langle \phi_n, f \rangle$$

This is the familiar Fourier series. Coefficients tell us the "frequency content" of a function.

Function can be reconstructed from its coefficients using

$$f = \sum_n c_n e^{2\pi in}$$

We determine coefficient representations by taking inner products with some set of vectors  $\phi_n$ . What sets of vectors will give suitable coefficients?

*Frames!* Frame theory gives necessary and sufficient condition for  $\phi_n$  to give "suitable" coefficients for every  $f \in \mathcal{H}$ .

# The Frame Operator

For a Hilbert space  $\mathcal{H}$  and a set of vectors  $\{\phi_n\}$ , define a mapping  $F$  such that  $\forall x \in \mathcal{H}$ ,

$$(Fx)_n = \langle \phi_n, x \rangle.$$

If the  $\phi_n$  meet the "frame condition," then  $F$  is called a frame operator. Note that  $F$  is linear.

## What Vectors Give Suitable Coefficients?

We want to determine which sets of vectors  $\phi_n$  can be used to represent other vectors in a "suitable" way.

1. Want *unique* coefficients for any  $f \in \mathcal{H}$ 
  - ▶ Frame operator should be *injective*
2. Want to be able to go from vector to coefficients, and back again, in a *numerically stable* way
  - ▶ Frame operator should be invertible, at least on its range
  - ▶ Frame operator and its pseudo-inverse should be *bounded*

For this we need the *frame condition*.

# The Frame Condition

A set of vectors  $\{\phi_n\} \in \mathcal{H}$  is called a *frame* if there exist  $0 < A, B < \infty$  such that

$$A\|x\|^2 \leq \|Fx\|^2 \leq B\|x\|^2$$

where

$$\|Fx\|^2 = \sum |\langle \phi_n, x \rangle|^2$$

for all  $x \in \mathcal{H}$ .

If  $A = B$ ,  $\phi_n$  is said to be a *tight frame*.

## Why this frame condition?

- ▶  $\|F_x\|^2 \geq A\|x\|^2$  guarantees *uniqueness* of frame coefficients:

$$\|F_x - F_y\| = 0 \implies \|x - y\| = 0$$

- ▶  $\|F_x\|^2 \leq B\|x\|^2$  guarantees *numerical stability* in computing frame coefficients, as long as  $B$  is reasonable:

$$\|F(x + \delta) - F(x)\| = \|F\delta\| \leq \sqrt{B}\|\delta\|$$

- ▶ Also have stability when reconstructing a vector from its coefficients; we will see that  $F$  has a left inverse bounded by  $\frac{1}{\sqrt{A}}$ .

# Bases are frames

Any basis is also a frame. An orthonormal basis<sup>1</sup> is a tight frame with  $A = B = 1$ . However, a frame does not have to be a basis.

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<sup>1</sup>An orthonormal basis is sometimes *defined* as a set of orthonormal vectors satisfying the frame condition with  $A = B = 1$

## Not every frame is a basis

There are frames which are not bases. Let

$$F = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{pmatrix}$$

This is a frame operator, and its rows are the frame vectors. Frame bounds come from the singular value decomposition:

$$\sigma_1 = \sqrt{3}, \sigma_2 = 1$$

so

$$\|x\|^2 \leq \|Fx\|^2 \leq 3\|x\|^2$$

## A frame in $L^2$ which is not a basis

A *windowed Fourier frame* of  $L^2$  has the form

$$\{g_{n,k}(t) = g(t - nu_0)e^{ik\xi_0 t}\}$$

for some window function  $g(t)$ . This is useful for performing *localized* frequency analysis.

The frame coefficients  $(Fx)_n$  can be used to reconstruct any  $x$ .  
But how?

For any frame  $\{\phi_n\}$  and any vector  $x$ , we have the reconstruction formula

$$x = \sum_n (Fx)_n \tilde{\phi}_n$$

The  $\tilde{\phi}_n$  are *not necessarily equal to the  $\phi_n$* ! They make up the *dual frame*.

# The Frame Operator's Adjoint

To define the dual frame, we will need the adjoint of  $F$ .

For a frame  $\phi_n$  with frame operator  $F$ , the adjoint is the function  $F^* l^2 \rightarrow \mathcal{H}$  given by

$$F^* c = \sum_j c_j \phi_j$$

## The Dual Frame

For a frame  $\phi_n$  with frame operator  $F$ , the dual frame is

$$\tilde{\phi}_n = (F^*F)^{-1}\phi_j$$

This is also a frame. Its frame operator<sup>2</sup> is  $\tilde{F} = F(F^*F)^{-1}$ . Its frame bounds are given by

$$\frac{1}{B}\|x\|^2 \leq \|\tilde{F}x\|^2 \leq \frac{1}{A}\|x\|^2$$

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<sup>2</sup>Note that if the frame is an orthonormal basis,  $F$  is unitary and  $\tilde{F} = F!$

## Reconstruction formula

The frame operator and the dual frame operator satisfy

$$\tilde{F}^* F = F^* \tilde{F} = I$$

which gives the reconstruction formula<sup>3</sup> for any  $x \in \mathcal{H}$ :

$$x = \sum_j \langle \phi_j, x \rangle \tilde{\phi}_j = \sum_j \langle \tilde{\phi}_j, x \rangle \phi_j$$

<sup>3</sup>If the frame is an orthonormal basis, this reduces to  $x = \sum_j \langle \phi_j, x \rangle \phi_j$

## Example of a dual frame

Returning to our earlier example, with

$$F = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{pmatrix},$$

the dual frame operator is

$$F = \begin{pmatrix} 2/3 & -1/3 \\ -1/3 & 2/3 \\ 1/3 & 1/3 \end{pmatrix}.$$

The dual frame vectors are the rows of this matrix.

## Example of a dual frame (cont.)

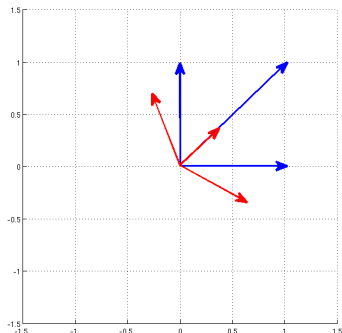


Figure: Frame from example (blue) and dual frame (red)

Say we want to expand  $x = (1, 2)^T$  in terms of this frame. We have  $\tilde{F}x = (0, 1, 1)^T$ , so

$$x = \begin{pmatrix} 1 \\ 2 \end{pmatrix} = 0 \begin{pmatrix} 1 \\ 0 \end{pmatrix} + 1 \begin{pmatrix} 0 \\ 1 \end{pmatrix} + 1 \begin{pmatrix} 1 \\ 1 \end{pmatrix}.$$

The dual set of coefficients is  $Fx = (1, 2, 3)^T$ , so the expansion in terms of the dual frame vectors is

$$x = \begin{pmatrix} 1 \\ 2 \end{pmatrix} = 1 \begin{pmatrix} 2/3 \\ -1/3 \end{pmatrix} + 2 \begin{pmatrix} -1/3 \\ 2/3 \end{pmatrix} + 3 \begin{pmatrix} 1/3 \\ 1/3 \end{pmatrix}.$$

## Classification of Euclidean space frames

A finite set of vectors  $\{\phi_n\} \in \mathbb{R}^n$  is a frame iff it spans  $\mathbb{R}^n$  and satisfies

$$\sum_n \|\phi_n\|^2 < \infty$$

Thus, any finite spanning set is a frame, and an infinite set can be a frame as long as the magnitudes of the frame vectors decay sufficiently quickly.

A special class of frames called Finite Normalized Tight Frames (FNTFs) exhibit a great deal of symmetry. These are tight frames consisting of finitely many unit vectors. Classification of these is more difficult, but has been done; see paper "Finite Normalized Tight Frames" in bibliography.

Some interesting results about FNTFs in  $\mathbb{R}^n$ :

- ▶ FNTFs in  $\mathbb{R}^n$  consisting of  $k$  vectors exist for all  $k \geq n$
- ▶ Frame bound of a FNTF measures "redundancy" of frame vectors; given by  $A = \frac{k}{n}$

- ▶ The FNTFs with  $A = 1$  are precisely the orthonormal bases in  $\mathbb{R}^n$
- ▶ The vertices of any Platonic solid are a frame in  $\mathbb{R}^n$ , as well as the vertices of some other highly symmetric shapes like the "soccer ball"
- ▶ Problem of classifying FNTFs is related to problem of equidistribution; how to "evenly" distribute  $k$  vectors on the unit sphere?

# The End

Thanks to Dr. Faris for advising this project.

References:

- ▶ A Wavelet Tour of Signal Processing, Stephan Mallat
- ▶ Ten Lectures on Wavelets, Ingrid Daubechies
- ▶ "Finite Normalized Tight Frames", J. Benedetto & M. Fickus, Advances in Computational Mathematics 18:357-385, 2003