

# Representation and approximation of data

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February 3, 2015



# Outline

## 1 Lecture 4: Overcomplete Representations

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# Frames

- Practical potential was not recognized until the 1990s.
- Among the generalizations of frames, many ideas have been proposed in the recent years, e.g., frames of subspaces (Casazza and Kutyniok), pseudo-frames (Li and Ogawa), fusion frames (Casazza, Fickus, Kutyniok), outer frames (Aldrubi, Cabrelli, and Molter),  $g$ -frames (Sun), and multiplicative frames (Benedetto).
- Frames have a simple interpretation in the context of finite dimensional vector spaces.

# Finite frames

## Definition

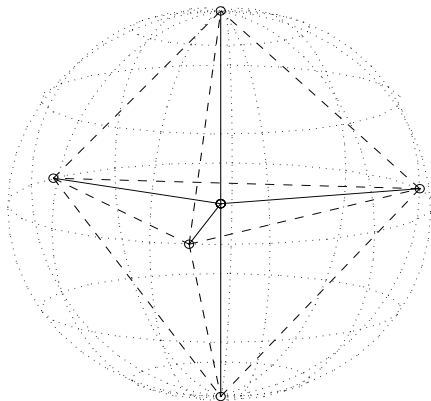
A collection  $\{x_n\}_{n=1}^N$  in a Hilbert space  $\mathbb{H}$  is a *frame* for  $\mathbb{H}$  if there exist  $0 < A \leq B < \infty$  such that

$$\forall x \in \mathbb{H}, A\|x\|^2 \leq \sum_{n=1}^N |\langle x, x_n \rangle|^2 \leq B\|x\|^2.$$

The constants  $A$  and  $B$  are the *frame bounds*. If  $A = B$ , the frame is an *A-tight* frame.

- Any spanning set of vectors in  $\mathbb{R}^d$  is a *frame* for  $\mathbb{R}^d$ .
- However, the spanning property does not indicate the value of frames for representation and stability in noisy environments.

# Example of a frame



Typical frames are redundant systems with more elements than the dimensionality of the space they represent.

# From data to frame operators

- Given data space  $X$  of  $N$  vectors  $x_n \in \mathbb{R}^D$ . Without loss of generality, assume  $\sum x_n = 0$  (subtract mean).
- Let  $P$  be  $D \times N$  matrix whose columns are the data vectors  $x_n$ .
- Let  $\mathbb{H} = \text{span}\{x_n\}_{n=1}^N \subseteq \mathbb{R}^D$ . Define  $L : \mathbb{H} \rightarrow \mathbb{R}^N$ ,

$$v \mapsto P^* v = L(v) = \{\langle v, x_n \rangle\} \in \mathbb{R}^N,$$

and its Hilbert space adjoint  $L^* : \mathbb{R}^N \rightarrow \mathbb{H} \subseteq \mathbb{R}^D$ ,

$$w \mapsto L^*(w) = \sum_{n=1}^N w[n]x_n, \quad w = (w[1], w[2], \dots, w[N]).$$

- $L$  is the *Bessel (analysis)* operator, and  $L^*$  is the *synthesis* operator.

# Finite frames

- Recall the Bessel operator  $L(v) = \{\langle v, x_n \rangle\} \in \mathbb{R}^N$ .
- The *frame operator* for  $\mathbb{H}$  is

$$S = L^*L : \mathbb{H} \rightarrow \mathbb{H}.$$

- $\{x_n\}_{n=1}^N$  is a *frame* for  $\mathbb{H}$  if

$$\exists 0 < A \leq B < \infty \quad \text{such that} \quad AI \leq S \leq BI.$$

- $AI \leq S \leq BI$  implies that  $S$  is invertible and that  $B^{-1}I \leq S \leq A^{-1}I$ .



# Finite frames

## Theorem

**a.**  $\{x_n\}_{n=1}^N$  is a frame for  $\mathbb{H}$  if and only if

$$\forall v \in \mathbb{H}, v = \sum_{n=1}^N \langle v, S^{-1}(x_n) \rangle x_n = \sum_{n=1}^N \langle v, x_n \rangle S^{-1}(x_n).$$

**b.**  $\{x_n\}_{n=1}^N$  is an  $A$ -tight frame for  $\mathbb{H}$  if and only if  $S = AI$ .

P. Casazza, "The art of frame theory," arXiv preprint math/9910168, 1999.

# Tight frame vs. ONB

## Theorem (Vitali, 1921)

Let  $H$  be a Hilbert space,  $\{x_n\} \subseteq H$ ,  $\|x_n\| = 1$ .

$\{x_n\}$  is 1-tight  $\Leftrightarrow \{x_n\}$  is an ONB.

*Proof.* If  $\{x_n\}$  is 1-tight, then  $\forall y \in H$ ,  $\|y\|^2 = \sum_n |\langle y, x_n \rangle|^2$ . Since each  $\|x_n\| = 1$ , we have

$$1 = \|x_n\|^2 = \sum_k |\langle x_n, x_k \rangle|^2 = 1 + \sum_{k, k \neq n} |\langle x_n, x_k \rangle|^2$$

$$\Rightarrow \sum_{k \neq n} |\langle x_n, x_k \rangle|^2 = 0 \Rightarrow \forall n \neq k, \langle x_n, x_k \rangle = 0.$$

# The role of covariance

- The frame operator  $S$  can be written as

$$S : \mathbb{H} \rightarrow \mathbb{H}, v \mapsto \sum_{n=1}^N \langle v, x_n \rangle x_n = (PP^*)v,$$

where  $PP^*$  is  $D \times D$ .

- Hence, up to a scaling factor and a translation,  $S$  is the linear operator identified with the  $D \times D$  symmetric covariance matrix  $C = \frac{1}{N}PP^*$  of the data space, i.e.

$$C = \frac{1}{N} \left( \sum_{j=1}^N x_j[m]x_j[n] \right)_{m,n=1}^D, \quad x_j = (x_j[1], \dots, x_j[D]) \in \mathbb{R}^D.$$

# The Grammian

- The *Grammian operator* for  $X$  is

$$G = LL^* : \mathbb{R}^N \rightarrow \mathbb{R}^N.$$

Thus,  $G$  is  $N \times N$  and

$$G = \{\langle x_m, x_n \rangle\}_{m,n=1}^N = P^*P.$$

- $G = LL^*$ ,  $N \times N$ , and  $S = L^*L$ ,  $D \times D$ , have the same non-zero eigenvalues, a fact we shall exploit.

# FUNTFs

Let  $\mathbb{K}$  be an  $r$ -dimensional Hilbert space, and let  $\psi_n \in \mathbb{K}$ ,  $\|\psi_n\| = 1$ ,  $n = 1, \dots, s$ .

- If  $\{\psi_n\}_{n=1}^s$  is a **finite unit norm tight frame (FUNTF)** for  $\mathbb{K} = \mathbb{R}^r$ , then

$$\forall y \in \mathbb{K}, y = \frac{s}{r} \sum_{n=1}^s \langle y, \psi_n \rangle \psi_n.$$

- *Problem:* Find FUNTFs analytically, effectively, and computationally.

# Characterization of FUNTFs

The total frame force potential energy is:

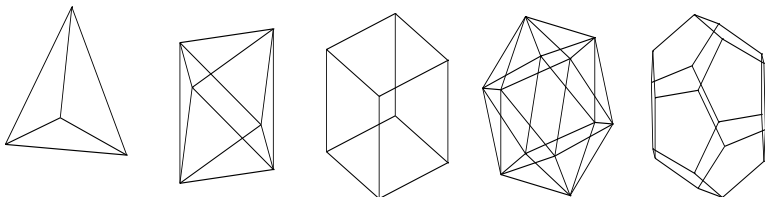
$$TFP(\{\psi_n\}) = \sum_{m=1}^s \sum_{n=1}^s |\langle \psi_m, \psi_n \rangle|^2.$$

## Theorem

- Let  $s = r$ . The minimum value of  $TFP$ , for the frame force and  $s$  variables, is  $s$ ; and the minimizers are precisely the orthonormal sets of  $s$  elements for  $\mathbb{K}$ .
- Let  $s > r$ . The minimum value of  $TFP$ , for the frame force and  $s$  variables, is  $s^2/r$ ; and the minimizers are precisely the FUNTFs of  $s$  elements for  $\mathbb{K}$ .

J. J. Benedetto and M. Fickus, "Finite normalized tight frames," Adv. Comp. Math., 2003, Vol. 18, pp. 357–385.

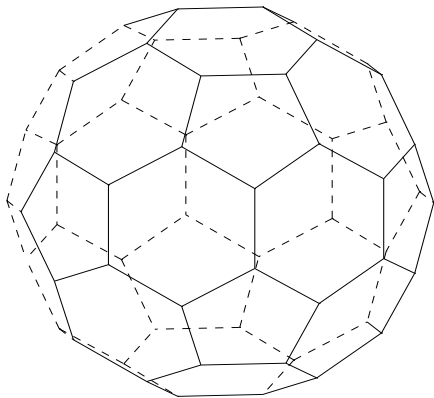
# Examples of FUNTFs



**Figure :** The vertices of the Platonic solids are examples of finite unit norm tight frames.

R. Vale, S. Waldron, "The vertices of the Platonic solids are tight frames," in: Proceedings of the Conference on Advances in Constructive Approximation (M. Neamtu, E. B. Saff, eds.). Brentwood, TN: Nashboro Press, 2004.

# Examples of FUNTFs



**Figure :** The truncated icosahedron (also known as the “soccer ball” or “bucky ball”) forms a tight frame for  $n$ -dimensional Euclidean space.



# Motivation for frames

- Different classes of interest may not be orthogonal to each other; however, they may be captured by different frame elements. It is plausible that classes may correspond to elements in a frame but not elements in a basis.
- A *frame* generalizes the concept of an orthonormal basis. Frame elements are non-orthogonal.
- Frames provide over-complete data decompositions, often useful for numerical stability and noise reduction.