

Frames

A frame is slightly more general than a Riesz basis — we remove the requirement of linear independence.

Definition: A frame for a Hilbert space H is a set $\{\varphi_n\}_{n \in T}$ such that $\exists 0 < A \leq B < \infty$ with

$$A \cdot \|f\|_H^2 \leq \sum_{n \in T} |\langle f, \varphi_n \rangle|^2 \leq B \cdot \|f\|_H^2$$

for all $f \in H$.

Interesting fact: There exist frames which are not orthonormal bases which have $A = B$ (a Parseval relation holds)

These are called tight frames.

Frame operator: $\Phi: H \rightarrow \ell_2(T)$ (infinite dimensions)

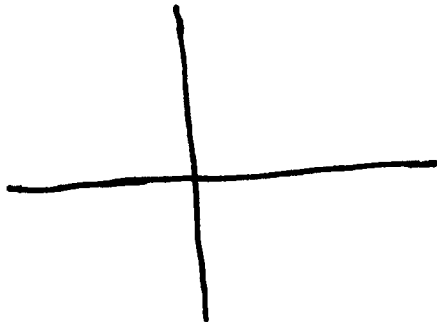
or $\Phi: \mathbb{C}^n \rightarrow \mathbb{C}^m$ (finite dimensions)

↳ just as before Φ has the φ_k^* as rows — it is an $m \times n$ matrix

Example: $H = \mathbb{R}^2$

$$\theta_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad \theta_2 = \begin{bmatrix} -1/2 \\ \sqrt{3}/2 \end{bmatrix}, \quad \theta_3 = \begin{bmatrix} -1/2 \\ -\sqrt{3}/2 \end{bmatrix}$$

Sketch:



Frame operator: $\Phi = \begin{bmatrix} 1 & 0 \\ -1/2 & \sqrt{3}/2 \\ -1/2 & -\sqrt{3}/2 \end{bmatrix}$

$$\Phi^* \Phi = \begin{bmatrix} & \\ & \end{bmatrix}$$

$$\Rightarrow \underline{\hspace{2cm}} \leq f^* \Phi^* \Phi f \leq \underline{\hspace{2cm}}$$

$$\Rightarrow A = B = \underline{\hspace{2cm}}$$

Example 2: Unions of orthobases

$$\{\theta_{k_1}^1\}_{k_1 \in T_1} \cup \{\theta_{k_2}^2\}_{k_2 \in T_2} \cup \dots \cup \{\theta_{k_M}^M\}_{k_M \in T_M}$$

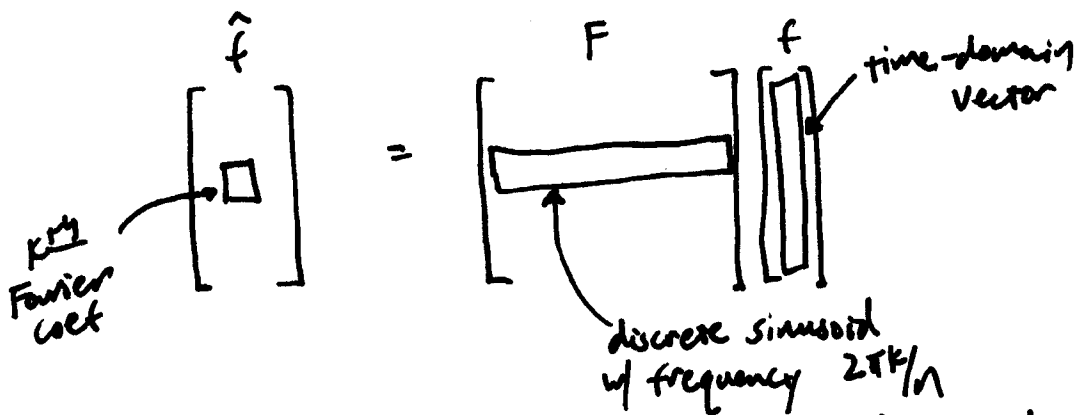
Then

$$\begin{aligned} \|\mathbb{E}f\|^2 &= \sum_{k_1 \in T_1} |\langle f, \theta_{k_1}^1 \rangle|^2 + \sum_{k_2 \in T_2} |\langle f, \theta_{k_2}^2 \rangle|^2 + \dots + \sum_{k_M \in T_M} |\langle f, \theta_{k_M}^M \rangle|^2 \\ &= \|f\|_H^2 + \|f\|_H^2 + \dots + \|f\|_H^2 \\ &= M \cdot \|f\|_H^2 \end{aligned}$$

Example 3: Oversampled DFT

Recall that the $n \times n$ DFT matrix F is

$$(F)_{k+l, l} = \frac{1}{\sqrt{n}} e^{-j \frac{2\pi k l}{n}} \quad l, k = 0, \dots, n-1$$



We know very well by now that the rows of F form an orthobasis for \mathbb{E}

n sinusoids at equally spaced freqs. between $0 \neq 2\pi$

We will look at an oversampled DFT,

$$(\Phi)_{k+l, l} = \frac{1}{\sqrt{n}} e^{-j \frac{2\pi k l}{m}} \quad \begin{array}{l} l = 0, \dots, n-1 \\ k = 0, \dots, m-1 \end{array}$$

Rows are $m > n$ complex sinusoids with equally spaced freqs between 0 & 2π .

In MATLAB: `fft(f, m)` $\text{length}(f) = n$

$$\begin{bmatrix} \hat{f} \end{bmatrix} = \begin{bmatrix} \Phi \end{bmatrix} \begin{bmatrix} f \end{bmatrix}$$

Now, $(\Phi \Phi^*)_{l+l, l+l} = \langle \cdot, \cdot \rangle$

$$= \sum_{k=0}^{m-1}$$

$$= \{$$

Thus

$$\sum_{k=1}^m |\langle f, \theta_k \rangle|^2 = f^* \Phi^* \Phi f$$

$$= \underline{\hspace{2cm}}$$

General rule:

Oversampled systems are frames.

When the sampling is equispaced, the system is a tight frame

(Much more about this on Homework #1)

Note that for a tight frame

$$\Phi^* \Phi = A \cdot I$$

So we have a Parseval theorem:

$$\|\Phi f\|_2^2 = \langle \Phi f, \Phi f \rangle = \langle f, \Phi^* \Phi f \rangle$$

$$= A \langle f, f \rangle$$

$$= A \cdot \|f\|_2^2$$

← energy is simply scaled by a constant

Also, inner products are preserved

$$\langle \Phi f, \Phi g \rangle = A \langle f, g \rangle$$

IMPORTANT NOTE:

It is NOT true that $\Phi \Phi^* = A \cdot I$

(It can't be, as it is a $m \times m$ matrix of rank n , $n < m$)

Reconstructing from frame coefficients

How do we go from $\{\langle f, \theta_k \rangle\}_{k \in T}$ back to f ?

Since the $\{\theta_k\}_{k \in T}$ are not linearly independent, given an arbitrary $\alpha \in L_2(T)$, there may not be an f with

$$\Phi f = \alpha$$

i.e. in finite dimensions, Φ is an $m \times n$ matrix

$$\begin{bmatrix} \alpha \end{bmatrix} = \begin{bmatrix} \Phi \end{bmatrix} \begin{bmatrix} f \end{bmatrix}$$

Can we solve this system for any α ?
(Of course not.)

Instead, we will take our best shot, reconstructing with the pseudo inverse.

Pseudo Inverse

The pseudo inverse of a frame operator $\Phi: H \rightarrow \ell_2(\Gamma)$ is denoted $\Phi^\dagger: \ell_2(\Gamma) \rightarrow H$, and satisfies:

① left inverse

$$\Phi^\dagger \Phi f = f \quad \forall f \in H$$

② minimum error

$$\Phi^\dagger \alpha = \operatorname{argmin}_f \|\alpha - \Phi f\|_{\ell_2(\Gamma)}$$

"Which signal has $\{\langle f, \theta_k \rangle\}_{k \in \Gamma}$ closest to α ?"

It is not hard to show that

$$\Phi^\dagger = (\Phi^* \Phi)^{-1} \Phi^*$$

→ this exists if the frame bounds satisfy $0 < A \leq B < \infty$

Note: If $\{\sigma_k\}_{k \in \Gamma}$ is tight, then

$$\Phi^* \Phi = A \cdot I$$

$$\Rightarrow \Phi^\dagger = \frac{1}{A} \cdot \Phi^*$$

pseudo inverse = applying the adjoint (rescaled), just as in the orthonormal case

Reproducing formula

$$\begin{aligned} f &= \mathbb{I}^+ \mathbb{I} f = (\mathbb{I}^+ \mathbb{I})^{-1} \mathbb{I}^+ \mathbb{I} f \\ &= \sum_{k \in \mathcal{T}} \langle f, \theta_k \rangle \tilde{\theta}_k \end{aligned}$$

where

$$\tilde{\theta}_k = (\mathbb{I}^+ \mathbb{I})^{-1} \theta_k$$

If $\{\theta_k\}_{k \in \mathcal{T}}$ is tight, $\tilde{\theta}_k = \underline{\hspace{2cm}}$

and the RF looks almost the same as in the orthonormal case

$$f = \frac{1}{A} \sum_{k \in \mathcal{T}} \langle f, \theta_k \rangle \theta_k$$

Conditioning

Arguing as in the non-orthogonal Riesz basis case

$$\frac{1}{B} \|f\|_H^2 \leq \sum_{k \in \mathcal{T}} |\langle f, \tilde{\theta}_k \rangle|^2 \leq \frac{1}{A} \|f\|_H^2$$

$$\text{If } \lambda_k = \langle f, \theta_k \rangle + \varepsilon_k$$

$$\text{and } \tilde{f} = \sum_{k \in \mathcal{T}} \lambda_k \tilde{\theta}_k$$

$$\text{then } \|f - \tilde{f}\|_H^2 \leq \frac{1}{A} \cdot \|\varepsilon\|_{\ell_2(\mathcal{T})}^2$$

BUT (and this is huge) this is really a worst case bound — the best case would be $z \in \text{Null}(I^T) \Rightarrow \|f - \tilde{f}\|_H^2 = 0$

If the error is random, then it is an "average" of these two cases.

Recall: Gaussian random vectors

Suppose $z \in \mathbb{C}^m$ is iid Gaussian
 $z(k) \sim \mathcal{N}(0, \sigma^2)$

then

$$E \|z\|_2^2 = m \cdot \sigma^2$$

Let A be an $n \times m$ matrix.

What is $E \|Az\|_2^2$?

$$\begin{aligned} E \|Az\|_2^2 &= E \langle Az, Az \rangle = E \langle z, A^* A z \rangle \\ &= E \langle z, U^* \Lambda U z \rangle = E \langle U z, \Lambda U z \rangle \end{aligned}$$

U : \perp -normal
 Λ : diagonal, ≥ 0

Set $z' = Uz$

z' is also iid Gaussian (since U \perp -normal)

$$z'(k) \sim \mathcal{N}(0, \sigma^2)$$

We have

$$E \|Az\|_2^2 = E \langle z', \Lambda z' \rangle$$

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Oversampling for noise reduction

Signal $f \in \mathbb{C}^n$

Say we observe the DFT coefficients of f in the presence of noise, and then attempt to reconstruct f :

$\Phi = \text{DFT}$ (orthonormal)

Observe: $y = \Phi f + \varepsilon$ $\varepsilon(k) \sim N(0, \sigma^2)$

Reconstruct: $\tilde{f} = \Phi^* y = \Phi^* (\Phi f + \varepsilon)$
 $= f + \Phi^* \varepsilon$

Mean-square error:

$$E \|f - \tilde{f}\|_2^2 = E \|\Phi^* \varepsilon\|_2^2 = n\sigma^2$$

Now, say we observe oversampled DFT coefficients in the presence of noise

$\Phi =$ oversampled DFT, $m \times n$, $m > n$

$$(\Phi)_{k+l, \ell+l} = \frac{1}{\sqrt{n}} e^{-j2\pi k \ell / m} \quad \begin{array}{l} \ell = 0, \dots, n-1 \\ k = 0, \dots, m-1 \end{array}$$

We saw previously that Φ is a tight frame with $A=B=n/m$, i.e.

$$\Phi^* \Phi = \frac{n}{m} \cdot I$$

Observe: $y = \Phi f + \varepsilon$ ($y \in \mathbb{C}^m$, $\varepsilon \in \mathbb{C}^m$)

Reconstruct w/ pseudo inverse:

$$\begin{aligned} \tilde{f} &= (\Phi^* \Phi)^{-1} \Phi^* y = (\Phi^* \Phi)^{-1} \Phi^* (\Phi f + \varepsilon) \\ &= f + \frac{n}{m} \Phi^* \varepsilon \end{aligned}$$

(since $(\Phi^* \Phi)^{-1} = \frac{m}{n} \cdot I$)

Mean-squared error:

$$\begin{aligned} E \|f - \tilde{f}\|_2^2 &= E \left\| \frac{n}{m} \Phi^* \varepsilon \right\|_2^2 \\ &= \left(\frac{n}{m}\right)^2 \cdot \text{Trace}(\Phi \Phi^*) \cdot \sigma^2 \\ &= \left(\frac{n}{m}\right)^2 \cdot m \cdot \sigma^2 \end{aligned}$$

$$\boxed{= \left(\frac{n}{m}\right) \cdot n \cdot \sigma^2}$$

Since $N < M$, the reconstruction error is smaller.

$\frac{1}{\text{oversampling}}$ = noise reduction factor!

A portion of the noise lies in the null space of $\underline{\mathbb{E}}^+$.

Key application: Oversampled A/D conversion
We can quantize samples more coarsely if we sample them faster.