



Mathematical Foundations of Signal Processing

Bases and Least Squares Approximation

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Previous lecture

- Geometry is key to intuition!
 - From finite to infinite dimensional spaces
- Hilbert spaces: Complete inner product spaces
- Linear operators and their adjoints
 - Bounded linear operators, eigenvalues and eigenvectors
- Projection: Oblique and orthogonal
 - Idempotency and self-adjointness
 - Projection theorem
- Approximation
 - Best/least squares approximation
- Direct sums and decompositions

Today

- 1 Basis representation and expansions
 - Riesz bases
 - Orthonormal bases
 - Biorthogonal bases
- 2 Frame representation and expansions
 - Definition
 - Tight frames
 - General frames
- 3 Matrix representation of linear operators
 - Change of basis
 - Matrices
- 4 Gram-Schmidt orthonormalization and polynomial approximation

Readings:

- Section 2.5 of Chapter 2, "From Euclid to Hilbert", of *Foundations of Signal Processing*

Bases

Definition (Basis)

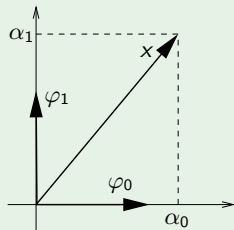
- $\Phi = \{\varphi_k\}_{k \in \mathcal{K}} \subset V$ is a **basis** when
 - ➊ Φ is linearly independent and
 - ➋ Φ is complete in V : $V = \overline{\text{span}}\{\Phi\}$
- **Expansion formula:** any $x \in V$, $x = \sum_{k \in \mathcal{K}} \alpha_k \varphi_k$
 - $\{\alpha_k\}_{k \in \mathcal{K}}$: is unique
 - α_k : expansion coefficients

Example

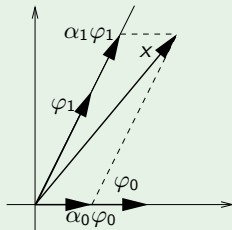
- The standard basis for \mathbb{R}^N
 $e_k = [0 \ 0 \ \cdots \ 0 \ 1 \ 0 \ 0 \ \cdots \ 0]^T, \ k = 0, \dots, N-1$
any $v \in \mathbb{R}^N$: $v = \sum_{k=0}^{N-1} v_k e_k$

Bases

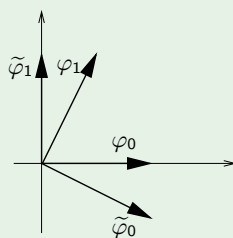
Examples



(a) Orthonormal basis



(b) Biorthogonal basis expansion



(c) and dual basis

Riesz bases

Definition (Riesz Basis)

- In a Hilbert space H , $\Phi = \{\varphi_k\}_{k \in \mathcal{K}}$ is a **Riesz basis** when

- 1 Φ is a basis

- 2 There exists $0 < \lambda_{\min} \leq \lambda_{\max} < \infty$ s.t.

for any $x = \sum_{k \in \mathcal{K}} \alpha_k \varphi_k \in H$

$$\lambda_{\min} \|x\|^2 \leq \sum_{k \in \mathcal{K}} |\alpha_k|^2 \leq \lambda_{\max} \|x\|^2$$

- Numerical stability when $\lambda_{\min} \approx \lambda_{\max}$

Example

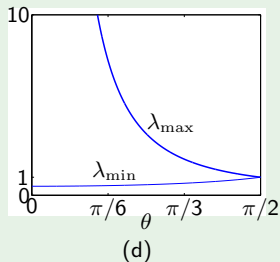
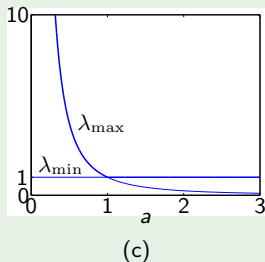
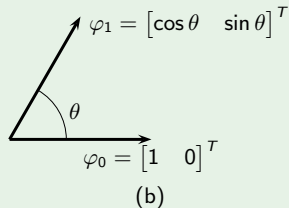
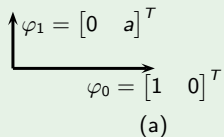
In \mathbb{R}^N

- Let $\Phi = [\varphi_0 \ \varphi_1 \ \dots \ \varphi_{N-1}]$ be a Riesz basis and define $G = \Phi^* \Phi$. Then
 $1/\lambda_{\max}$ = minimum eigenvalue of G
 $1/\lambda_{\min}$ = maximum eigenvalue of G

Riesz bases

Example

Eigenvalues of $\Phi_a^* \Phi_a$



Operators associated with Riesz bases

Definition (Basis synthesis operator)

- **Synthesis** operator

- $\Phi : \ell^2(\mathcal{K}) \rightarrow H$, s.t. $\Phi\alpha = \sum_{k \in \mathcal{K}} \alpha_k \varphi_k$

- $\|\Phi\| \leq 1/\sqrt{\lambda_{\min}}$ (follows from definition)

- **Adjoint:** $y \in H$

$$\langle \Phi\alpha, y \rangle = \left\langle \sum_{k \in \mathcal{K}} \alpha_k \varphi_k, y \right\rangle = \sum_{k \in \mathcal{K}} \alpha_k \langle y, \varphi_k \rangle^*$$

Definition (Basis analysis operator)

- **Analysis** operator

- $\Phi^* : H \rightarrow \ell^2(\mathcal{K})$ $(\Phi^*x)_k = \langle x, \varphi_k \rangle, \quad k \in \mathcal{K}$

- $\|\Phi^*\| \leq 1/\sqrt{\lambda_{\min}}$

- Note that the analysis operator is the adjoint of the synthesis operator

Orthonormal bases

Definition (Orthonormal basis)

- $\Phi = \{\varphi_k\}_{k \in \mathcal{K}} \subset H$ is an **orthonormal basis** for H when

- 1 Φ is a basis for H and

- 2 Φ is an orthonormal set

$$\langle \varphi_i, \varphi_k \rangle = \delta_{i-k} \text{ for all } i, k \in \mathcal{K}$$

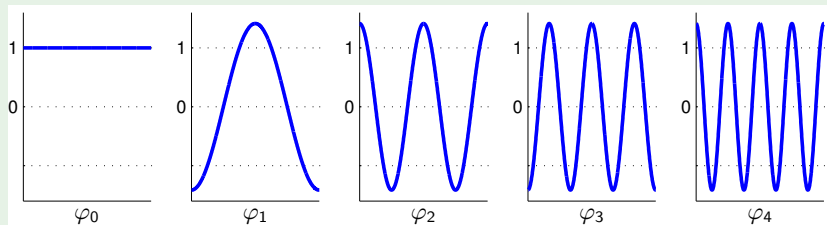
- If Φ is an orthogonal **set**, then it is linearly independent
- If $\overline{\text{span}}\{\Phi\} = H$ and Φ is an orthogonal **set**, then Φ is an orthogonal **basis** for H

If we also have $\|\varphi_k\| = 1$, then Φ is an orthonormal basis

Orthonormal basis

Example

Consider the Hilbert space $H = \{f \in \mathcal{L}^2([-\frac{1}{2}, \frac{1}{2}]) : f(-x) = f(x)\}$

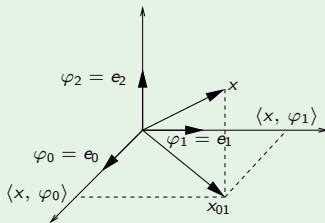


Orthonormal basis expansions

Definition (Orthonormal basis expansions)

- $\Phi = \{\varphi_k\}_{k \in \mathcal{K}}$ orthonormal basis for H , then for any $x \in H$:
 $\alpha_k = \langle x, \varphi_k \rangle$ for $k \in \mathcal{K}$, or $\alpha = \Phi^* x$, α is unique
- Synthesis:
$$\begin{aligned} x &= \sum_{k \in \mathcal{K}} \langle x, \varphi_k \rangle \varphi_k \\ &= \Phi \alpha = \Phi \Phi^* x \end{aligned}$$

Example



Orthonormal basis: Parseval's equality

Theorem (Parseval's equalites)

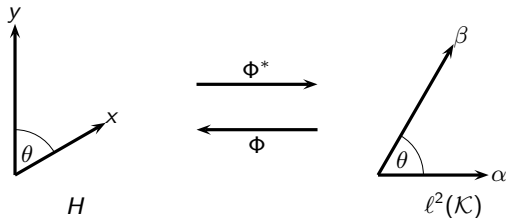
- $\Phi = \{\varphi_k\}_{k \in \mathcal{K}}$ orthonormal basis for H

$$\|x\|^2 = \sum_{k \in \mathcal{K}} |\langle x, \varphi_k \rangle|^2 = \|\Phi^* x\|^2 = \|\alpha\|^2$$

- In general:

$$\langle x, y \rangle = \langle \Phi^* x, \Phi^* y \rangle = \langle \alpha, \beta \rangle$$

with $\alpha_k = \langle x, \varphi_k \rangle, \beta_k = \langle y, \varphi_k \rangle$



Orthonormal bases

- On $\ell^2(\mathcal{K})$: $\Phi^* \Phi = I$
- On H : $\Phi \Phi^* = I$
- Analysis and synthesis operators associated with an orthonormal basis are **unitary**
- Isometry between any separable Hilbert space H and $\ell^2(\mathbb{Z})$
- How about approximation?

Orthogonal projection and decomposition

Theorem

- $\Phi = \{\varphi_k\}_{k \in \mathcal{I}} \subset H, \quad \mathcal{I} \subset \mathcal{K}$

$$P_{\mathcal{I}}x = \sum_{k \in \mathcal{I}} \langle x, \varphi_k \rangle \varphi_k = \Phi_{\mathcal{I}} \Phi_{\mathcal{I}}^* x$$

is the *orthogonal projection* of x onto $S_{\mathcal{I}} = \overline{\text{span}}(\{\varphi_k\}_{k \in \mathcal{I}})$

- Φ induces an orthogonal decomposition

$$H = \bigoplus_{k \in \mathcal{K}} S_{\{k\}} \quad \text{where } S_{\{k\}} = \text{span}(\varphi_k)$$

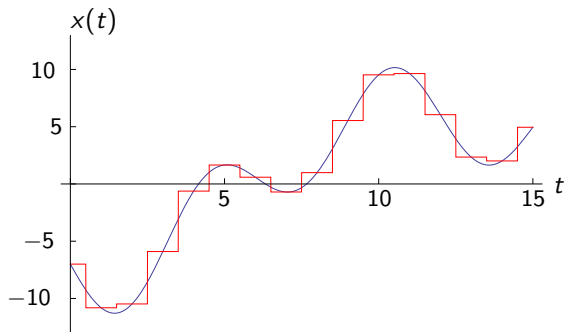
Example

$$S = \{[x_0 \quad x_1 \quad x_2]^T \in \mathbb{C}^3 \mid x_1 = x_0 + x_2\}$$

$$= \text{span}(\{\frac{1}{\sqrt{2}}[1 \quad 1 \quad 0]^T, \frac{1}{\sqrt{6}}[-1 \quad 1 \quad 2]^T\})$$

$$P_S x = \sum_{k=0}^1 \langle x, \varphi_k \rangle \varphi_k = \frac{1}{3} \begin{bmatrix} 2 & 2 & -1 \\ 1 & 1 & 1 \\ -1 & 1 & 2 \end{bmatrix} x$$

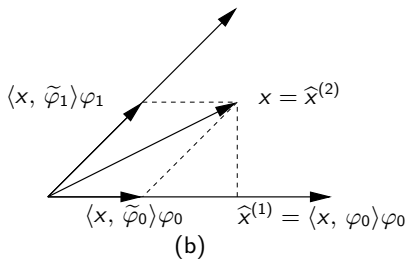
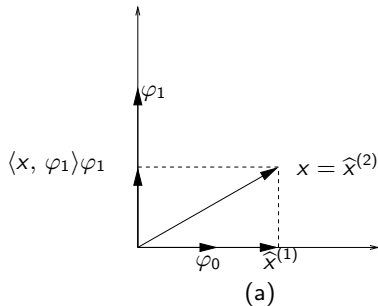
Orthogonal projection on $\mathcal{L}^2(\mathbb{R})$



The local averaging operator A and its adjoint A^* are such that $AA^* = I$, so $P = A^*A$ is an orthogonal projection

Best approximation in orthogonal bases

- $\{\varphi_k\}_{k \in \mathcal{K}}$ orthogonal basis
- $\hat{x}^{(k)}$ = best approximation of x in $\{\varphi_0, \dots, \varphi_{k-1}\}$
- $\hat{x}^{(0)} = 0$ and $\hat{x}^{(k+1)} = \hat{x}^{(k)} + \langle x, \varphi_k \rangle \varphi_k$ for $k = 0, 1, \dots$
 successive approximation



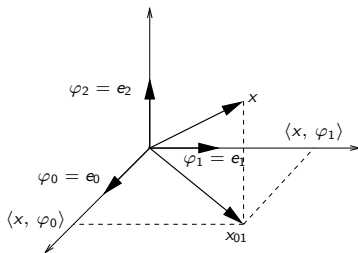
Bessel's inequality

Theorem (Bessel's inequality)

- $\Phi = \{\varphi_k\}_{k \in \mathcal{I}} \subset H$, Φ orthonormal set

$$\begin{aligned}\|x\|^2 &\geq \sum_{k \in \mathcal{I}} |\langle x, \varphi_k \rangle|^2 \\ &= \|\Phi_{\mathcal{I}}^* x\|^2\end{aligned}$$

- Equality when Φ is complete, i.e. an orthonormal basis
- Example: Note that $\|x_{01}\| \leq \|x\|$



Biorthogonal pairs of bases

Definition

Biorthogonal pairs of bases

- $\Phi = \{\varphi_k\}_{k \in \mathcal{K}} \subset H$, and $\tilde{\Phi} = \{\tilde{\varphi}_k\}_{k \in \mathcal{K}} \subset H$ is a biorthogonal pair of bases when
 - 1 Φ and $\tilde{\Phi}$ are **both bases for H**
 - 2 Φ and $\tilde{\Phi}$ are **biorthogonal**
- $$\langle \varphi_i, \tilde{\varphi}_k \rangle = \delta_{i-k} \text{ for all } i, k \in \mathcal{K}$$
- Roles of Φ and $\tilde{\Phi}$ are interchangeable

Example

$$\begin{aligned} \varphi_0 &= \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, & \varphi_1 &= \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}, & \varphi_2 &= \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, & \Phi &= \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 1 \\ 0 & 1 & 1 \end{bmatrix} \\ \tilde{\varphi}_0 &= \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}, & \tilde{\varphi}_1 &= \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, & \tilde{\varphi}_2 &= \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}, & \Phi^{-1} &= \begin{bmatrix} 0 & 1 & -1 \\ -1 & 1 & 0 \\ 1 & -1 & 1 \end{bmatrix} \end{aligned}$$

Biorthogonal basis expansion

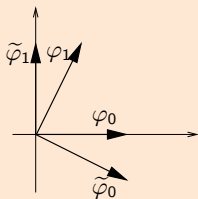
Theorem

- $\Phi = \{\varphi_k\}_{k \in \mathcal{K}}, \tilde{\Phi} = \{\tilde{\varphi}_k\}_{k \in \mathcal{K}}$ *biorthogonal pair of bases for H*
- Any $x \in H$ has *expansion coefficients*

$$\alpha_k = \langle x, \tilde{\varphi}_k \rangle, \quad k \in \mathcal{K}, \quad \text{or } \alpha = \tilde{\Phi}^* x$$

- *Synthesis:*
$$\begin{aligned} x &= \sum_{k \in \mathcal{K}} \langle x, \tilde{\varphi}_k \rangle \varphi_k \\ &= \Phi \alpha = \Phi \tilde{\Phi}^* x \end{aligned}$$

- Also
$$x = \sum_{k \in \mathcal{K}} \langle x, \varphi_k \rangle \tilde{\varphi}_k$$



Biorthogonal bases: Parseval's equality

Theorem

Parseval's equalities for biorthogonal pairs of bases

- $\Phi = \{\varphi_k\}_{k \in \mathcal{K}}, \tilde{\Phi} = \{\tilde{\varphi}_k\}_{k \in \mathcal{K}}$ biorthogonal pair of bases for H
- $$\begin{aligned}\|x\|^2 &= \sum_{k \in \mathcal{K}} \langle x, \varphi_k \rangle \langle x, \tilde{\varphi}_k \rangle^* \\ &= \langle \Phi^* x, \tilde{\Phi}^* x \rangle = \langle \tilde{\alpha}, \alpha \rangle\end{aligned}$$
- In general
$$\begin{aligned}\langle x, y \rangle &= \sum \langle x, \varphi_k \rangle \langle y, \tilde{\varphi}_k \rangle^* \\ &= \langle \Phi^* x, \tilde{\Phi}^* y \rangle = \langle \tilde{\alpha}, \beta \rangle\end{aligned}$$
- $\tilde{\Phi}^*$ is the inverse of Φ :
$$\begin{aligned}\Phi \tilde{\Phi}^* &= I \quad \text{on } H \text{ and} \\ \tilde{\Phi}^* \Phi &= I \quad \text{on } \ell^2(\mathcal{K})\end{aligned}$$

Gram matrix

- $G = \Phi^* \Phi$ is the **Gram matrix**

$$G_{ik} = \langle \varphi_k, \varphi_i \rangle \quad \text{for every } i, k \in \mathcal{K},$$

$$G = \begin{bmatrix} \vdots & \vdots & \vdots \\ \cdots & \langle \varphi_{-1}, \varphi_{-1} \rangle & \langle \varphi_0, \varphi_{-1} \rangle & \langle \varphi_1, \varphi_{-1} \rangle & \cdots \\ \cdots & \langle \varphi_{-1}, \varphi_0 \rangle & \boxed{\langle \varphi_0, \varphi_0 \rangle} & \langle \varphi_1, \varphi_0 \rangle & \cdots \\ \cdots & \langle \varphi_{-1}, \varphi_1 \rangle & \langle \varphi_0, \varphi_1 \rangle & \langle \varphi_1, \varphi_1 \rangle & \cdots \\ \vdots & \vdots & \vdots \end{bmatrix}$$

- Assume $x = \Phi\alpha$, $y = \Phi\beta$ then

$$\langle x, y \rangle = \langle \Phi\alpha, \Phi\beta \rangle = \langle \Phi^* \Phi\alpha, \beta \rangle = \langle G\alpha, \beta \rangle = \beta^* G \alpha$$

The inner product in H becomes an inner product in $\ell^2(\mathcal{K})$!

Dual basis

- How to compute the dual basis $\tilde{\Phi}$?

Theorem (Dual basis)

- Let $A = (\Phi^* \Phi)^{-1}$ be the inverse of the Gram matrix
- Given a Riesz basis $\Phi = \{\varphi_k\}_{k \in \mathcal{K}}$ for Hilbert space H , the set $\tilde{\Phi} = \{\tilde{\varphi}_k\}_{k \in \mathcal{K}}$ defined via

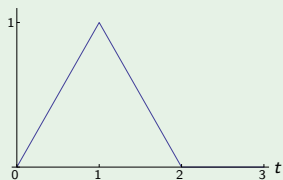
$$\tilde{\varphi}_k = \sum_{\ell \in \mathcal{K}} a_{\ell, k} \varphi_\ell, \quad \text{for each } k \in \mathcal{K},$$

$$\tilde{\Phi} = \Phi A = \Phi (\Phi^* \Phi)^{-1},$$

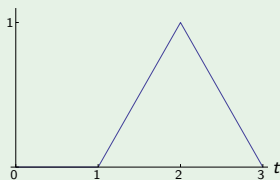
is a basis for H , called the **dual basis**, and the sets Φ and $\tilde{\Phi}$ are a biorthogonal pair of bases

Dual basis of periodic hat functions

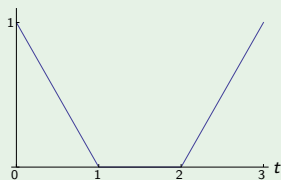
Example



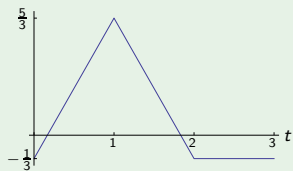
$\varphi_1(t)$



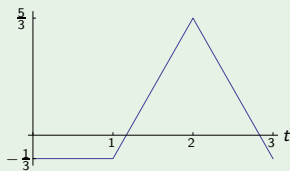
$\varphi_2(t)$



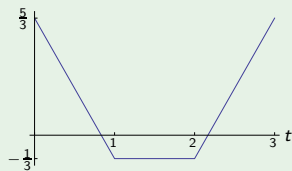
$\varphi_3(t)$



$\tilde{\varphi}_1(t)$



$\tilde{\varphi}_2(t)$



$\tilde{\varphi}_3(t)$

Dual basis of periodic hat functions: Computing the dual basis

- Consider the functions in $\mathcal{L}^2([0, 3])$ such that $\varphi_0 = \begin{cases} t, & \text{for } t \in [0, 1); \\ 2 - t & \text{for } t \in (1, 2]; \\ 0 & \text{for } t \in (2, 3] \end{cases}$

and their circular shifts by 1.

- Let $\Phi = \{\varphi_1, \varphi_2, \varphi_3\}$ be the basis for $\text{span}\{\varphi_1, \varphi_2, \varphi_3\} =$ subspace of functions x satisfying $x(0) = x(3)$ and are piecewise linear on $[0, 3]$ with breakpoints at 1 and 2

- $G = \begin{bmatrix} 2/3 & 1/6 & 1/6 \\ 1/6 & 2/3 & 1/6 \\ 1/6 & 1/6 & 2/3 \end{bmatrix}$

- We find a dual basis for Φ by using $\tilde{\Phi} = \Phi A = \Phi G^{-1}$ where

$$G^{-1} = \begin{bmatrix} 5/3 & -1/3 & -1/3 \\ -1/3 & 5/3 & -1/3 \\ -1/3 & -1/3 & 5/3 \end{bmatrix}$$

Dual bases and projection

- How to compute the projection?

Theorem

- Given sets $\Phi_{\mathcal{I}} = \{\varphi_k\}_{k \in \mathcal{I}} \subset H$ and $\tilde{\Phi}_{\mathcal{I}} = \{\tilde{\varphi}_k\}_{k \in \mathcal{I}} \subset H$ satisfying

$$\langle \varphi_i, \tilde{\varphi}_k \rangle = \delta_{i-k} \quad \text{for every } i, k \in \mathcal{I},$$

- Then, for any x in H ,

$$P_{\mathcal{I}} x = \sum_{k \in \mathcal{I}} \langle x, \tilde{\varphi}_k \rangle \varphi_k = \Phi_{\mathcal{I}} \tilde{\Phi}_{\mathcal{I}}^* x$$

is an **oblique projection** of x onto $S_{\mathcal{I}} = \overline{\text{span}}(\{\varphi_k\}_{k \in \mathcal{I}})$.

- The residual satisfies $x - P_{\mathcal{I}} x \perp \tilde{S}_{\mathcal{I}}$, where $\tilde{S}_{\mathcal{I}} = \overline{\text{span}}(\{\tilde{\varphi}_k\}_{k \in \mathcal{I}})$.

Projection onto a subspace

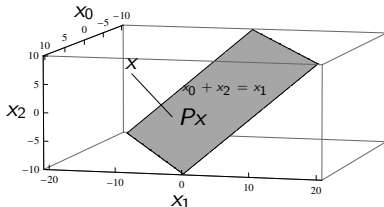
Example (Projection onto a subspace)

Let $A = \begin{bmatrix} 1/2 & 1/2 & -1/2 \\ -1/2 & 1/2 & 1/2 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 1 \end{bmatrix}$.

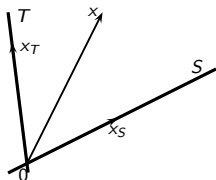
Since A is a left inverse of B , then

$$P = BA = \begin{bmatrix} 1/2 & 1/2 & -1/2 \\ 0 & 1 & 0 \\ -1/2 & 1/2 & 1/2 \end{bmatrix} \text{ is a projection operator.}$$

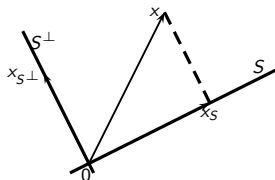
We can verify $P^2 = P$. Does $P^* = P$?



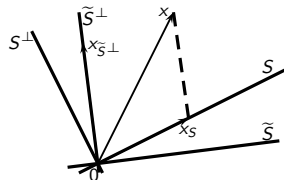
Direct sum decomposition



(a)
Direct sum in
a vector space
 $V = S \oplus T$



(b)
Orthogonal projection
on the subspace S
 $x = x_S + x_{S^\perp}$



(c)
Oblique projection
on the subspace S
 $x = x_S + x_{\tilde{S}^\perp}$

Best approximation & the normal equations

Theorem (Normal equations)

- $x \in H$ and $\{\phi_k\}_{k \in \mathcal{I}}$ a Riesz basis for a closed subspace S
- The closest vector to x in S is

$$\hat{x} = \sum_{k \in \mathcal{I}} \beta_k \phi_k = \Phi \beta$$

where β is the unique solution to:

$$\Phi^* \Phi \beta = \Phi^* x \quad \text{or}$$

$$\sum_{k \in \mathcal{I}} \beta_k \langle \phi_k, \phi_i \rangle = \langle x, \phi_i \rangle \quad \text{for all } i \in \mathcal{I}$$

Normal equations

- $\hat{x} = \Phi(\Phi^* \Phi)^{-1} \Phi^* x = Px$

P is an orthogonal projection

- We can verify that $P^2 = \Phi \underbrace{(\Phi^* \Phi)^{-1} \Phi^* \Phi}_{I} (\Phi^* \Phi)^{-1} \Phi^* = P$ and that $P^* = P$

Frames: Overcomplete representations

Definition

Frame

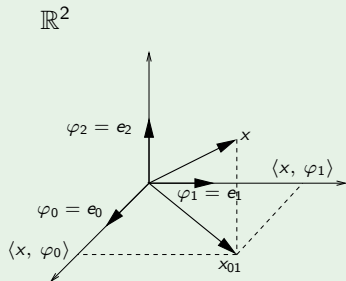
- $\Phi = \{\varphi_k\}_{k \in \mathcal{J}} \subset H$ is a **frame** for the Hilbert space H when there exist $0 < \lambda_{\min} \leq \lambda_{\max} < \infty$ s.t. for any $x \in H$

$$\lambda_{\min} \|x\|^2 \leq \sum_{k \in \mathcal{J}} |\langle x, \varphi_k \rangle|^2 \leq \lambda_{\max} \|x\|^2$$

- $\lambda_{\min}, \lambda_{\max}$: **frame bounds**
- **Synthesis** operator: $\Phi : \ell^2(\mathcal{J}) \rightarrow H, \quad \Phi \alpha = \sum_{k \in \mathcal{J}} \alpha_k \varphi_k$
- **Analysis** operator: $\Phi^* : H \rightarrow \ell^2(\mathcal{J}), \quad (\Phi^* x)_k = \langle x, \varphi_k \rangle, \quad k \in \mathcal{J}$
- $\lambda_{\min} I \leq \Phi \Phi^* \leq \lambda_{\max} I$

Frames

Example



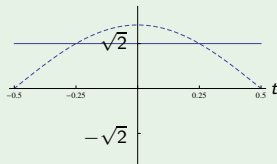
- Given $\Phi = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & -1 \end{bmatrix}$
- $\{\varphi_0, \varphi_1, \varphi_2\}$ is a frame for \mathbb{R}^2
- λ_{\min} and λ_{\max} computed as **smallest** and **largest eigenvalues** of $\Phi\Phi^*$

$$\bullet \Phi\Phi^* = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} = \underbrace{\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}}_V \underbrace{\begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix}}_\Lambda \underbrace{\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}}_{V^{-1}}$$

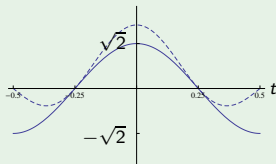
$$\bullet \lambda_{\min} = 1, \lambda_{\max} = 3$$

Example: Frame of cosine functions

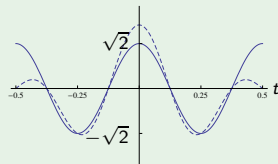
Example



(a) $\varphi_0(t)$ and $\psi_0(t)$



(b) $\varphi_1(t)$ and $\psi_1(t)$

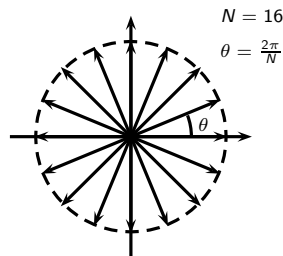
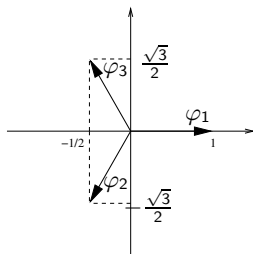


(c) $\varphi_2(t)$ and $\psi_2(t)$

Tight frames

Definition

- $\Phi = \{\varphi_k\}_{k \in \mathcal{J}} \subset H$ is a tight frame or λ -tight frame for H when $\lambda_{\min} = \lambda_{\max} = \lambda$
- Counterpart of orthonormal basis
- When $\lambda = 1$: Parseval tight frame



Tight frames expansions

Theorem (Tight frame expansions)

- $\Phi = \{\varphi_k\}_{k \in \mathcal{J}}$ a 1-tight frame for H , $x \in H$

- *Expansion coefficients* of x w.r.t. Φ :

$$\alpha_k = \langle x, \varphi_k \rangle \quad k \in \mathcal{J}, \text{ or } \alpha = \Phi^* x$$

- *Expansion* $x = \sum_{k \in \mathcal{J}} \langle x, \varphi_k \rangle \varphi_k = \Phi \Phi^* x$

- $\Phi \Phi^* = I$ but (in general) $\Phi^* \Phi \neq I$

- Note: similar to orthonormal basis, but \mathcal{J} is **overcomplete** and vectors Φ are linearly **dependent**

Tight frames Parseval's equality

Theorem (Parseval's equalities for 1-tight frames)

- $\Phi = \{\varphi_k\}_{k \in \mathcal{J}}$ a 1-tight frame for H
- $\|x\|^2 = \sum_{k \in \mathcal{J}} |\langle x, \varphi_k \rangle|^2 = \|\Phi^* x\|^2 = \|x\|^2$
- In general:
$$\begin{aligned} \langle x, y \rangle &= \sum_{k \in \mathcal{J}} \langle x, \varphi_k \rangle \langle y, \varphi_k \rangle^* \\ &= \langle \Phi^* x, \Phi^* y \rangle = \langle \alpha, \beta \rangle \end{aligned}$$

General frames I

Definition (Dual pair of frames)

- Dual frame pairs and **expansion**

$\Phi = \{\varphi_k\}_{k \in \mathcal{J}} \in H$, $\tilde{\Phi} = \{\tilde{\varphi}_k\}_{k \in \mathcal{J}} \in H$ form a dual pair of frames when:

- 1 Each is a frame for H
- 2 for any x in H ,

$$x = \sum_{k \in \mathcal{K}} \langle x, \tilde{\varphi}_k \rangle \varphi_k = \Phi \tilde{\Phi}^* x$$

General frames II

Example

- Let $\Phi = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & -1 \end{bmatrix}$ be the frame synthesis operator for \mathbb{R}^2 of rank 2
- Φ has infinitely many right inverses
- Examples: $\tilde{\Phi} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$, $\tilde{\Phi} = \begin{bmatrix} 0 & -1 & -1 \\ 1 & 2 & 1 \end{bmatrix}$ and
$$\tilde{\Phi} = \begin{bmatrix} 2/3 & -1/3 & -1/3 \\ -1/3 & 2/3 & -1/3 \end{bmatrix}$$
- Check that $\Phi\tilde{\Phi}^* = I_2$

General frames: Operators I

- **Inner product:** Assume $x = \Phi\alpha, y = \Phi\beta$, $\langle x, y \rangle_H = \beta^* G \alpha$

- **Oblique projection:** $P = \tilde{\Phi}^* \Phi$

$$P^2 = (\tilde{\Phi}^* \Phi)(\tilde{\Phi}^* \Phi) = \tilde{\Phi}^* \underbrace{(\Phi \tilde{\Phi}^*)}_I \Phi = \tilde{\Phi}^* \Phi = P$$

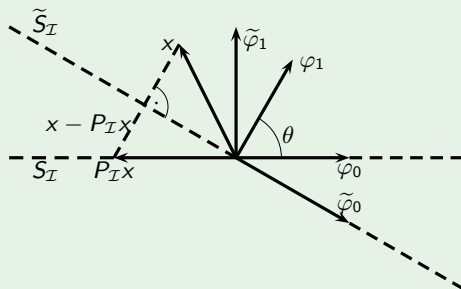
Does $P^* = P$?

- **Canonical dual:** $\tilde{\Phi} = (\Phi\Phi^*)^{-1}\Phi$

$P = \tilde{\Phi}^* \Phi$ is an orthogonal projection

General frames: Operators II

Example



- $\Phi_I = \begin{bmatrix} 1 & 0 \end{bmatrix}^T$, $\tilde{\Phi}_I = \begin{bmatrix} 1 & -\cot \theta \end{bmatrix}^T$

- $P_I x = \Phi_I \tilde{\Phi}_I^* x$

Change of basis: Orthonormal basis

- How are the expansion coefficients in two **orthonormal** bases related?
- Assume $x = \Phi\alpha = \Psi\beta$. Then $\Psi^*\Phi\alpha = \Psi^*\Psi\beta = \beta$
- The change of basis from Φ to Ψ that maps α to β is the operator

$$C_{\Phi,\Psi} : \ell^2(\mathcal{K}) \rightarrow \ell^2(\mathcal{K}) \text{ s.t. } C_{\Phi,\Psi} = \Psi^*\Phi \text{ (or } \beta = \Psi^*\Phi\alpha)$$

- As a matrix

$$C_{\Phi,\Psi} = \begin{bmatrix} \vdots & \vdots & \vdots \\ \cdots & \langle \varphi_{-1}, \psi_{-1} \rangle & \langle \varphi_0, \psi_{-1} \rangle & \langle \varphi_1, \psi_{-1} \rangle & \cdots \\ \cdots & \langle \varphi_{-1}, \psi_0 \rangle & \boxed{\langle \varphi_0, \psi_0 \rangle} & \langle \varphi_1, \psi_0 \rangle & \cdots \\ \cdots & \langle \varphi_{-1}, \psi_1 \rangle & \langle \varphi_0, \psi_1 \rangle & \langle \varphi_1, \psi_1 \rangle & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix}$$

Change of basis: Biorthogonal basis

- How are the expansion coefficients in two **biorthogonal** bases related?
- Assume $x = \Phi\alpha = \Psi\beta$. Then $\Psi^{-1}\Phi\alpha = \Psi^{-1}\Psi\beta = \beta$ and $\beta = \Psi^{-1}\Phi\alpha$
- The change of basis from Φ to Ψ that maps α to β is the operator

$$C_{\Phi,\Psi} : \ell^2(\mathcal{K}) \rightarrow \ell^2(\mathcal{K}) \text{ s.t. } C_{\Phi,\Psi} = \Psi^{-1}\Phi$$

- As a matrix

$$C_{\Phi,\Psi} = \begin{bmatrix} \vdots & \vdots & \vdots \\ \cdots & \langle \varphi_{-1}, \tilde{\psi}_{-1} \rangle & \langle \varphi_0, \tilde{\psi}_{-1} \rangle & \langle \varphi_1, \tilde{\psi}_{-1} \rangle & \cdots \\ \cdots & \langle \varphi_{-1}, \tilde{\psi}_0 \rangle & \boxed{\langle \varphi_0, \tilde{\psi}_0 \rangle} & \langle \varphi_1, \tilde{\psi}_0 \rangle & \cdots \\ \cdots & \langle \varphi_{-1}, \tilde{\psi}_1 \rangle & \langle \varphi_0, \tilde{\psi}_1 \rangle & \langle \varphi_1, \tilde{\psi}_1 \rangle & \cdots \\ \vdots & \vdots & \vdots \end{bmatrix}$$

Matrix representation: Orthonormal basis

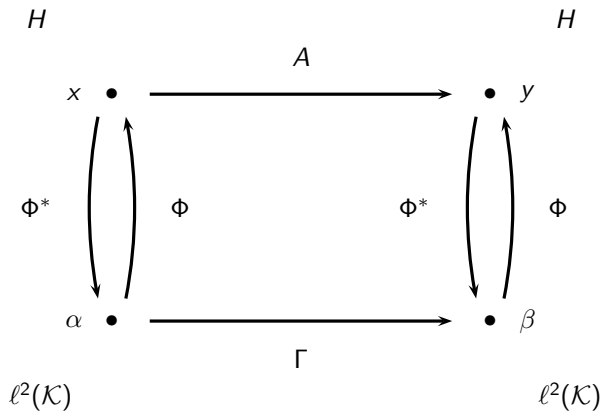
- How are the expansion coefficients of x and y related when $A : H \rightarrow H$, s.t. $y = Ax$?
 - $\{\varphi_k\}_{k \in \mathcal{K}}$ orthonormal basis of H
 - $x = \Phi\alpha, y = \Phi\beta$
- A matrix representation allows A to be computed directly on coefficient sequences

$$\Gamma : \ell^2(\mathcal{K}) \rightarrow \ell^2(\mathcal{K}) \text{ s.t. } \beta = \Gamma\alpha$$

- As a matrix

$$\Gamma = \begin{bmatrix} \vdots & \vdots & \vdots & \\ \cdots & \langle A\varphi_{-1}, \varphi_{-1} \rangle & \langle A\varphi_0, \varphi_{-1} \rangle & \langle A\varphi_1, \varphi_{-1} \rangle & \cdots \\ \cdots & \langle A\varphi_{-1}, \varphi_0 \rangle & \boxed{\langle A\varphi_0, \varphi_0 \rangle} & \langle A\varphi_1, \varphi_0 \rangle & \cdots \\ \cdots & \langle A\varphi_{-1}, \varphi_1 \rangle & \langle A\varphi_0, \varphi_1 \rangle & \langle A\varphi_1, \varphi_1 \rangle & \cdots \\ \vdots & \vdots & \vdots & \vdots & \end{bmatrix}$$

Matrix representation: Orthonormal basis



Matrix representation: Orthonormal basis

- How are the expansion coefficients of x and y related when $A : H_0 \rightarrow H_1$, s.t. $y = Ax$?
 - Φ orthonormal basis of H_0
 - Ψ orthonormal basis of H_1
 - $x = \Phi\alpha, y = \Psi\beta$
- A matrix representation allows A to be computed directly on coefficient sequences

$$\Gamma : \ell^2(\mathcal{K}) \rightarrow \ell^2(\mathcal{K}) \text{ s.t. } \beta = \Gamma\alpha$$

- As a matrix

$$\Gamma = \begin{bmatrix} \vdots & \vdots & \vdots & \vdots \\ \cdots & \langle A\varphi_{-1}, \psi_{-1} \rangle & \langle A\varphi_0, \psi_{-1} \rangle & \langle A\varphi_1, \psi_{-1} \rangle & \cdots \\ \cdots & \langle A\varphi_{-1}, \psi_0 \rangle & \boxed{\langle A\varphi_0, \psi_0 \rangle} & \langle A\varphi_1, \psi_0 \rangle & \cdots \\ \cdots & \langle A\varphi_{-1}, \psi_1 \rangle & \langle A\varphi_0, \psi_1 \rangle & \langle A\varphi_1, \psi_1 \rangle & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix}$$

Example: Averaging Operator I

Example

- Let $A : H_0 \rightarrow H_1$,
$$y(t) = Ax(t) = \frac{1}{2} \int_{2\ell}^{2(\ell+1)} x(\tau) d\tau$$

for $2\ell \leq t < 2(\ell+1)$, $\ell \in \mathbb{Z}$

H_0 : space of piecewise-constant, finite-energy functions with breakpoints at integers

H_1 : space of piecewise-constant, finite-energy functions with breakpoints at even integers.

- Given $\chi_I(t) = \begin{cases} 1, & \text{for } t \in I; \\ 0, & \text{otherwise} \end{cases}$,

$$\begin{aligned} \text{Let } \Phi &= \{\varphi_k(t)\}_{k \in \mathbb{Z}} = \{\chi_{[k, k+1)}(t)\}_{k \in \mathbb{Z}}, \\ \Psi &= \{\psi_i(t)\}_{i \in \mathbb{Z}} = \left\{ \frac{1}{\sqrt{2}} \chi_{[2i, 2(i+1))}(t) \right\}_{i \in \mathbb{Z}}, \end{aligned}$$

be orthogonal bases for H_0, H_1 respectively

Example: Averaging Operator II

Example (Cont.)

- $A\varphi_0(t) = \frac{1}{2} \chi_{[0,2)}(t) \Rightarrow \langle A\varphi_0, \psi_0 \rangle = \int_0^2 \frac{1}{2} \frac{1}{\sqrt{2}} d\tau = \frac{1}{\sqrt{2}}$

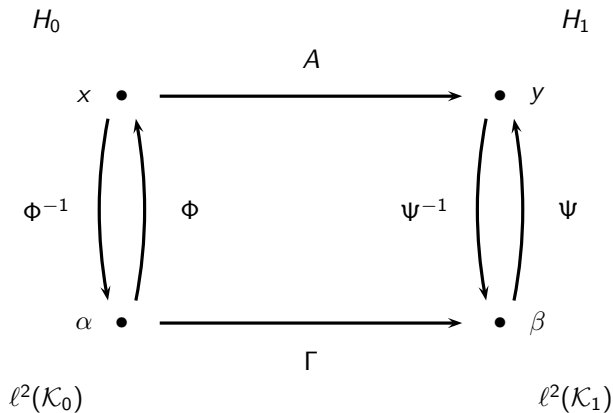
- Then $\Gamma = \frac{1}{\sqrt{2}} \begin{bmatrix} \ddots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \\ \dots & 1 & 1 & 0 & 0 & 0 & 0 & \dots \\ \dots & 0 & 0 & \boxed{1} & 1 & 0 & 0 & \dots \\ \dots & 0 & 0 & 0 & 0 & 1 & 1 & \dots \\ & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix}.$

Matrix representation: Biorthogonal basis

- How are the expansion coefficients of x and y related when $A : H_0 \rightarrow H_1$, s.t. $y = Ax$?
 - $\Phi, \tilde{\Phi}$ biorthogonal pair of bases of H_0
 - $\Psi, \tilde{\Psi}$ biorthogonal pair of bases of H_1
- A matrix representation is similar to the orthogonal case, but involves the dual bases
- As a matrix

$$\Gamma = \begin{bmatrix} \vdots & \vdots & \vdots & \\ \cdots & \langle A\varphi_{-1}, \tilde{\psi}_{-1} \rangle & \langle A\varphi_0, \tilde{\psi}_{-1} \rangle & \langle A\varphi_1, \tilde{\psi}_{-1} \rangle & \cdots \\ \cdots & \langle A\varphi_{-1}, \tilde{\psi}_0 \rangle & \boxed{\langle A\varphi_0, \tilde{\psi}_0 \rangle} & \langle A\varphi_1, \tilde{\psi}_0 \rangle & \cdots \\ \cdots & \langle A\varphi_{-1}, \tilde{\psi}_1 \rangle & \langle A\varphi_0, \tilde{\psi}_1 \rangle & \langle A\varphi_1, \tilde{\psi}_1 \rangle & \cdots \\ \vdots & \vdots & \vdots & \end{bmatrix}$$

Matrix representation: Biorthogonal basis



Example: Derivative Operator I

Example

- Let $A : H_0 \rightarrow H_1$ be the derivative operator

H_0 : space of piecewise-linear, continuous, finite-energy functions with breakpoints at integers

H_1 : space of piecewise-constant, finite-energy functions with breakpoints at integers.

- Let $\Phi = \{\varphi_k(t)\}_{k \in \mathbb{Z}} = \{\varphi(t - k)\}_{k \in \mathbb{Z}}$, $\varphi(t) = \begin{cases} 1 - |t|, & |t| < 1; \\ 0, & \text{otherwise} \end{cases}$
- Let $\Psi = \{\psi_i(t)\}_{i \in \mathbb{Z}} = \{\chi_{[i, i+1)}(t)\}_{i \in \mathbb{Z}}$.

Example: Derivative Operator II

Example (Cont.)

- We evaluate $\langle A\varphi_k, \tilde{\psi}_i \rangle$ for all k and i .

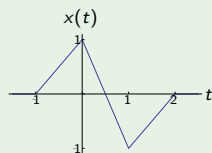
$$A\varphi(t) = \varphi'(t) = \begin{cases} 1, & \text{for } -1 < t < 0; \\ -1, & \text{for } 0 < t < 1; \\ 0, & \text{for } |t| > 1, \end{cases}$$

- Then $\langle A\varphi_0, \tilde{\psi}_i \rangle = \begin{cases} 1, & \text{for } i = -1; \\ -1, & \text{for } i = 0; \\ 0, & \text{otherwise.} \end{cases}$ and $\langle A\varphi_k, \tilde{\psi}_i \rangle = \begin{cases} 1, & \text{for } i = k - 1; \\ -1, & \text{for } i = k; \\ 0, & \text{otherwise.} \end{cases}$

$$\text{and } \Gamma = \begin{bmatrix} \ddots & \ddots & \ddots & \vdots & \vdots & \vdots & \vdots & \\ \cdots & 0 & -1 & 1 & 0 & 0 & 0 & \cdots \\ \cdots & 0 & 0 & -1 & 1 & 0 & 0 & \cdots \\ \cdots & 0 & 0 & 0 & -1 & 1 & 0 & \cdots \\ & \vdots & \vdots & \vdots & \vdots & \ddots & \ddots & \ddots \end{bmatrix}.$$

Example: Derivative Operator III

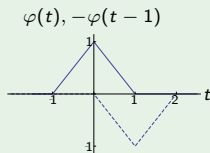
Example (Cont.)



(a)

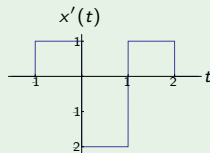
Original function

$x(t)$



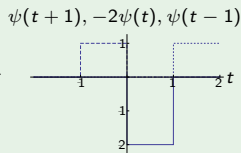
(b)

its decomposition
in the basis for H_0



(c)

Derivative function
 $x'(t)$



(d)

its decomposition
in the basis for H_1

- Expansion in Φ :
$$x(t) = \varphi(t) - \varphi(t-1)$$
$$\alpha = \left[\dots \quad 0 \quad \boxed{1} \quad -1 \quad 0 \quad 0 \quad \dots \right]^T$$
- Expansion of the derivative in Ψ :
$$x'(t) = \psi(t+1) - 2\psi(t) + \psi(t-1),$$
$$\beta = \left[\dots \quad 0 \quad 1 \quad -2 \quad 1 \quad 0 \quad \dots \right]^T$$
- Check: $\beta = \Gamma \alpha$

Example: Polynomial approximation

- Hilbert space $P_N[-1, 1] \subset \mathcal{L}^2[-1, 1]$
- a self-evident, naive basis: $\mathbf{s}^{(k)} = t^k$, $k = 0, 1, \dots, N - 1$
- naive basis is not orthonormal
- goal: approximate $\mathbf{x} = \sin t$ over $P_3[-1, 1]$

Building an orthonormal basis

Gram-Schmidt orthonormalization procedure:

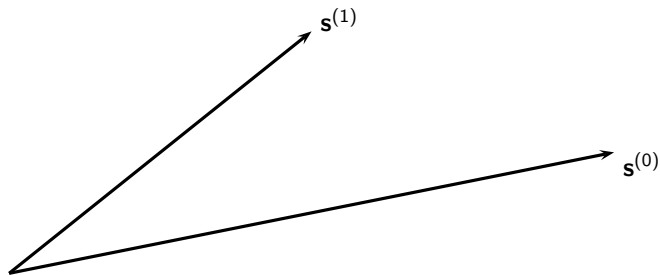
$$\begin{array}{ccc} \{\mathbf{s}^{(k)}\} & \longrightarrow & \{\mathbf{u}^{(k)}\} \\ \text{original set} & & \text{orthonormal set} \end{array}$$

Algorithmic procedure: at each step k

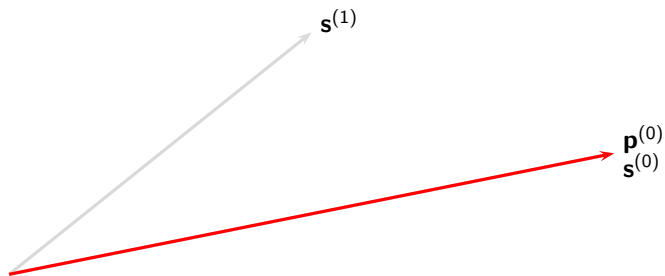
1 $\mathbf{p}^{(k)} = \mathbf{s}^{(k)} - \sum_{n=0}^{k-1} \langle \mathbf{u}^{(n)}, \mathbf{s}^{(k)} \rangle \mathbf{u}^{(n)}$

2 $\mathbf{u}^{(k)} = \mathbf{p}^{(k)} / \|\mathbf{p}^{(k)}\|$

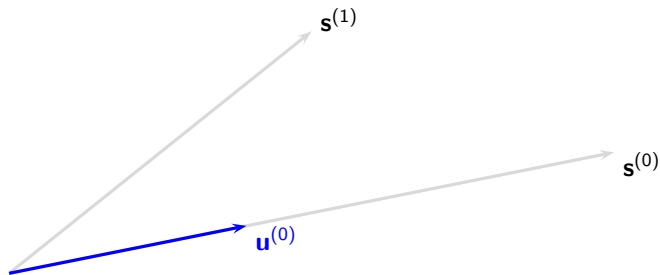
Least Squares Approximation



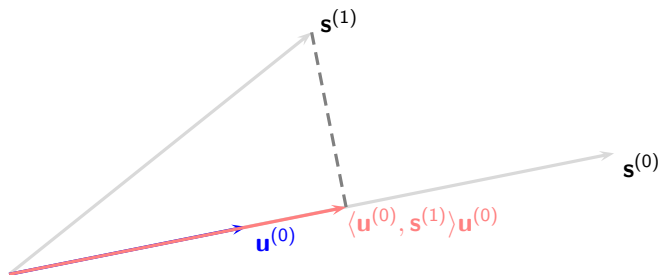
Least Squares Approximation



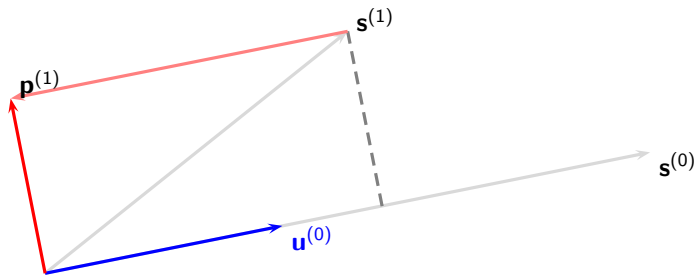
Least Squares Approximation



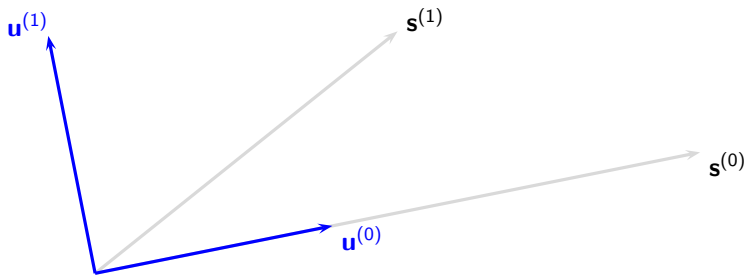
Least Squares Approximation



Least Squares Approximation



Least Squares Approximation



Building an orthonormal basis

Gram-Schmidt orthonormalization of the naive basis: $\{\mathbf{s}^{(k)}\} \rightarrow \{\mathbf{u}^{(k)}\}$

- $\mathbf{s}^{(0)} = 1$

- $\mathbf{p}^{(0)} = \mathbf{s}^{(0)} = 1$

- $\|\mathbf{p}^{(0)}\|^2 = 2$

- $\mathbf{u}^{(0)} = \mathbf{p}^{(0)} / \|\mathbf{p}^{(0)}\| = \sqrt{1/2}$

- $\mathbf{s}^{(1)} = t$

- $\langle \mathbf{u}^{(0)}, \mathbf{s}^{(1)} \rangle = \int_{-1}^1 t / \sqrt{2} = 0$

- $\mathbf{p}^{(1)} = \mathbf{s}^{(1)} = t$

- $\|\mathbf{p}^{(1)}\|^2 = 2/3$

- $\mathbf{u}^{(1)} = \sqrt{3/2} t$

- $\mathbf{s}^{(2)} = t^2$

- $\langle \mathbf{u}^{(0)}, \mathbf{s}^{(2)} \rangle = \int_{-1}^1 t^2 / \sqrt{2} = 2/3\sqrt{2}$

- $\langle \mathbf{u}^{(1)}, \mathbf{s}^{(2)} \rangle = \int_{-1}^1 t^3 / \sqrt{2} = 0$

- $\mathbf{p}^{(2)} = \mathbf{s}^{(2)} - (2/3\sqrt{2})\mathbf{u}^{(0)} = t^2 - 1/3$

- $\|\mathbf{p}^{(2)}\|^2 = 8/45$

- $\mathbf{u}^{(2)} = \sqrt{5/8}(3t^2 - 1)$

Legendre polynomials

An orthonormal basis for $P_N([-1, 1])$

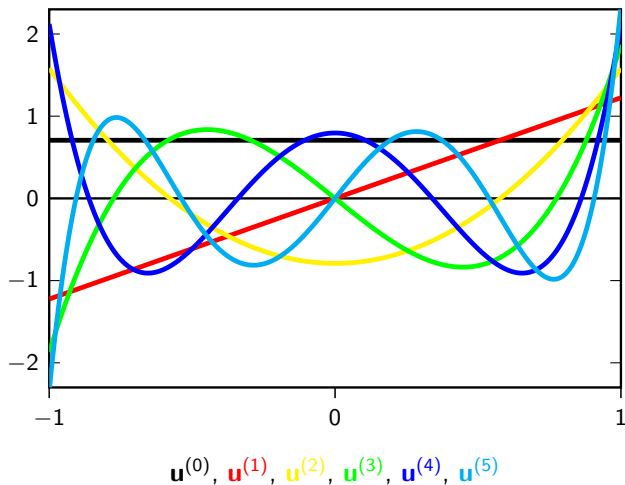
$$\mathbf{u}^{(0)} = \sqrt{1/2}$$

$$\mathbf{u}^{(1)} = \sqrt{3/2} t$$

$$\mathbf{u}^{(2)} = \sqrt{5/8}(3t^2 - 1)$$

$$\mathbf{u}^{(3)} = \dots$$

Legendre Polynomials



Orthogonal projection over $P_3[-1, 1]$

$$\alpha_k = \langle \mathbf{u}^{(k)}, \mathbf{x} \rangle = \int_{-1}^1 u_k(t) \sin t \, dt$$

- $\alpha_0 = \langle \sqrt{1/2}, \sin t \rangle = 0$
- $\alpha_1 = \langle \sqrt{3/2} t, \sin t \rangle \approx 0.7377$
- $\alpha_2 = \langle \sqrt{5/8}(3t^2 - 1), \sin t \rangle = 0$

Approximation

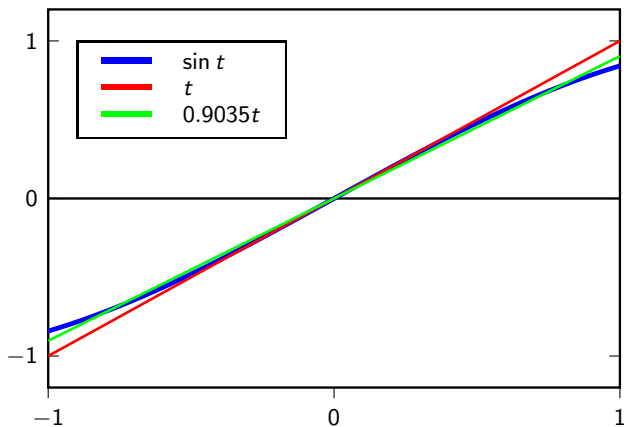
Using the orthogonal projection over $P_3[-1, 1]$:

$$\sin t = \alpha_1 \mathbf{u}^{(1)} \approx 0.9035 t$$

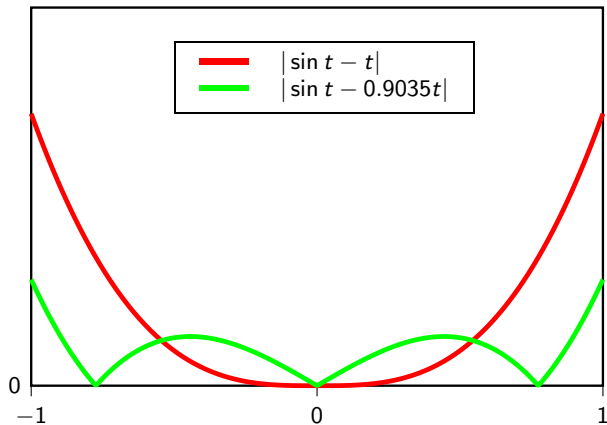
Using Taylor's series:

$$\sin t \approx t$$

Sine approximation



Approximation error



Orthogonal projection over $P_3[-1, 1]$:

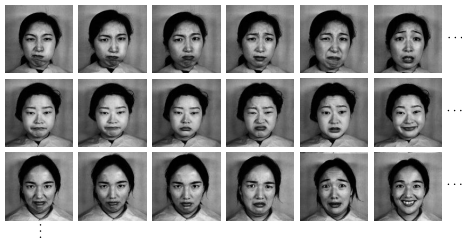
$$\|\sin t - \alpha_1 t\| \approx 0.0337$$

Taylor series:

$$\|\sin t - t\| \approx 0.0857$$

Example: Eigenfaces

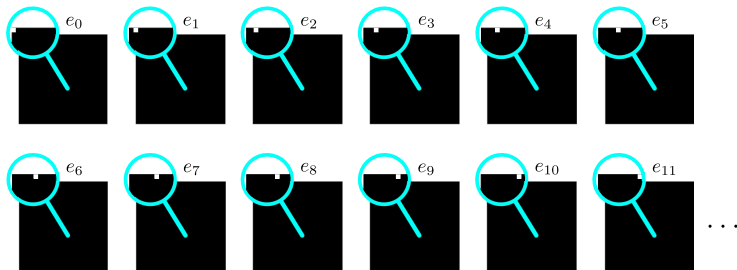
- Images of ten women with about 20 facial expressions (213 images in total)*
- Images of size 256×256 : dimensionality *too high* (65536)
- Goal: represent images in *few dimensions*



*Michael J. Lyons, Miyuki Kamachi, Jiro Gyoba. Japanese Female Facial Expressions (JAFFE), Database of digital images (1997).
<http://www.kasrl.org/jaffe.html>

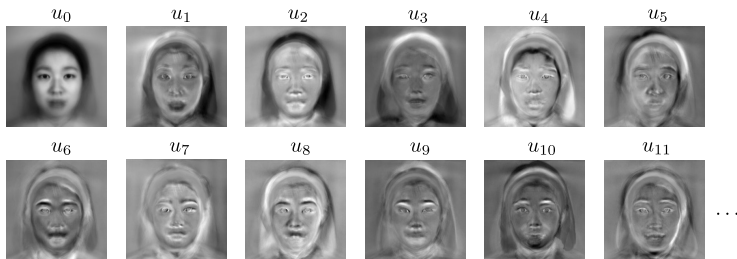
Example: Eigenfaces

Canonical (usual) basis



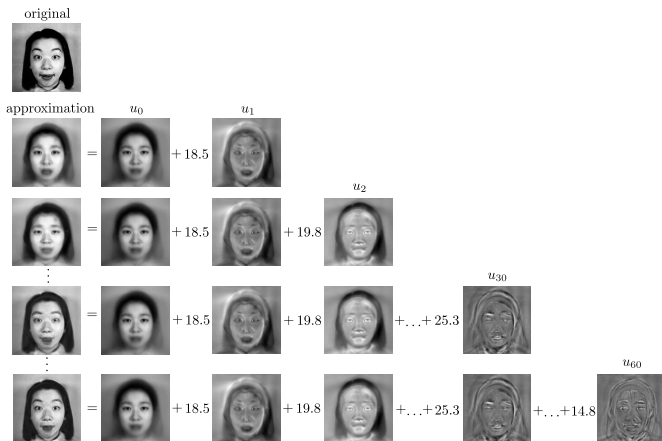
Example: Eigenfaces

Eigenface basis



Example: Eigenfaces

Reconstruction with eigenfaces



Summary

- Bases
 - Bases span a space S
 - Well behaved bases are Riesz bases (conditioning)
 - Best behaved bases are orthonormal bases
 - Biorthogonal bases have dual bases
 - Parseval's equalities
 - Approximations and normal equations
- Matrix representation of operators
 - Change of basis (orthonormal, biorthogonal)
 - Matrix representations of operators