

Math 6610: Analysis of Numerical Methods, I Orthogonal, Projection, and Permutation matrices

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Fall 2026

Resources: Trefethen and Bau [1997](#), Lectures 2, 6
Atkinson [1989](#), Sections 7.1
Salgado and Wise [2022](#), Sections 1.1, 1.2, 5.2

Unitary and orthogonal matrices

A *very* special and important class of matrices:

Definition (Unitary/orthogonal matrices)

A matrix $\mathbf{U} \in \mathbb{C}^{n \times n}$ is **unitary** if $\mathbf{U}^* \mathbf{U} = \mathbf{I}_n$

A matrix $\mathbf{U} \in \mathbb{R}^{n \times n}$ is **orthogonal** if $\mathbf{U}^T \mathbf{U} = \mathbf{I}_n$

A unitary matrix is one whose columns are an orthonormal basis for $\mathbb{C}^n$ under the $\ell^2(\mathbb{C}^n)$ inner product.

$$\mathbf{U} = \begin{pmatrix} | & & | \\ u_1 & \cdots & u_n \\ | & & | \end{pmatrix} \quad \mathbf{U}^* \mathbf{U} = \begin{pmatrix} \langle u_1, u_1 \rangle & \cdots & \langle u_1, u_n \rangle \\ \vdots & \ddots & \vdots \\ \langle u_n, u_1 \rangle & \cdots & \langle u_n, u_n \rangle \end{pmatrix}$$

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Theorem

Let $\mathbf{U} \in \mathbb{C}^{n \times n}$ be a unitary matrix. Then

- $\mathbf{U}^{-1} = \mathbf{U}^*$ ← Because $(\mathbf{U}^*)\mathbf{U} = \mathbf{I}$

- $\mathbf{U}\mathbf{U}^* = \mathbf{I}$

- $\|\mathbf{U}\mathbf{x}\|_2 = \|\mathbf{x}\|_2$ for all $\mathbf{x} \in \mathbb{C}^n$.

(i) $\mathbf{U}\mathbf{x} = \sum_{j=1}^n x_j \mathbf{u}_j \rightarrow \text{Pythagorean Thm}$

(ii) $\|\mathbf{U}\mathbf{x}\|_2^2 = \langle \mathbf{U}\mathbf{x}, \mathbf{U}\mathbf{x} \rangle = (\mathbf{U}\mathbf{x})^* \mathbf{U}\mathbf{x} = \mathbf{x}^* \mathbf{U}^* \mathbf{U} \mathbf{x} = \|\mathbf{x}\|_2^2$

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ℓ^2

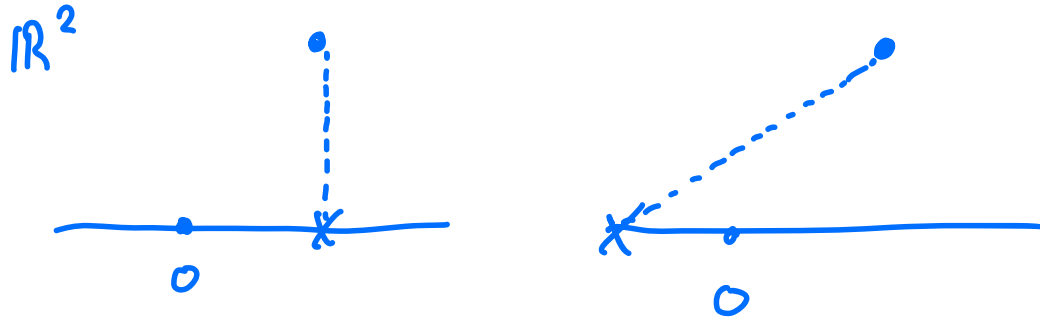
This last property in particular shows: unitary matrices are *isometries*, i.e., are ℓ^2 norm-preserving.

Unitary matrices have nice geometric interpretations: they are precisely rigid rotations and/or reflections.

Projections and projection matrices: Geometric construction

With $\mathbb{C}^n$ the ambient space, we want to define square matrix “projections”. Informally, we want projections to

- Act like the identity on some subspace (the range)
- Annihilate components in another subspace (the kernel)



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In order to be well-defined, we need an additional condition.

Definition (Geometric definition of projection matrices)

A matrix $\mathbf{P} \in \mathbb{C}^{n \times n}$ is a projection matrix if

1. $\mathbf{P}\mathbf{v} = \mathbf{v}$ for all $\mathbf{v} \in V = \text{range}(\mathbf{P})$
2. $\mathbf{P}\mathbf{w} = \mathbf{0}$ for all $\mathbf{w} \in W = \text{ker}(\mathbf{P})$
3. $\text{range}(\mathbf{P}) \oplus \text{ker}(\mathbf{P}) = \mathbb{C}^n$ and $\text{range}(\mathbf{P}) \cap \text{ker} \mathbf{P} = \{0\}$. $\Rightarrow n = \dim V + \dim W$

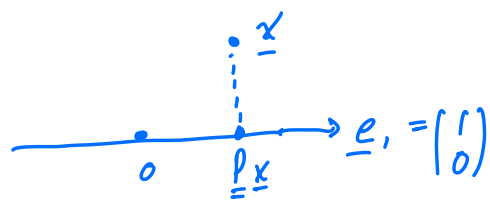
We typically say that $\mathbf{P}$ projects *onto* V , and projects *along* W .

Ex: $\mathbf{P} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$

$$V = \text{range}(\mathbf{P}) = \text{span} \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right\}, \quad W = \text{ker}(\mathbf{P}) = \text{span} \left\{ \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\}$$

$$2 = \dim V + \dim W$$

$$V \cap W = \left\{ \begin{pmatrix} 0 \\ 0 \end{pmatrix} \right\}, \quad \mathbb{C}^2 = V \oplus W$$

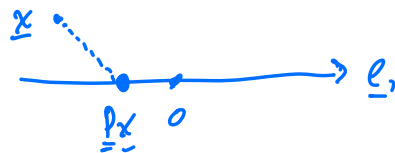


$$\underline{\text{Ex}}: P = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}$$

$$V = \text{range}(P) = \text{span} \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right\}, \quad W = \ker(P) = \text{span} \left\{ \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right\}$$

$$V \oplus W = \mathbb{C}^2, \quad V \cap W = \{0\}$$

$$\nwarrow \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$



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We typically say that $\mathbf{P}$ projects *onto* V , and projects *along* W .

This last condition is *necessary*:

- A projector $\mathbf{P}$ in $\mathbb{R}^3$ that projects onto $\text{span}\{(1, 0, 0)\}$ along $\text{span}\{(0, 0, 1)\}$ is undefined for input $(0, 1, 0)$.
- A projector $\mathbf{P}$ in $\mathbb{R}^2$ that projects onto $\mathbb{R}^2$ along $\text{span}\{(1, 0)\}$ is ill-defined.

Projections exist and are unique

Does this definition make sense?

Theorem

If V and W are $\mathbb{C}^N$ -subspaces such that $V \cap W = \{\mathbf{0}\}$ and $\dim V + \dim W = N$, then $\exists$! projection matrix $\mathbf{P} \in \mathbb{C}^{N \times N}$ such that $\text{range}(\mathbf{P}) = V$ and $\ker(\mathbf{P}) = W$.

(pf: If P_1, P_2 are these projectors, show $P_1 v = P_2 v \ \forall v \in \mathbb{C}^N$)

How to construct projectors?

Initial computation: claim: Let $\underline{V} = \begin{pmatrix} \underline{v}_1 & \dots & \underline{v}_n \end{pmatrix}$, $\underline{W} = \begin{pmatrix} \underline{w}_1 & \dots & \underline{w}_{N-n} \end{pmatrix}$
be matrices whose columns hold a basis for the range, kernel,
resp.

$\begin{bmatrix} \underline{V} & \underline{W} \end{bmatrix} \in \mathbb{C}^{N \times N}$ is invertible.

Claim: $\begin{bmatrix} \underline{V} & \underline{W} \end{bmatrix}^{-1} \underline{x}$ contains coordinates of $\underline{x}$ in the joint basis $\{\underline{v}_1, \dots, \underline{v}_n, \underline{w}_1, \dots, \underline{w}_{N-n}\}$.

(Why? $\underline{x} = \underline{v}_1$, what is $\underline{y} = [\underline{V} \ \underline{W}]^{-1} \underline{x}$?

$$\begin{aligned} [\underline{V} \ \underline{W}] \underline{y} &= \underline{x} = \underline{v}_1 \\ \Rightarrow \underline{y} &= \underline{e}_1 \end{aligned}$$

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This theorem ensures that any construction for $\mathbf{P}$ identifies the unique projector.

Let $n = \dim V$.

If $\mathbf{V} \in \mathbb{C}^{N \times n}$ satisfies $\text{range}(\mathbf{V}) = V$, and $\mathbf{W} \in \mathbb{C}^{N \times (N-n)}$ satisfies $\text{range}(\mathbf{W}) = W$, then

$$\mathbf{P} = [\mathbf{V} \ \mathbf{0}_{N \times (N-n)}] [\mathbf{V} \ \mathbf{W}]^{-1}$$

(i) $\underline{\underline{P}} \underline{\underline{v}} = \underline{\underline{v}} \quad \forall \underline{\underline{v}} \in V$
(ii) $\underline{\underline{P}} \underline{\underline{w}} = \underline{\underline{0}} \quad \forall \underline{\underline{w}} \in W$ } $\Rightarrow \underline{\underline{P}}$ is the unique proj. matrix.

Notice: $I - P = \begin{bmatrix} \underline{V} & \underline{W} \end{bmatrix} \begin{bmatrix} \underline{V} & \underline{W} \end{bmatrix}^{-1} - \begin{bmatrix} \underline{V} & \underline{0} \end{bmatrix} \begin{bmatrix} \underline{V} & \underline{W} \end{bmatrix}^{-1}$
 $= \begin{bmatrix} \underline{0} & \underline{W} \end{bmatrix} \begin{bmatrix} \underline{V} & \underline{W} \end{bmatrix}^{-1}$

$\Rightarrow I - P$ is the (unique) projector onto W along V .

Projection matrices: Algebraic definitions

There is a more algebraically convenient characterization of projection matrices.

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Theorem

$\mathbf{P} \in \mathbb{C}^{n \times n}$ is a projection matrix iff it is idempotent.

Proof: Assume $P = P^2$

Given $\underline{x} \in \mathbb{C}^n$, $\underline{x} = \underbrace{P\underline{x}}_{\in \text{range}(P)} + \underbrace{(I-P)\underline{x}}_{\in \ker(P)}$ since $P(I-P)\underline{x} = (P-P^2)\underline{x} = 0$

can $\underline{x} \in \text{range}(P)$ also lie in $\ker(P)$ if $\underline{x} \neq 0$?

$$0 \neq \underline{x} = \underline{P}\underline{y} = \underline{P}^2\underline{y} = P(P\underline{y}) = P\underline{x}$$

This shows $\text{range}(P) \oplus \ker(P) = \mathbb{C}^n$
and $\text{range}(P) \cap \ker(P) = \{0\}$.

Is $Px = x$ for $x \in \text{range}(P)$?

$$x \in \text{range}(P) \Rightarrow x = Py = P^2 y = P(Py) = Px$$

"Yes"

Therefore: $P^2 = P \Rightarrow P$ is a projector.

Assume P is a projector. Let $x \in \mathbb{C}^n$ be arbitrary vector

$$x = Px + (I - P)x$$

$$Px = P^2 x + \underbrace{P(I - P)x}_{\in \ker(P)}$$

$$\Rightarrow Px = P^2 x \quad \forall x \Rightarrow P = P^2. \quad \blacksquare$$

$$\text{Ex: } \underline{P_1} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \quad \underline{P_2} = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}$$

$$P_1^2 = P_1, \quad P_2^2 = P_2$$

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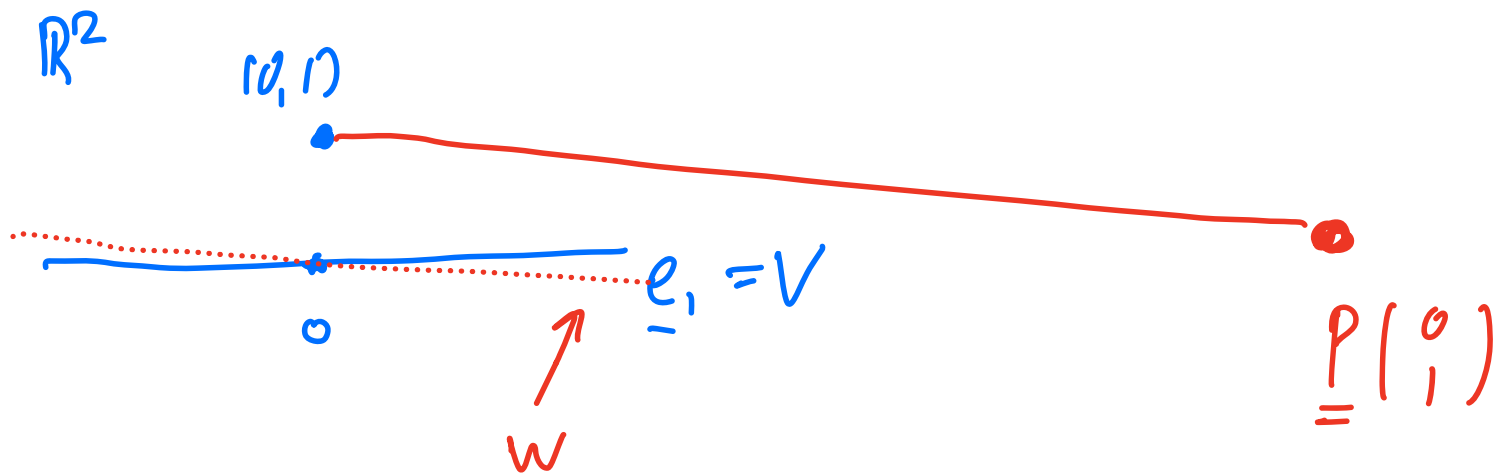
This motivates the more common definition of a projection matrix:

Definition (Algebraic definition of projection matrices)

$\mathbf{P} \in \mathbb{C}^{n \times n}$ is a projection matrix if $\mathbf{P} = \mathbf{P}^2$.

Geometry for orthogonal projections

Projection matrices can in general inflate the size (norm) of non-trivial vectors by an arbitrary amount.



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However, projecting along $\text{range}(\mathbf{P})^\perp$ is a norm non-expansive operation, i.e.,

$$\ker(\mathbf{P}) = \text{range}(\mathbf{P})^\perp \quad \implies \quad \|\mathbf{P}\mathbf{v}\|_2 \leq \|\mathbf{v}\|_2.$$

$$\begin{aligned} \|\mathbf{v}\|_2^2 &= \|\mathbf{P}\mathbf{v} + (\mathbf{I} - \mathbf{P})\mathbf{v}\|_2^2 \\ &= \|\mathbf{P}\mathbf{v}\|_2^2 + \|(\mathbf{I} - \mathbf{P})\mathbf{v}\|_2^2 \\ &\geq \|\mathbf{P}\mathbf{v}\|_2^2 \end{aligned}$$

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Projection matrices $\mathbf{P}$ satisfying $\ker(\mathbf{P}) = (\text{range} \mathbf{P})^\perp$ are *orthogonal* projections.

Definition (Geometric definition of orthogonal projection matrices)

Let $\mathbf{P}$ be a projection matrix. It's an orthogonal projector if $\ker(\mathbf{P}) = \text{range}(\mathbf{P})^\perp$.

If P is an orthogonal projector: choose basis for $\ker(P)$ and basis for $\text{range}(P)$ so that the joint basis is orthonormal.

$\Rightarrow [\underline{V} \ \underline{W}]$ has orthonormal col's.

$$P = [\underline{V} \ \underline{0}] [\underline{V} \ \underline{W}]^{-1} = [\underline{V} \ \underline{0}] \begin{bmatrix} \underline{V}^* \\ \underline{W}^* \end{bmatrix}$$

$$= \underline{V} \underline{V}^* + \underline{0} \underline{W}^*$$

$$P = \underline{V} \underline{V}^*$$

Orthogonal projectors

There is also an algebraically convenient characterization of orthogonal projection matrices.

Theorem

Let $\mathbf{P} \in \mathbb{C}^{n \times n}$ be a projection matrix. Then it is an orthogonal projector iff it is Hermitian.

$$\mathbf{P} = \mathbf{P}^*$$

Proof (half)

Assume $\mathbf{P}$ is an orthogonal projector (geometrically).

Take $x, y \in \mathbb{C}^n$, arbitrary.

$$p_x \in \text{range}(p)$$

$$(I-p) \cancel{x}^y \in \text{ker}(p)$$

$$\langle p_x, (I-p)y \rangle = 0$$

$$y^* (I-p^*) p_x = 0 \quad \forall x, y$$

$$\Rightarrow p - p^* p = 0$$

$$p = p^* p \quad p^* = (p^* p)^* = p^* p \quad \square$$

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Permutation

A final class of matrices we'll consider is permutation matrices.

Definition

For a fixed $n \in \mathbb{N}$, $\pi : [n] \rightarrow [n]$ is a permutation if it is a bijection.

Every permutation of the set of indices of a vector corresponds to a linear operator on $\mathbb{C}^n$, so it can be encoded by an $n \times n$ matrix.

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Definition

$\mathbf{P} \in \mathbb{C}^{n \times n}$ is a permutation matrix if there is a permutation map π on $[n]$ such that $\mathbf{P}\mathbf{e}_j = \mathbf{e}_{\pi(j)}$ for all $j \in [n]$.

$\mathbf{P}$'s col's are a rearrangement of col's of $\mathbf{I}$.

↑
the j -th column of $\mathbf{P}$ is $\mathbf{e}_{\pi(j)}$

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Permutation matrices are orthogonal/unitary, and hence if $\mathbf{P}$ is a permutation matrix, then $\mathbf{P}^* = \mathbf{P}^{-1}$.

The space of permutation matrices is closed under matrix multiplication, so that if $\mathbf{P}$ and $\mathbf{Q}$ are permutation matrices, then so is $\mathbf{PQ}$.

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