

Review

$$V = W \oplus W^\perp$$

Definition

Let V be a vector space with an inner product $\langle \cdot, \cdot \rangle$. Let W be a subspace of V . We define the *orthogonal complement* of W in V to be the set

$$W^\perp = \{v \in V \mid \langle v, w \rangle = 0, \forall w \in W\}.$$

Proposition

Let V be a vector space with an inner product $\langle \cdot, \cdot \rangle$. Let W be a subspace of V . Then W^\perp is a subspace of V and

$$W \cap W^\perp = \{0\}$$

Further, if V is finite-dimensional then $V = W + W^\perp$ and

$$\dim V = \dim W + \dim W^\perp$$

Orthogonal Decomposition Theorem

Theorem

Let V be a finite dimensional vector space with an inner product $\langle \cdot, \cdot \rangle$. Let W be a subspace of V . Then every vector $v \in V$ can be written uniquely in the form

$$v = w + \tilde{w}$$



where $w \in W$ and $\tilde{w} \in W^\perp$.

Definition

Let V, W, v, w and \tilde{w} be as in the above theorem. We define the *orthogonal projection of v onto W* to be

$$\text{proj}_W v = w$$

and the *component of v orthogonal to W* to be $\tilde{w} = v - \text{proj}_W v$.

Proof: We already know

that $V = W + W^\perp$, and

that $W \cap W^\perp = \{0\}$.

Let $v \in V$.

Claim: The expression

$v = w + \tilde{w}$, where $w \in W, \tilde{w} \in W^\perp$

is unique
i.e. if $v = w_1 + \tilde{w}_1 = w_2 + \tilde{w}_2$ where

$\underbrace{\omega_1, \omega_2 \in W}, \quad \text{and} \quad \underbrace{\tilde{\omega}_1, \tilde{\omega}_2 \in W}^{\sim}$
 then $\omega_1 = \tilde{\omega}_1$, and $\omega_2 = \tilde{\omega}_2$.

Suppose $v = \omega_1 + \tilde{\omega}_1 = \omega_2 + \tilde{\omega}_2$,
 as above.

$$\Rightarrow \omega_1 - \omega_2 = \tilde{\omega}_2 - \tilde{\omega}_1 \quad \text{--- (1)}$$

$$\text{Now } \omega_1 - \omega_2 \in W, \quad \tilde{\omega}_2 - \tilde{\omega}_1 \in W^+ \quad \text{--- (2)}$$

From ① & ②

$$w_1 - w_2 \in W \cap W^\perp = \{0\}$$

$$\text{and } \tilde{w}_2 - \tilde{w}_1 \in W \cap W^\perp = \{0\}$$

$$\therefore w_1 = w_2, \quad \tilde{w}_1 = \tilde{w}_2.$$

□

Formula for Orthogonal Projection with respect to a Fixed Orthogonal Basis

Let V be a vector space of dimension n . Let $\langle \cdot, \cdot \rangle$ be an inner product defined on V .

Let W be a subspace of V .

Let $\{w_1, \dots, w_m\}$ be an orthogonal basis of W and $\{w_{m+1}, \dots, w_n\}$ be an orthogonal basis of W^\perp .

Let $v \in V$. Then

$$\text{proj}_W v = \frac{\langle v, w_1 \rangle}{\langle w_1, w_1 \rangle} w_1 + \dots + \frac{\langle v, w_m \rangle}{\langle w_m, w_m \rangle} w_m$$

and

$$v - \text{proj}_W v = \frac{\langle v, w_{m+1} \rangle}{\langle w_{m+1}, w_{m+1} \rangle} w_{m+1} + \dots + \frac{\langle v, w_n \rangle}{\langle w_n, w_n \rangle} w_n$$

Orthogonal Projection in \mathbb{R}^n

$$\frac{y \cdot u_1}{u_1 \cdot u_1} = y \cdot u_1$$

Theorem

If $\{\mathbf{u}_1, \dots, \mathbf{u}_p\}$ is an orthonormal basis for a subspace W of \mathbb{R}^n , then

$$\text{proj}_W \mathbf{y} = (\mathbf{y} \cdot \mathbf{u}_1) \mathbf{u}_1 + (\mathbf{y} \cdot \mathbf{u}_2) \mathbf{u}_2 + \dots + (\mathbf{y} \cdot \mathbf{u}_p) \mathbf{u}_p$$

If $U = [\mathbf{u}_1 \ \mathbf{u}_2 \ \dots \ \mathbf{u}_p]$, then

$$\text{proj}_W \mathbf{y} = UU^T \mathbf{y}, \quad \forall \mathbf{y} \in \mathbb{R}^n$$

$$A = [a_1 \ \dots \ a_n]$$

$$\mathbf{x} = (x_1, \dots, x_n)$$

Proof:

$$\text{proj}_W \mathbf{y} = U \begin{bmatrix} (\mathbf{y} \cdot \mathbf{u}_1) \\ (\mathbf{y} \cdot \mathbf{u}_2) \\ \vdots \\ (\mathbf{y} \cdot \mathbf{u}_p) \end{bmatrix} = U \begin{bmatrix} \mathbf{u}_1^T \mathbf{y} \\ \mathbf{u}_2^T \mathbf{y} \\ \vdots \\ \mathbf{u}_p^T \mathbf{y} \end{bmatrix} = UU^T \mathbf{y}$$

$$A\mathbf{y} = x_1 a_1 + \dots + x_n a_n$$

$$U^T = \begin{bmatrix} u_1^T \\ \vdots \\ u_p^T \end{bmatrix} \quad p \times n$$

$$y = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}$$

$$Uy = \begin{bmatrix} u_1^T \\ \vdots \\ u_p^T \end{bmatrix} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}$$

$$= \begin{bmatrix} u_1^T \cdot y \\ \vdots \\ u_p^T \cdot y \end{bmatrix}$$

Example

$$U^T U = I.$$

Compute $\text{proj}_W \mathbf{y}$, where

$$\mathbf{u}_1 = \begin{bmatrix} -\frac{1}{\sqrt{3}} \\ 1 \\ \frac{1}{\sqrt{3}} \\ 1 \\ \frac{1}{\sqrt{3}} \end{bmatrix}, \mathbf{u}_2 = \begin{bmatrix} \frac{2}{\sqrt{6}} \\ 1 \\ \frac{1}{\sqrt{6}} \\ 1 \\ \frac{1}{\sqrt{6}} \end{bmatrix}, \mathbf{y} = \begin{bmatrix} -9 \\ 1 \\ 6 \end{bmatrix}$$

and

$$W = \text{Span} \{ \mathbf{u}_1, \mathbf{u}_2 \}.$$

$$UU^T = \begin{bmatrix} -\frac{1}{\sqrt{3}} & \frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{6}} \end{bmatrix} \begin{bmatrix} -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ \frac{2}{\sqrt{6}} & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

So $\text{proj}_W \mathbf{y} = UU^T \mathbf{y} = \begin{bmatrix} -9 \\ 7 \\ \frac{7}{2} \\ \frac{7}{2} \end{bmatrix}$

QR Factorization

Definition

Let A be an $\underline{m \times n}$ matrix having linearly independent columns. A *QR-factorization* of A is a matrix factorization of the form

$$\underline{A = QR}$$

where Q is an $m \times n$ matrix having orthonormal columns and R is an upper triangular $n \times n$ matrix with strictly positive diagonal entries.

QR Algorithm

Let $\mathbf{a}_1, \dots, \mathbf{a}_n$ be the columns of A . Apply the Gram-Schmidt algorithm the the vectors $\mathbf{a}_1, \dots, \mathbf{a}_n$ to obtain an orthogonal set $\mathbf{b}_1, \dots, \mathbf{b}_n$, i.e.

$$\mathbf{b}_1 = \mathbf{a}_1$$

$$\mathbf{b}_k = \mathbf{a}_k - \sum_{i=1}^{k-1} \frac{\langle \mathbf{a}_k, \mathbf{b}_i \rangle}{\langle \mathbf{b}_i, \mathbf{b}_i \rangle} \mathbf{b}_i, \quad \text{for } k = 2, \dots, n.$$

Normalize the vectors \mathbf{b}_k to obtain an orthonormal set,

$$\mathbf{q}_k = \frac{\mathbf{b}_k}{\|\mathbf{b}_k\|}, \quad \text{for } k = 1, \dots, n.$$


Thus we obtain the matrix

$$Q = [\mathbf{q}_1 \quad \dots \quad \mathbf{q}_n]$$

having orthonormal columns. Therefore $Q^T Q = I$. We put $R = Q^T A$, and this gives us the required matrix factorization.

If time permits on Thursday, we will see a proof of why this works.

- ① Why is R upper triangular
 - ② Why is R strictly positive diagonal entries?
- with $A = QR$?

Algebra of Linear Transformations; the space $L(V, W)$

Lemma

Let V and W be vector spaces. Let \mathcal{U} be the set of all functions from V to W . Then \mathcal{U} is a vector space under pointwise addition and pointwise scalar multiplication of functions.

Proof: Exercise. Verify all axioms. (HW) .

Definition

Let V and W be vector spaces. The set of all linear transformations from V to W is called $L(V, W)$.

Proposition

$L(V, W)$ is a vector space under pointwise addition of functions and pointwise multiplication of functions by real numbers.

Proof: Let $T, S \in L(V, W)$

Claim: $T + S$ is a linear transformation.

Let $v, w \in V$.

$$(T+S)(v+w)$$

$$= \underbrace{T(v+w)} + \underbrace{S(v+w)}$$

$$= T(v) + T(w) + S(v) + S(w)$$

$$= \underbrace{T(v) + S(v)} + \underbrace{T(w) + S(w)}$$

$$= (T + S)(v) + (T + S)(w).$$

\therefore Let $c \in \mathbb{R}$, $v \in V$.

$$(T + S)(cv) = T(cv) + S(cv)$$

$$= cT(v) + cS(v)$$

$$= c(T + S)(v).$$

$$\therefore T + S \in L(v, w)$$

(HW) If $T \in L(v, w)$ then $cT \in L(v, w)$
 $\forall c \in \mathbb{R}$.

$$T \in L(V, W) \longrightarrow [T]_{\beta, \phi}$$

Theorem

Let $\dim V = n$ and $\dim W = m$. Then $L(V, W)$ is isomorphic to the vector space $M_{m \times n}(\mathbb{R})$ of $m \times n$ matrices.

(In other words there exists a bijective linear transformation from $L(V, W)$ to $M_{m \times n}(\mathbb{R})$).

Fix a basis β for V
and a basis ϕ for W .

$$L(v, w) \xrightarrow{\alpha} M_{m \times n}(\mathbb{R})$$

$T \mapsto$ matrix of T
wrt β & \mathcal{C} .

$$\alpha(T) = [T]_{\beta, \mathcal{C}}.$$

We will show, ^{that} $\alpha: L(V, W) \rightarrow M_{m \times n}(\mathbb{R})$
sending $T \mapsto [T]_{\beta, \gamma}$
is a linear transformation.

Let $T_1, T_2 \in L(V, W)$.

$$\text{Claim: } \underbrace{[T_1 + T_2]_{\beta, \gamma}} = \underbrace{[T_1]_{\beta, \gamma} + [T_2]_{\beta, \gamma}}$$

Let $v \in V$.

$$[(T_1 + T_2)(v)]_{\mathcal{B}}$$

$$= [T_1(v) + T_2(v)]_{\mathcal{B}}$$

$$= [T_1(v)]_{\mathcal{B}} + [T_2(v)]_{\mathcal{B}}$$

(It can be seen that $x \mapsto (x)_{\mathcal{B}}$ is linear)

$$= \underbrace{\begin{bmatrix} T_1 \end{bmatrix}_{\beta, \ell}}_{\beta} \underbrace{\begin{bmatrix} v \end{bmatrix}_{\beta}}_{\beta} + \underbrace{\begin{bmatrix} T_2 \end{bmatrix}_{\beta, \ell}}_{\beta} \underbrace{\begin{bmatrix} v \end{bmatrix}_{\beta}}_{\beta}$$

$$\rightarrow \left(\begin{bmatrix} T_1 \end{bmatrix}_{\beta, \ell} + \begin{bmatrix} T_2 \end{bmatrix}_{\beta, \ell} \right) \begin{bmatrix} v \end{bmatrix}_{\beta}$$

$$\therefore \begin{bmatrix} T_1 + T_2 \end{bmatrix}_{\beta, \ell} = \begin{bmatrix} T_1 \end{bmatrix}_{\beta, \ell} + \begin{bmatrix} T_2 \end{bmatrix}_{\beta, \ell}$$

$\therefore \alpha$ respects ^{vector} addition
in $L(V, W)$.

(HW) : α also respects
scalar multiplication.
i.e. $[cT]_{\beta, \ell} = c[T]_{\beta, \ell}$
 $\forall c \in \mathbb{R}, \forall T \in L(V, W)$

Let us show that α is
injective.

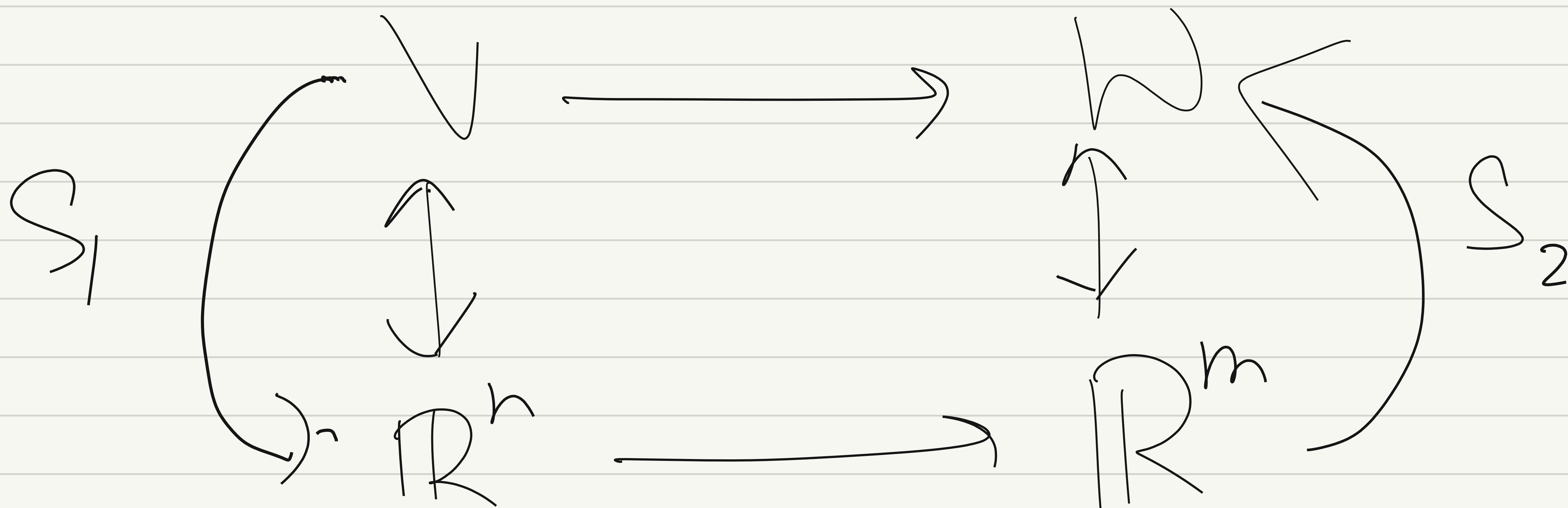
Let $T \in \text{Ker } \alpha$, i.e.

$$\alpha(T) = 0.$$

$$\begin{aligned} \Rightarrow [T(v)]_{\mathcal{C}} &= \alpha(T)[v]_{\beta} = 0, \\ \Rightarrow T(v) &= 0, \forall v \in V. \Rightarrow T = 0. \end{aligned} \quad \forall v \in V.$$

$\therefore \alpha$ is injective.

Claim: α is surjective.



S_1 : coordinate map from
 $V \longrightarrow \mathbb{R}^n$ sending
 $v \longmapsto [v]_\beta$

S_2 : coordinate map from
 $W \longrightarrow \mathbb{R}^m$

sending

$w \longmapsto [w]_e$.

Given n matrix

$$A \in M_{m \times n}(\mathbb{R})$$

define

$$T(v) = \sum_{i=1}^n (A S_i(v)).$$

This is

composite of linear maps,
therefore linear.

$$[T(v)]_{\mathcal{P}}$$

$$= [S_1^{-1} (A S_1(v))]_{\mathcal{P}}$$

$$= A S_1(v)$$

$$= [A(v)]_{\mathcal{B}}, \quad \forall v \in V.$$

$$\therefore A = [T]_{\beta, \mathcal{L}}$$



Theorem

Let U, V and W be finite dimensional vector spaces of dimensions m, n and k respectively, having bases \mathcal{A}, \mathcal{B} and \mathcal{C} respectively.

Let $S \in L(U, V)$ and $T \in L(V, W)$. Then $T \circ S \in L(U, W)$ and

$$\underline{[T \circ S]_{\mathcal{A}, \mathcal{C}}} = \underline{[T]_{\mathcal{B}, \mathcal{C}} [S]_{\mathcal{A}, \mathcal{B}}}$$

Proof : We first show $T \circ S \in L(U, W)$

Let $u_1, u_2 \in U$.

$$(T \circ S)(u_1 + u_2)$$

$$= T(S(u_1 + u_2))$$

$$= T(S(u_1) + S(u_2))$$

$$= T(S(u_1)) + T(S(u_2))$$

$$= TOS(u_1) + (TOS)(u_2)$$

\therefore TOS satisfies first
condition for being a
linear transformation.

Let $u \in U$, $c \in \mathbb{R}$,

$T \circ S(u)$

$\vdash T(S(u))$

$\vdash T(c S(u))$

$\vdash c T(S(u))$
 $\vdash c T \circ S(u)$

$\therefore TOS \in L(V, w)$
~~let~~ $u \in U$

$$[(TOS)(u)]_{\ell}$$

$$\begin{aligned}
 &= [T(S(u))]_{\ell} \\
 &= [T]_{\beta, \ell} [S(u)]_{\beta}
 \end{aligned}$$

$$= \left(\begin{bmatrix} T \end{bmatrix}_{\beta, \ell} \begin{bmatrix} S \end{bmatrix}_{A, \beta} \begin{bmatrix} u \end{bmatrix}_A \right)$$

$$\therefore \begin{bmatrix} T \text{ } S \end{bmatrix}_{A, \ell} = \begin{bmatrix} T \end{bmatrix}_{\beta, \ell} \begin{bmatrix} S \end{bmatrix}_{A, \beta}$$