

6C Orthogonal Complements and Minimization Problems

- V and W continue to denote inner product spaces over \mathbf{F} .

Orthogonal Complements

- Definition 6.46: The **orthogonal complement** of any subset $U \subseteq V$ is the subset $U^\perp = \{v \in V \mid u \perp v \text{ for all } u \in U\}$.
 - We are mainly interested in the case in which U is a subspace.
- Examples 6.47:
 - The orthogonal complement of a line/plane through the origin in \mathbf{R}^3 is a plane/line through the origin.
 - In \mathbf{F}^n , the orthogonal complement of $\text{span}(e_1, \dots, e_k)$ is $\text{span}(e_{k+1}, \dots, e_n)$.
 - This holds more generally whenever e_1, \dots, e_n is an orthonormal basis of V .
- Proposition 6.48:
 - (a) U^\perp is a subspace
 - (b) $\{0\}^\perp = V$
 - (c) $V^\perp = \{0\}$
 - (d) $U \cap U^\perp \subseteq \{0\}$
 - (e) if $U_1 \subseteq U_2$, then $U_2^\perp \subseteq U_1^\perp$.
- Proposition 6.49: if $U \subseteq V$ is a finite-dimensional subspace, then $V = U \oplus U^\perp$.
- Definition 6.55: if $U \subseteq V$ is a finite-dimensional subspace, then the **orthogonal projection** onto U is the operator $P_U \in \mathcal{L}(V)$ such that $P_U(u + w) = u$ for $u \in U$ and $w \in U^\perp$.
 - Explicitly, if u_1, \dots, u_k is an orthogonal basis of U , then $P_U(w) = \sum_{i=1}^k \frac{\langle w, u_i \rangle}{\langle u_i, u_i \rangle} u_i$.
- Corollary 6.51: if V is finite-dimensional and $U \subseteq V$ is a subspace, then $\dim U^\perp = \dim V - \dim U$.
- Corollary 6.52: if $U \subseteq V$ is a finite-dimensional subspace, then $(U^\perp)^\perp = U$.
- Corollary 6.54: if $U \subseteq V$ is a finite-dimensional subspace, then $U^\perp = \{0\} \iff U = V$.
- Proposition 6.57: if U is a finite-dimensional subspace of V , then (i) P_U is a projection operator onto U (see below), and (ii) $\|P_U v\| \leq \|v\|$ for all $v \in V$.

Aside on projection operators

- This is something that we should have covered when we discussed direct sums; during this aside, let V be an arbitrary vector space (rather than an inner product space).
- Definition: given a subspace $U \subseteq V$ of a vector space V , a **projection operator** onto U is an operator $P \in \mathcal{L}(V)$ with $\text{range}(P) = U$ and $P|_U = I$.
- Proposition: $P \in \mathcal{L}(V)$ is a projection operator if and only if $P^2 = P$ (an operator with this property is also called **idempotent**), in which case it is a projection operator onto $U = \text{range}(P)$.
 - Proof: if P is a projection onto $U \subseteq V$, then $P(v) \in U$ for any $v \in V$, hence $P(P(v)) = P(v)$.
 - Conversely, if $P^2 = P$, we may set $U = \text{range } P$; then any $u \in U$ is of the form $u = Pv$, and hence $Pu = PPv = Pv = u$.
- Definition: given a subspace $U \subseteq V$, a **complement** of U is a subspace $W \subseteq V$ with $U \oplus W = V$.
- Theorem: let $U \subseteq V$ be a subspace. Then
 - (a) if P is a projection onto U , then $\text{null } P$ is a complement of U ;
 - (b) conversely, if W is any complement of U , there is a unique projection P onto U with $\text{null } P = W$.
 - Proof of (a): if $u \in U \cap \text{null } P$, then $u = Pu = 0$, where the first equality follows from $P|_U = I$; and given any $v \in V$, we have $P(v - Pv) = Pv - PPv = Pv - Pv = 0$, hence $v - Pv \in \text{null } P$, and so $v = Pv + (v - Pv) \in U + \text{null } P$.
 - Proof of (b): we define a projection onto U as follows: given $v \in V$, we can write v uniquely as $v = u + w$ with $u \in U$ and $w \in W$, and we set $Pv = u$; it's easy to check that this is a projection on to U with nullspace W . To prove uniqueness, given any projection P onto U with nullspace W , if $v = u + w$ as above, then we must have $Pv = Pu + Pw = u + 0 = u$.
- Proposition: if P is a projection onto U , then $I - P$ is a projection onto the complement $\text{null } P$ of U .
 - Proof: we have $(I - P)^2 = I^2 - 2P + P^2 = I - 2P + P = I - P$, hence $I - P$ is a projection operator; it remains to see that $\text{range}(I - P) = \text{null } P$.
 - For any $v \in V$, we have $P((I - P)v) = P(v - Pv) = Pv - PPv = 0$, so $\text{range}(I - P) \subseteq \text{null } P$.
 - And for any $v \in \text{null } P$, we have $v = v - Pv = (I - P)v \in \text{range}(I - P)$, so $\text{null } P \subseteq \text{range}(I - P)$.

Orthogonal complements and annihilators

- We return to our convention that V and W denote inner product spaces over \mathbf{F} .
- Proposition: if V is finite-dimensional, then for any subspace $U \subseteq V$, the isomorphism $\overline{V} \rightarrow V'$ defined in the proof of (6.42) satisfies $F(U^\perp) = U^0$.
 - This follows directly from the definition: $v \in U^\perp$ if and only if $\langle u, v \rangle = 0$ for all $u \in U$ if and only if $F(v)|_U = 0$ if and only if $F(v) \in U^0$.

Minimization Problems

- Proposition 6.61: if U is a finite-dimensional subspace of V , then for any $v \in V$, $P_U v$ is the closest point to v in U , in the sense that $\|v - P_U v\| < \|v - u\|$ for any other $u \in U$.
- Example 6.63: we can use this to approximate the function $\sin: [-\pi, \pi]$ by a polynomial by letting $V = \mathcal{C}^0([-\pi, \pi])$ with inner product $\langle f, g \rangle = \int_{-\pi}^{\pi} fg$, and letting $U = \text{span}(1, x, \dots, x^n)$.

7A Self-Adjoint and Normal Operators

- V and W continue to denote inner product spaces over \mathbf{F} .
- We will now introduce self-adjoint operators—which are an analogue for operators of symmetric matrices—and normal operators, which are a generalization of self-adjoint operators.
- The most important fact about these, which we will prove in the next section, is the (finite-dimensional) **spectral theorem** (Theorems 7.29 and 7.31), which says that
 - (a) every self-adjoint operator T on a finite-dimensional vector space V is diagonalizable with respect to an orthonormal basis,
 - (b) if $\mathbf{F} = \mathbf{C}$, then more generally, every **normal** operator T is diagonalizable with respect to an orthonormal basis,
 - (c) conversely, if an operator is diagonalizable with respect to an orthonormal basis then, if $\mathbf{F} = \mathbf{R}$, it is self-adjoint, and if $\mathbf{F} = \mathbf{C}$, it is normal.

Adjoints

- Proposition/Definition 7.1 and 7.4: for any $T \in \mathcal{L}(V, W)$, there exists at most one linear map $T^*: W \rightarrow V$ satisfying $\langle Tv, w \rangle = \langle v, T^*w \rangle$ for all $v \in V$ and $w \in W$; if it exists, T^* is called the **adjoint** of T . Moreover, if $\dim V, \dim W < \infty$, then the adjoint T^* always exists.
- Remark: supposing $\dim V, \dim W < \infty$, let $\varphi_V: \overline{V} \xrightarrow{\sim} V'$ and $\varphi_W: \overline{W} \xrightarrow{\sim} W'$ be the isomorphisms given by $\varphi_V(v) = \langle -, v \rangle$ and $\varphi_W(w) = \langle -, w \rangle$. Then for any $T: \mathcal{L}(V, W)$ and $w \in \overline{W} = W$, we have $\varphi_V(T^*w) = T'(\varphi_W w)$.

$$\begin{array}{ccc}
 \overline{W} & \xrightarrow{T^*} & \overline{V} \\
 \varphi_W \downarrow \wr & & \wr \downarrow \varphi_V \\
 W' & \xrightarrow{T'} & V'
 \end{array}$$

- Thus, roughly speaking, the adjoint of a linear map between inner product spaces is “isomorphic” to the dual map; in particular, all the properties of the dual map carry over to give properties of the adjoint map.
- Example: for a linear map $T: \mathbf{R}^{m,1} \rightarrow \mathbf{R}^{n,1}$ (with the Euclidean inner product) given by multiplying by a matrix A , the adjoint $T^*: \mathbf{R}^{n,1} \rightarrow \mathbf{R}^{m,1}$ is given by multiplying by A^t .
 - Indeed, one way of writing the Euclidean inner product on $\mathbf{R}^{n,1}$ is as the matrix product $\langle u, v \rangle = u^t \cdot v \in \mathbf{R}^{1,1}$.

– Hence $\langle A \cdot u, v \rangle = (A \cdot u)^t \cdot v = u^t \cdot A^t \cdot v = u^T \cdot (A^t \cdot v) = \langle u, A^t \cdot v \rangle$.

- Proposition 7.5: Suppose V and W are finite-dimensional. Then
 - (a,b) the map $T \mapsto T^*$ is an anti-linear map $\mathcal{L}(V, W) \rightarrow \mathcal{L}(W, V)$
 - (c) $(T^*)^* = T$ for any $T \in \mathcal{L}(V, W)$
 - (d) $(ST)^* = T^*S^*$ for $T \in \mathcal{L}(V, W)$ and $S \in \mathcal{L}(W, U)$ (where U is a finite-dimensional inner-product space over \mathbf{F})
 - (e) $I_V^* = I_V$
 - (f) if $T \in \mathcal{L}(V, W)$ is invertible, then T^* is invertible and $(T^*)^{-1} = (T^{-1})^*$.
- Proposition 7.6: if V and W are finite-dimensional, then for any $T \in \mathcal{L}(V, W)$:
 - (a) $\text{null } T^* = (\text{range } T)^\perp$
 - (b) $\text{range } T^* = (\text{null } T)^\perp$
 - (c) $\text{null } T = (\text{null } T^*)^\perp$
 - (d) $\text{range } T = (\text{null } T^*)^\perp$
- Definition 7.7: the **conjugate transpose** of a matrix $A \in \mathbf{F}^{m,n}$ is the matrix A^* defined by $(A^*)_{j,k} = \overline{A_{k,j}}$.
- Proposition 7.9: given $T \in \mathcal{L}(V, W)$ and orthonormal bases for V and W , we have $\mathcal{M}(T^*) = \overline{\mathcal{M}(T)}$.

Self-Adjoint Operators

- Definition 7.10: $T \in \mathcal{L}(V)$ is **self-adjoint** if $T = T^*$, i.e., if $\langle Tu, v \rangle = \langle u, Tv \rangle$ for all $u, v \in V$.
 - Remark: given an orthonormal basis V , T is self-adjoint if and only if $\mathcal{M}(T)$ is **conjugate-symmetric** (also known as being **Hermitian**), meaning that $\mathcal{M}(T)^* = \mathcal{M}(T)$.
- Proposition 7.12: Every eigenvalue of a self-adjoint operator is real.
 - Of course, this is only an interesting result when $\mathbf{F} = \mathbf{C}$.
- Proposition 7.13: if $\mathbf{F} = \mathbf{C}$, then for any $T \in \mathcal{L}(V)$, we have $Tv \perp v$ for all $v \in V$ if and only if $T = 0$.
 - This is not true when $\mathbf{F} = \mathbf{R}$, for instance when $T(x, y) = (-y, x)$, a rotation by 90° .
- Proposition 7.14: if $F = \mathbf{C}$, then $T \in \mathcal{L}(V)$ is self-adjoint if and only if $\langle Tv, v \rangle \in \mathbf{R}$ for all $v \in V$.
 - This is obviously false when $\mathbf{F} = \mathbf{R}$, since there are real operators that are not self-adjoint.
- Proposition 7.16: if $T \in \mathcal{L}(V)$ is self-adjoint, then $Tv \perp v$ for all $v \in V$ if and only if $T = 0$.
 - When $\mathbf{F} = \mathbf{C}$, this is implied by the stronger result (7.13), which does not have the self-adjoint hypothesis.

Normal operators

- Definition 7.18: $T \in \mathcal{L}(V)$ is **normal** if T and T^* commute.
 - Note that every self-adjoint operator is normal.
 - The main significance of normal operators is the **spectral theorem** in section 7B: if $F = \mathbf{C}$ and $\dim V < \infty$, then T is normal if and only if there is an orthonormal basis \mathcal{B} such that $\mathcal{M}(T, \mathcal{B})$ is diagonal.
- Example 7.19: the operator on \mathbf{F}^2 with matrix $\begin{bmatrix} 2 & -3 \\ 3 & 2 \end{bmatrix}$ is normal but not self-adjoint.
- Proposition 7.20: $T \in \mathcal{L}(V)$ is normal iff $\|Tv\| = \|T^*v\|$ for all $v \in V$.
- Proposition 7.21: if $T \in \mathcal{L}(V)$ is normal, then
 - (a) $\text{null } T = \text{null } T^*$
 - (b) $\text{range } T = \text{range } T^*$ assuming $\dim V < \infty$
 - (c) $V = \text{null } T \oplus \text{range } T$ assuming $\dim V < \infty$
 - (d) $T - \lambda I$ is normal for all $\lambda \in \mathbf{F}$
 - (e) $Tv = \lambda v$ iff $T^*v = \bar{\lambda}v$ for $v \in V$ and $\lambda \in F$.
- Proposition 7.22: if $T \in \mathcal{L}(V)$ is normal, then eigenvectors of T with distinct eigenvalues are orthogonal.
 - This has a simpler proof in the special case that T is self-adjoint.
- Proposition 7.23: when $F = \mathbf{C}$, then $T \in \mathcal{L}(V)$ is normal if and only if there are self-adjoint commuting $A, B \in \mathcal{L}(V)$ with $T = A + iB$.
 - A version of the statement that makes sense over \mathbf{R} is: T is normal if and only if there are commuting $A, B \in \mathcal{L}(V)$ with $T = A + B$ and $A = A^*$ and $B = -B^*$.

7B Spectral Theorem

- V and W continue to denote inner product spaces over \mathbf{F} .

Real Spectral Theorem

- Lemma 7.26: if $\dim V < \infty$ and $T \in \mathcal{L}(V)$ is self-adjoint and $b, c \in \mathbf{R}$ satisfy $b^2 - 4c < 0$, then $T^2 + bT + cI \in \mathcal{L}(V)$ is invertible.
- Lemma 7.27: if $\dim V < \infty$ and $T \in \mathcal{L}(V)$ is self-adjoint, then the minimal polynomial m_T is a product of real linear factors.
- Theorem 7.29 (real finite-dimensional spectral theorem): if $\mathbf{F} = \mathbf{R}$ and $\dim V < \infty$, then any $T \in \mathcal{L}(V)$ is self-adjoint if and only if it is diagonalizable with respect to some orthonormal basis.
 - Remark: an equivalent condition to being diagonalizable with respect to an orthonormal basis is that the eigenspaces of T are mutually orthogonal and their sum is V .

Complex Spectral Theorem

- Theorem 7.31 (complex finite-dimensional spectral theorem): if $\mathbf{F} = \mathbf{C}$ and $\dim V < \infty$, then any $T \in \mathcal{L}(V)$ is normal if and only if it is diagonalizable with respect to some orthonormal basis.