

Recall: • bilinear form $b: V \times V \rightarrow k \iff \varphi_b \in \text{Hom}(V, V^*)$
 $\varphi_b(v) = b(v, \cdot) = \left(w \mapsto b(v, w) \right)$

This gives an isom. $B(V) \cong \text{Hom}(V, V^*)$.

• b is nondegenerate if φ_b is an isom.

• in a basis (e_1, \dots, e_n) , represent b by a matrix A with entries $a_{ij} = b(e_i, e_j)$

$$b(\sum x_i e_i, \sum y_j e_j) = \sum a_{ij} x_i y_j = X^T A Y.$$

b is symmetric iff A is symmetric ($a_{ij} = a_{ji}$)

nondegenerate iff A is invertible

* Def: If $S \subseteq V$ is a subspace of a vector space equipped with a bilinear form $b: V \times V \rightarrow k$, we define its orthogonal complement
 $S^\perp = \{ v \in V \mid b(v, w) = 0 \ \forall w \in S \}$. This is a vector subspace.

* corrected from an earlier version where v, w were swapped, for consistency with the lemma below.

△ This is most intuitive if b is symmetric or skew. Otherwise we have to worry about "left-orthogonal" vs. "right-orthogonal" to S .

* Lemma: If b is nondegenerate then $\dim S^\perp = \dim V - \dim S$ (else \geq)

Proof: $S^\perp = \text{Ker} \left(\begin{array}{c} V \rightarrow S^* \\ v \mapsto \varphi_b(v)|_S \end{array} \right)$ composition of $\varphi_b: V \rightarrow V^*$ and restriction $r: V^* \rightarrow S^*$
 $l \mapsto l|_S$

By dim-formula, $\dim S^\perp = \dim V - \text{rank}(r \circ \varphi_b)$. If b is nondegenerate then

φ_b isomorphism and r surjective $\Rightarrow \text{rank}(r \circ \varphi_b) = \dim S^* = \dim S$; in general \leq □

Ex: • $V = \mathbb{R}^n$ with the standard dot product $b(v, w) = \sum_{i=1}^n v_i w_i$; then

$V = S \oplus S^\perp$ the "usual" orthogonal complement

because: $S \cap S^\perp = \{0\}$ (see below) and $\dim S + \dim S^\perp = \dim V$.

• but for $b: k^2 \times k^2 \rightarrow k$

$$b((x_1, x_2), (y_1, y_2)) = x_1 y_2 - x_2 y_1 \quad (\text{skew-symmetric, nondegenerate})$$

$S \subseteq k^2$ 1-dim! subspace spanned by any nonzero vector $v \Rightarrow \underline{S^\perp = S!!}$
 (because $b(v, w) = 0 \iff w \in k \cdot v = S$)

Inner product spaces:

Defⁿ: An inner product space is a vector space V over \mathbb{R} together with
 a symmetric definite positive bilinear form $\langle \cdot, \cdot \rangle: V \times V \rightarrow \mathbb{R}$

Symmetric: $\langle u, v \rangle = \langle v, u \rangle$ Def. positive: $\langle u, u \rangle \geq 0 \ \forall u \in V$, and $\langle u, u \rangle = 0$ iff $u = 0$.

This definition only makes sense over an ordered field so " $\langle u, u \rangle \geq 0$ " makes sense. (2)
 In practice this means \mathbb{R} . We can't do this over \mathbb{C} . (we'll see a workaround: Hermitian forms)

• Let $\varphi: V \rightarrow V^*$ be the linear map corresponding to $\langle \cdot, \cdot \rangle$.
 $v \mapsto \langle v, \cdot \rangle$

$\langle \cdot, \cdot \rangle$ definite positive $\Rightarrow \varphi$ is injective (since $\forall v \neq 0, \varphi(v) \neq 0! \varphi(v)(v) > 0$).
 \Rightarrow (assuming $\dim V < \infty$) φ is an iso. $V \xrightarrow{\sim} V^*$, i.e. $\langle \cdot, \cdot \rangle$ is nondegenerate. (The converse is false: $\langle \cdot, \cdot \rangle$ nondegenerate \nRightarrow positive).

Prop: $\| V$ finite-dim inner product space, $S \subset V$ subspace $\Rightarrow V = S \oplus S^\perp$.

Pf: • we've seen: $\langle \cdot, \cdot \rangle$ is nondegenerate so $\dim S^\perp = \dim V - \dim S$.

• since $\langle \cdot, \cdot \rangle$ is positive definite, $v \in S \cap S^\perp \Rightarrow \langle v, v \rangle = 0 \Rightarrow v = 0$.

So $S \cap S^\perp = \{0\}$. Since dimensions add up to $\dim V$, this implies $S \oplus S^\perp = V$. \square

Def: • The norm of a vector is $\|v\| = \sqrt{\langle v, v \rangle}$.

• $v, w \in V$ are orthogonal if $\langle v, w \rangle = 0$.

Observe: $\|v-w\|^2 = \langle v-w, v-w \rangle = \|v\|^2 + \|w\|^2 - 2\langle v, w \rangle$.



\rightarrow if v and w are orthogonal then $\|v-w\|^2 = \|v\|^2 + \|w\|^2$ Pythagorean theorem

\rightarrow in general, by analogy with law of triangles, we define the angle b/w 2 vectors

$\angle(v, w) = \cos^{-1} \left(\frac{\langle v, w \rangle}{\|v\| \|w\|} \right)$. This only makes sense if $|\langle v, w \rangle| \leq \|v\| \|w\|$?

Theorem (Cauchy-Schwarz inequality) $\| \forall u, v \in V, |\langle u, v \rangle| \leq \|u\| \|v\|$.

Pf: The inequality is unaffected by scaling so we can assume $\|u\| = 1$.

Decompose v along $V = S \oplus S^\perp$ where $S = \text{span}(u) \subset V$. Explicitly,

$v = v_1 + v_2$, $v_1 = \langle v, u \rangle u \in \text{span}(u)$, $v_2 = v - \langle v, u \rangle u$ orthogonal to u .

Then $v_1 \perp v_2$ so $\|v\|^2 = \|v_1\|^2 + \|v_2\|^2 \geq \|v_1\|^2 = \langle v, u \rangle^2$.

This is the desired inequality for $\|u\| = 1$. \square

Def: $\| V$ finite dim! / \mathbb{R} with inner product $\langle \cdot, \cdot \rangle$. A basis v_1, \dots, v_n of V is said to be orthonormal if $\langle v_i, v_j \rangle = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$ (i.e. $\|v_i\| = 1$)
 (i.e. $v_i \perp v_j$)

In such a basis, $(V, \langle \cdot, \cdot \rangle) \cong (\mathbb{R}^n \text{ with standard dot product})$.

Thm: $\|$ Every finite-dimensional inner product space $(V, \langle \cdot, \cdot \rangle)$ has an orthonormal basis. (3)

Proof 1: By induction on $\dim(V)$: choose $v \neq 0 \in V$, let $v_1 = \frac{v}{\|v\|}$ so $\|v_1\| = 1$.

Then let $S = \text{span}(v_1)$, $V = S \oplus S^\perp$.

Let v_2, \dots, v_n be an orthonormal basis for S^\perp (the restriction of $\langle \cdot, \cdot \rangle$ to S^\perp is an inner product!).

Then v_1, \dots, v_n is an orthonormal basis for V (check!). □

Proof 2: start with any basis w_1, \dots, w_n of V and use the Gram-Schmidt process.

First set $v_1 = \frac{w_1}{\|w_1\|}$. Then take $w_2 - \langle w_2, v_1 \rangle v_1$ which is $\perp v_1$

(and nonzero by independence of w_i), set $v_2 = \frac{w_2 - \langle w_2, v_1 \rangle v_1}{\|w_2 - \langle w_2, v_1 \rangle v_1\|}$

and so on, set $v_j = \frac{w_j - \sum_{i=1}^{j-1} \langle w_j, v_i \rangle v_i}{\|w_j - \sum_{i=1}^{j-1} \langle w_j, v_i \rangle v_i\|}$. Then (v_1, \dots, v_n) is an orthonormal basis □

So: every finite dimⁿ inner product space $(V, \langle \cdot, \cdot \rangle)$ is isomorphic (as an inner product space, not just as a vector space) to standard \mathbb{R}^n , $n = \dim V$.

Operators on inner product spaces: Let $(V, \langle \cdot, \cdot \rangle)$ inner product space. There are two special classes of linear operators on V of interest to us.

Def: $\|$ Say $T: V \rightarrow V$ is an orthogonal operator if it respects the inner product, i.e. $\langle Tu, Tv \rangle = \langle u, v \rangle \forall u, v \in V$.

(In other terms, T "preserves lengths and angles").

Remarks: 1) orthogonal operators map orthonormal bases to orthonormal bases!

$$\langle Te_i, Te_j \rangle = \langle e_i, e_j \rangle = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{otherwise} \end{cases}$$

in particular, orthogonal operators are always invertible!

2) If T is orthogonal then T^{-1} is orthogonal

$$\langle T^{-1}u, T^{-1}v \rangle = \langle T(T^{-1}u), T(T^{-1}v) \rangle = \langle u, v \rangle \forall u, v$$

\uparrow
 T orthogonal

If T_1, T_2 are orthogonal then so is $T_1 T_2$ (check!)

Hence: $\|$ orthogonal operators form a subgroup of $\text{Aut}(V)$.

3) If M is the matrix representing T in an orthonormal basis, then $M^T M = I$.

(4)

Indeed: entries of $M^T M =$ dot products of columns of M !

$$(M^T M)_{ij} = \sum_k M_{ik}^T M_{kj} = \sum_k M_{ki} M_{kj} = \langle M(e_i), M(e_j) \rangle = \langle e_i, e_j \rangle.$$

Def: Let $T: V \rightarrow V$ linear operator on an inner product space $(V, \langle \cdot, \cdot \rangle)$. There exists a unique linear operator $T^*: V \rightarrow V$, called the adjoint of T , such that $\langle v, T(w) \rangle = \langle T^*(v), w \rangle \quad \forall v, w \in V$.

Indeed: given $v \in V$, the linear functional $V \rightarrow \mathbb{R}$
 $w \mapsto \langle v, T(w) \rangle$

is, using nondegeneracy of $\langle \cdot, \cdot \rangle$, given by the inner product of w with some element of V , which we call $T^*(v)$; then check this has linear dependence on v .

Equivalently: $\langle \cdot, \cdot \rangle$ defines an isom. $\varphi: V \xrightarrow{\sim} V^*$. Then T^* is the composition

$$\text{of } V \xrightarrow{\varphi} V^* \xrightarrow{T^t} V^* \xrightarrow{\varphi^{-1}} V$$

$$v \mapsto \langle v, \cdot \rangle \mapsto \langle v, T(\cdot) \rangle = \langle T^*(v), \cdot \rangle \mapsto T^*(v).$$

Def: $T: V \rightarrow V$ is self-adjoint if $T^* = T$. (ie. $\langle v, Tw \rangle = \langle Tv, w \rangle \quad \forall v, w$).

* In an orthonormal basis (e_1, \dots, e_n) of V , $\langle v, w \rangle = v^t w$, so
 if matrix of T is M , T^* is N ,
transpose gives a row vector column vector

$$\left. \begin{aligned} \langle v, T(w) \rangle &= v^t M w \\ \langle T^*(v), w \rangle &= (Nv)^t w = v^t N^t w \end{aligned} \right\} \Rightarrow \text{comparing: } N^t = M, \text{ so } N = M^t.$$

Hence: $M(T^*) = M(T)^t$ in orthonormal basis; T is self-adjoint $\Leftrightarrow M(T)$ symmetric

Note that self-adjoint operators (\sim symmetric matrices) need not be invertible.

For example 0 is a self-adjoint operator...

Prop: If T is self-adjoint and $S \subset V$ is an invariant subspace ($T(S) \subset S$) then S^\perp is also an invariant subspace ($T(S^\perp) \subset S^\perp$)

Pf: Let $v \in S^\perp$, then $\forall w \in S, T(w) \in S$, so $\langle Tv, w \rangle \stackrel{?}{=} \langle v, Tw \rangle \stackrel{?}{=} 0$.

Since $\langle Tv, w \rangle = 0 \quad \forall w \in S$, we get: $Tv \in S^\perp$. ($T^* = T$) ($v \in S^\perp, Tw \in S$) \square

Theorem (the spectral theorem for real self-adjoint operators)

If $T: V \rightarrow V$ is self-adjoint then T is diagonalizable, with real eigenvalues.

Even more, T can be diagonalized in an orthonormal basis of $(V, \langle \cdot, \cdot \rangle)$!