

5 dual space

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Def. 10.1 Given a vector space E , the vector space $E^* = \text{Hom}(E, K)$ of linear maps from E to the field K is called the **dual space (or dual)** of E .
The linear maps in E^* are called **linear forms**, or **covectors**.
The dual space E^{**} of the space E^* is the **bidual** of E .

Notation: Linear forms $f: E \rightarrow K$ will be denoted with a star, e.g. u^* , x^* .

$\forall f^* \in E^*$, if (u_1, \dots, u_n) is a basis of E , $\forall x = x_1 u_1 + \dots + x_n u_n \in E$,

$$f^*(x) = f^*(u_1)x_1 + \dots + f^*(u_n)x_n \\ \equiv c_1 x_1 + \dots + c_n x_n$$

Therefore, f^* can be represented by

$$f^*(x) = (c_1 \dots c_n) \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = c^T x$$

Def. 10.2 Given a vector space E and any basis $(u_i)_{i \in I}$ for E , for every $i \in I$, there is a unique linear form u_i^* such that

$$u_i^*(u_j) = \begin{cases} 1, & \text{if } i=j \\ 0, & \text{if } i \neq j \end{cases} \quad (\text{Kronecker delta } \delta_{ij}.)$$

u_i^* is called the **coordinate form** of index i w.r.t. the basis $(u_i)_{i \in I}$.

It's called the **coordinate form** because if E is finite dimensional,

$$\forall v = d_1 u_1 + \dots + d_n u_n, \quad u_i^*(v) = d_i.$$

So u_i^* is the linear function that returns the i th coordinate.

Thm 2.3/2.20 Let E be a vector space of $\dim n$. For every basis (u_1, \dots, u_n) of E , the family of coordinate

Thm 2.3/2.20 Let E be a vector space. For every basis (u_1, \dots, u_n) of E , the family of coordinate forms (u_1^*, \dots, u_n^*) is a basis of E^* , called the **dual basis** of (u_1, \dots, u_n) .

Aside: If $P = \begin{pmatrix} u_1 & \dots & u_n \end{pmatrix}$, then $P^{-1} = \begin{pmatrix} u_1^* \\ \vdots \\ u_n^* \end{pmatrix}$

Notation: Denote $u^*(v)$ by $\langle u^*, v \rangle$, a bilinear map

$\langle -, - \rangle : E^* \times E \rightarrow \mathbb{K}$ satisfying

$$\langle u_1^* + u_2^*, v \rangle = \langle u_1^*, v \rangle + \langle u_2^*, v \rangle$$

$$\langle u_1^*, v_1 + v_2 \rangle = \langle u_1^*, v_1 \rangle + \langle u_1^*, v_2 \rangle$$

$$\langle \lambda u^*, v \rangle = \lambda \langle u^*, v \rangle$$

$$\langle u^*, \lambda v \rangle = \lambda \langle u^*, v \rangle.$$

This bilinear map is the **canonical pairing** between E^* and E .

We also have the map $\text{eval}_v : E^* \rightarrow \mathbb{K}$, $\text{eval}_v(u^*) = \langle u^*, v \rangle$.

Def. 10.3 $v \in E$ and $u^* \in E^*$ are **orthogonal** iff $\langle u^*, v \rangle = 0$.

Two subspaces V of E and U of E^* are orthogonal iff

$$\langle u^*, v \rangle = 0 \quad \forall u^* \in U \text{ and } v \in V.$$

The **orthogonal** V° of V is the subspace V° of E^*

defined s.t. $V^\circ = \{ u^* \in E^* \mid \langle u^*, v \rangle = 0 \quad \forall v \in V \}$ (also called annihilator of V)

The **orthogonal** U° of U is

$$U^\circ = \{ v \in E \mid \langle u^*, v \rangle = 0 \quad \forall u^* \in U \}$$

Although we've discussed all of these in the context of general vector spaces, we can specialize to Euclidean spaces.

Def. 11.1 A Euclidean space is a real vector space E equipped

vector spaces, we can specialize to

Def. 11.1 A Euclidean space is a real vector space E equipped with a symmetric bilinear form $\varphi: E \times E \rightarrow \mathbb{R}$ that is positive definite.

Notation: $\varphi(u, v) = u \cdot v = \langle u, v \rangle = (u | v)$
 $\|u\|^2 = \varphi(u, u) = \|u\|^2$.

Def. 11.2 Given a Euclidean space E , $u, v \in E$ are orthogonal/perpendicular iff $u \cdot v = 0$. Given a family $(u_i)_{i \in I}$ of vectors in E , $(u_i)_{i \in I}$ is orthogonal iff $u_i \cdot u_j = 0$ iff $i \neq j$. $(u_i)_{i \in I}$ is orthonormal iff in addition $\|u_i\| = u_i \cdot u_i = 1 \quad \forall i \in I$. For any subset F of E , the set $F^\perp = \{v \in E \mid u \cdot v = 0, \forall u \in F\}$ of all vectors orthogonal to all vectors in F , is called the orthogonal complement of F .

Def. 11.3 For any $u \in E$, let $\varphi_u = E \rightarrow \mathbb{R}$ be the map defined s.t.
 $\varphi_u(v) = u \cdot v, \quad \forall v \in E$.

Note that φ_u is a linear form in E^* .

Thm 11.1 / 11.6 The map $b: E \rightarrow E^*$ defined s.t.
 $b(u) = \varphi_u$.

is linear and injective. When E is finite dim., b is a canonical isomorphism.

proof. b is linear because \langle, \rangle is bilinear.


If $\varphi_u = \varphi_v$, then $\varphi_u(w) = \varphi_v(w) \quad \forall w \in E$

$\Rightarrow u \cdot w = v \cdot w \quad \forall w \in E$

$\Rightarrow (v - u) \cdot w = 0$

$\Rightarrow u = v$, because \langle, \rangle is pos. def.

$\Rightarrow b$ is injective

If $\dim(E) = n < \infty$, $\dim(E) = \dim(E^*)$, so b is bijective. 

Notation: The inverse of the isomorphism $b: E \rightarrow E^*$ is denoted $\# = E^* \rightarrow E$.

Corollary 11.1/11.7 If $\dim(E) = n < \infty$, every linear form $f \in E^*$ corresponds to a unique $u \in E$ s.t.
$$f(v) = u \cdot v, \quad \forall v \in E.$$

If f is not the 0-form, then $\text{Ker } f$ is a hyperplane of all vectors orthogonal to u .

Note: b is not surjective if E has infinite dim.

Can be salvaged by looking only at continuous linear maps in a Hilbert space (complete normed space)
(Riesz representation theorem)

Prop. 11.6/11.8 If $\dim(E) = n < \infty$, for every linear map $f: E \rightarrow E$, there is a unique linear map $f^*: E \rightarrow E$ s.t.
$$f^*(u) \cdot v = u \cdot f(v) \quad \forall u, v \in E.$$

proof. Suppose $a_1 \cdot v = u \cdot f(v)$ and $a_2 \cdot v = u \cdot f(v)$
 $\Rightarrow a_1 \cdot v = a_2 \cdot v$
 $\Rightarrow (a_1 - a_2) \cdot v = 0$
 $\Rightarrow a_1 = a_2$ or $v = 0$.

But since v is any element of V , $\Rightarrow a_1 = a_2$.

This implies that $f^*(u) = a_1$ is uniquely defined, so we just need linearity.

Given $u_1, u_2 \in E$, $(u_1 + u_2) \cdot f(v) = u_1 \cdot f(v) + u_2 \cdot f(v)$, $\forall v \in E$.

and $(f^*(u_1) + f^*(u_2)) \cdot v = f^*(u_1) \cdot v + f^*(u_2) \cdot v$ $\forall v \in E$.

By assumption, $f^*(u_1) \cdot v = u_1 \cdot f(v)$ and $f^*(u_2) \cdot v = u_2 \cdot f(v)$, $\forall v \in E$.

$\Rightarrow (f^*(u_1) + f^*(u_2)) \cdot v = (u_1 + u_2) \cdot f(v) = f^*(u_1 + u_2) \cdot v$, $\forall v \in E$.

Since b is bijective $\Rightarrow f^*(u_1 + u_2) = f^*(u_1) + f^*(u_2)$.

Similarly, $\forall v, f^*(\lambda u) \cdot v = \lambda u \cdot f(v)$

$$\begin{aligned}
 &= \lambda(u \cdot f(v)) \\
 &= \lambda(f^*(u) \cdot v) \\
 &= \lambda f^*(u) \cdot v
 \end{aligned}$$

$\Rightarrow f^*(\lambda u) = \lambda f^*(u)$

Def. 11.4 If $\dim(E) = n < \infty$, \forall linear map $f: E \rightarrow E$, the unique linear map $f^*: E \rightarrow E$ s.t.

$f^*(u) \cdot v = u \cdot f(v), \forall u, v \in E$

is called the *adjoint* of f (w.r.t. the inner prod.). Linear maps $f: E \rightarrow E$ s.t. $f = f^*$ are called *self-adjoint*.

Definition

Given Euclidean spaces $(E, \langle \cdot, \cdot \rangle_1)$ and $(F, \langle \cdot, \cdot \rangle_2)$, and a linear map $f: E \rightarrow F$, we can similarly show that there exists a unique linear map $f^*: F \rightarrow E$ s.t.

$\langle f(u), v \rangle_2 = \langle u, f^*(v) \rangle_1, \forall u \in E, v \in F.$

f^* is called the *adjoint* of f .

Properties

(1) $f^{**} = f$

(2) $(f+g)^* = f^* + g^*$

$(\lambda f)^* = \lambda f^*$

(3) $(g \circ f)^* = f^* \circ g^*$.

Prop. 11.7 / 11.9 Given any nontrivial Euclidean space E of finite dim. $n \geq 1$, \exists an orthonormal basis (u_1, \dots, u_n) for E .

proof. Induction on n . Base case $n=1$: take any $v \neq 0$ and let $u_1 = \frac{v}{\|v\|}$.

For $n \geq 2$, start with $u_1 = \frac{v}{\|v\|}$ for $v \neq 0$.

Let $H = \{u_1\}^\perp$, the hyperplane that is the orthogonal complement of $\{u_1\}$.

Let $H = \{u_1\}^\perp$, the hyperplane that is the orthogonal complement of $\{u_1\}$.
 $\dim(H) = n - 1$ because $H = \text{Ker } \varphi_{u_1}$, where φ_{u_1} is the linear form associated with u_1 .

Thus, by the inductive hypo, H has an orthonormal basis (u_2, \dots, u_n) .

Then (u_1, \dots, u_n) is an orthonormal basis of $E = H \oplus \mathbb{R}u_1$. □

Prop. 11.8/11.10 Given any nontrivial Euclidean space of finite dim $n \geq 1$, from any basis (e_1, \dots, e_n) for E , we can construct an orthonormal basis (u_1, \dots, u_n) s.t. $\forall k$, $\text{span}\{e_1, \dots, e_k\} = \text{span}\{u_1, \dots, u_k\}$.

proof sketch Let $v_k = e_k - \sum_{i=1}^{k-1} \varphi_{u_i}(e_k)u_i$, $u_k = \frac{v_k}{\|v_k\|}$.

We remove from e_k its projection onto $\text{span}\{u_1, \dots, u_{k-1}\} = \text{span}\{e_1, \dots, e_{k-1}\}$. □

This is called Gram-Schmidt orthonormalization.

Prop. 11.9/11.11 Given any nontrivial Euclidean space of finite dim $n \geq 1$, for any subspace F of dim k , the orthogonal complement F^\perp has dim $n - k$ and $E = F \oplus F^\perp$. Furthermore $F^{\perp\perp} = F$.

proof sketch Start with an orthonormal basis on F , extend to a basis on E , then orthonormalize it to get an orthonormal basis (u_1, \dots, u_n) on E where $\text{span}\{u_1, \dots, u_k\} = F$
 $\text{span}\{u_{k+1}, \dots, u_n\} = F^\perp$. □

Def. 11.5 Given any two nontrivial Euclidean spaces E and F of the same finite dim n , a function $f: E \rightarrow F$ is an orthogonal transformation, or linear isometry, if it is linear and $\|f(u)\| = \|u\|$ for all $u \in E$.

Prop. 11.10/11.12 Given any two nontrivial Euclidean spaces E and F of the same finite dim n , $\forall f: E \rightarrow F$, the following are equiv:
 (i) f is a linear map and $\|f(u)\| = \|u\| \quad \forall u \in E$

the same finite dim n , $\forall f: E \rightarrow F$, the following are equiv:

(1) f is a linear map and $\|f(u)\| = \|u\|$, $\forall u \in E$

(2) $\|f(v) - f(u)\| = \|v - u\|$, $\forall u, v \in E$ and $f(0) = 0$.

(3) $f(u) \cdot f(v) = u \cdot v$ $\forall u, v \in E$.

Furthermore, such a map is bijective.

proof. (1) \rightarrow (2): $\|f(v) - f(u)\| = \|f(v - u)\| = \|v - u\|$.
 \uparrow linearity

let's prove something slightly more general

(2) \rightarrow (3): Claim = If $\|f(v) - f(u)\| = \|v - u\|$ $\forall u, v \in E$, then $\forall \tau \in E$, the function $g: E \rightarrow F$ defined by $g(u) = f(\tau + u) - f(\tau)$ $\forall u \in E$ is a linear map s.t. $g(0) = 0$ and $g(u) \cdot g(v) = u \cdot v$.

proof. Clearly, $g(0) = f(\tau) - f(\tau) = 0$.

$$\begin{aligned} \|g(v) - g(u)\| &= \|f(\tau + v) - f(\tau) - f(\tau + u) + f(\tau)\| \\ &= \|f(\tau + v) - f(\tau + u)\| \\ &= \|\tau + v - (\tau + u)\| = \|v - u\| \quad \forall u, v \in E. \end{aligned}$$

Also, setting $u=0$, $\|g(v)\| = \|v\|$, so g preserves norms and distances.

Further, $\|u - v\|^2 = \langle u - v, u - v \rangle = \|u\|^2 + \|v\|^2 + 2u \cdot v$

$$\Rightarrow 2u \cdot v = \|u\|^2 + \|v\|^2 - \|u - v\|^2$$

Thus $2g(u) \cdot g(v) = \|g(u)\|^2 + \|g(v)\|^2 - \|g(u) - g(v)\|^2$

$$= \|u\|^2 + \|v\|^2 - \|u - v\|^2$$

$$= 2u \cdot v$$

$$\Rightarrow g(u) \cdot g(v) = u \cdot v \quad \forall u, v \in E. \quad \square$$

If we let $\tau = 0$, then $g = f$, showing that $f(u) \cdot f(v) = u \cdot v$.

(3) \rightarrow (1): $f(u) \cdot f(u) = u \cdot u$

$$\Rightarrow \|f(u)\|^2 = \|u\|^2 \quad \forall u \in E.$$

$$\Rightarrow \|f(u)\| = \|u\|$$

Now just need to show f is linear.

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Now just need to show f is isometry.

Let (e_1, \dots, e_n) be an orthonormal basis of E .

$$\text{So } u = \sum_{i=1}^n (u \cdot e_i) e_i = \sum_{i=1}^n u_i e_i \quad \text{for } u \in E.$$


Then $(f(e_1), \dots, f(e_n))$ is another orthonormal basis of E .

$$\begin{aligned} \text{Then } f(u) &= \sum_{i=1}^n (f(u) \cdot f(e_i)) f(e_i) \\ &= \sum_{i=1}^n (u \cdot e_i) f(e_i) = \sum_{i=1}^n u_i f(e_i). \end{aligned}$$

This proves that f is linear by using this basis.

Last: If $f(u) = f(v)$, then $f(v-u) = 0$, so $\|f(v-u)\| = 0$

$$\Rightarrow \|v-u\| = 0 \Rightarrow u = v.$$

Thus, f is injective, and since $\dim(E) = \dim(F) = n$, f is bijective. 

Prop. 11.11 / 11.13 Given any two nontrivial Euclidean spaces E and F of

the same finite dim n , $\forall f: E \rightarrow F$, if

$$\|f(v) - f(u)\| = \|v - u\| \quad \forall u, v \in E,$$

then f is an affine map, and its associated linear map g is an isometry.

proof. See Prop. 11.10 / 11.12 (2) \rightarrow (3) proof.

Prop. 11.12 / 11.14 Let E be any Euclidean space of finite dim n , and let $f: E \rightarrow E$ be any linear map. Then

(1) f is an isometry iff $f \circ f^* = f^* \circ f = \text{id}$.

(2) For any orthonormal basis (e_1, \dots, e_n) of E , if the matrix of f is A , then the matrix of f^* is A^T , and f is an isometry iff $AA^T = A^T A = I_n$.

iff columns of A form an orthonormal basis of \mathbb{R}^n .

iff columns of A form an orthonormal basis of \mathbb{R}^n
 iff rows of A form an orthonormal basis of \mathbb{R}^n .

proof.

$$\begin{aligned} (1) \quad & f(u) \cdot f(v) = u \cdot v && \text{by def. of isometry} \\ \Leftrightarrow & f^*(f(u)) \cdot v = u \cdot v && \text{by def. of adjoint} \\ \Leftrightarrow & f^*(f(u)) = u && \text{because true } \forall v. \\ \Leftrightarrow & f^* \circ f = \text{id}. \end{aligned}$$

But an endomorphism f in a finite-dim space with a left-inverse is an isomorphism, so $f^* \circ f = \text{id}$.

(2) If (e_1, \dots, e_n) is an orthonormal basis of E , let $A = (a_{ij})$ be the matrix of f and $B = (b_{ij})$ be the matrix of f^* .
 i.e. columns of A are the images of f under the basis.

$$A = \left(f(e_1) \quad \dots \quad f(e_n) \right) \quad B = \left(f^*(e_1) \quad \dots \quad f^*(e_n) \right)$$

Let $w = w_1 e_1 + \dots + w_n e_n \in E$. Then $w_k = w_i w$.

$$\text{Then } b_{ji} = f^*(e_i) \cdot e_j = e_i \cdot f(e_j) = a_{ij}$$

$$\Rightarrow B = A^T$$

$$\text{Then } A^T A = \begin{pmatrix} f(e_1)^T \\ \vdots \\ f(e_n)^T \end{pmatrix} \begin{pmatrix} f(e_1) & \dots & f(e_n) \end{pmatrix} = \begin{pmatrix} f(e_1) \cdot f(e_1) & f(e_1) \cdot f(e_2) & \dots & f(e_1) \cdot f(e_n) \\ \vdots & \ddots & \ddots & \vdots \\ f(e_n) \cdot f(e_1) & \dots & \dots & f(e_n) \cdot f(e_n) \end{pmatrix}$$

$\left(\text{But } (f(e_i))_{i \in I} \text{ form another orthonormal basis because } f \text{ preserves inner prod.} \right) = I_n$.

$$\text{Similarly, } (AA^T)_{ij} = f^*(e_i) \cdot f^*(e_j) = \delta_{ij}, \text{ so } AA^T = I_n,$$

and we also get columns of A^T are orthonormal. ☑

Def. 11.6

A real $n \times n$ matrix is an orthogonal matrix if

$$A A^T = A^T A = I$$

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(Equivalent to $A^T = A^{-1}$)