

Thm if V is fin dim and $T : V$ to W is linear,
then $\dim V = \dim \ker(T) + \dim \operatorname{im}(T)$

Pf Outline

pick basis $\{w_1, \dots, w_r\}$ for $\operatorname{im}(T)$

pick u_1, \dots, u_r s.t. $T(u_i) = w_i$ for all i

let $U = \operatorname{span}(u_1, \dots, u_r)$

pick basis v_1, \dots, v_k for $\ker(T)$

I) show that $\ker(T) + U = V$

II) show that $\ker(T) + U$ is a direct sum

then $\dim V = \dim \ker(T) + \dim U$
 $= \dim \ker(T) + \dim \operatorname{im}(T) \quad \square$

Pf of I) pick v in V

know $T(v) = \sum_i b_{iw_i}$ for some b_i

so $T(v) = \sum_i b_i T(u_i) = T(\sum_i b_i u_i)$

so $T(v - \sum_i b_i u_i) = \mathbf{0}_W$

so $v - \sum_i b_i u_i$ in $\ker(T)$

so v in $\ker(T) + U$

Pf of II) recall: $W + U$ is a direct sum iff

$$W \cap U = \{\mathbf{0}\}$$

so want $\ker(T) \cap U = \{\mathbf{0}_V\}$

pick v in $\ker(T) \cap U$

since v in U , have $v = \sum_i a_i u_i$ for some a_i

since v in $\ker(T)$, have $\sum_i a_i w_i = T(v) = \mathbf{0}_W$

but $\{w_i\}_i$ lin. indep. so $a_i = 0$ for all i

assuming V, W are both finite-dimensional:

Cor if $T : V$ to W is linear and $\dim V > \dim W$,
then T is not injective

Pf $\dim \operatorname{im}(T) \leq \dim W < \dim V$,
so
 $\dim \ker(T) = \dim V - \dim \operatorname{im}(T) > 0$

Cor if $T : V$ to W is linear and $\dim V < \dim W$,
then T is not surjective

Pf exercise

Cor if $T : V$ to W is linear and bijective,
then $\dim V = \dim W$ [converse is false!]

(Axler §3D)

Thm TFAE for a linear map $T : V$ to W :

- 1) T is bijective
- 2) T takes any basis for V onto a basis for W
- 3) T takes some basis for V onto one for W
- 4) there is a linear map $S : V$ to W s.t.
 $S(T(v)) = v$ and $T(S(w)) = w$

Df in the situation above, we say that T is
a linear isomorphism from V onto W

Pf of Thm not hard to show that
2) implies 3) implies 4) implies 1)
remains to show 1) implies 2)

so suppose $\{e_i\}_i$ is a basis for V

claim that $\{T(e_i)\}_i$ spans W :

for all w in W , have $w = T(v)$ for some v in V

by surjectivity of T

$v = \sum_i a_i e_i$ for some a_i

so $w = \sum_i a_i T(e_i)$

claim that $\{T(e_i)\}_i$ is lin. indep.

suppose $\sum_i b_i T(e_i) = \mathbf{0}_W$

then $T(\sum_i b_i e_i) = \mathbf{0}_W$

so $\sum_i b_i e_i$ in $\ker(T)$

so $\sum_i b_i e_i = \mathbf{0}_V$

by injectivity of T

so $b_i = 0$ for all i

by lin. independence of $\{e_i\}_i$ \square

(Axler §3C) recap:

Slogan #1 if we know a basis for V
then a linear map V to W is det by
 where it sends the basis
 and any choices will do

in particular, a linear map F^n to W is det by
 where it sends $(1, 0, 0, \dots)$, $(0, 1, 0, \dots)$, \dots

Slogan #2 a linear isomorphism is
 a linear map taking bases to bases

in particular, a linear iso F^n to W is det by
 an ordered basis for W
 [images of $(1, 0, 0, \dots)$, $(0, 1, 0, \dots)$, \dots]

[in this sense, linear maps out of F^n are easy]

[what else can we do with linear maps?]

a composition of linear maps is linear: given linear

$$A : V \text{ to } W,$$

$$B : W \text{ to } U,$$

$$B(A(v + v')) = B(A(v) + A(v')) = B(A(v)) + B(A(v')),$$

$$B(A(c \cdot v)) = B(c \cdot A(v)) = c \cdot B(A(v))$$

so $B \circ A : V \text{ to } U$ is also linear

suppose $V = F^n$ with std basis (v_1, \dots, v_n) ,

$W = F^m$ with std basis (w_1, \dots, w_m) ,

$U = F^\ell$ with std basis (u_1, \dots, u_ℓ)

$$A(v_i) = \sum_j a_{\{j,i\}} w_j,$$

$$B(w_j) = \sum_k b_{\{k,j\}} u_k$$

$$B(A(v_i)) = B(\sum_j a_{\{j,i\}} w_j)$$

$$= \sum_j a_{\{j,i\}} B(w_j)$$

$$= \sum_j a_{\{j,i\}} \sum_k b_{\{k,j\}} u_k$$

$$= \sum_{\{k, j\}} a_{\{j,i\}} b_{\{k,j\}} u_k$$

$$= \sum_k c_{\{k,i\}} u_k$$

$$\text{where } c_{\{k,i\}} = \sum_j a_{\{j,i\}} b_{\{k,j\}}$$

matrix multiplication = shorthand for these calc's

Df

matrix of A wrt the ordered bases

$(v_i)_{\{i=1\}^n}, (w_j)_{\{j=1\}^m}$:

$a_{11} \ a_{12} \ \dots \ a_{1n}$ # rows is n (dim domain),

$a_{21} \ a_{22} \ \dots \ a_{2n}$ # cols is m (dim target)

$\dots \ \dots \ \dots$

$a_{m1} \ a_{m2} \ \dots \ a_{mn}$

henceforth write F^n , F^m , etc in column notation:

$$v = \sum_i c_i v_i: \begin{matrix} c_1 \\ c_2 \\ \dots \\ c_n \end{matrix}$$

matrix \times vector rule for Av :

$$\begin{matrix} a_{11} & \dots & a_{1n} \\ a_{21} & \dots & a_{2n} \\ \dots & & \dots \\ a_{m1} & \dots & a_{mn} \end{matrix} \begin{matrix} c_1 \\ c_2 \\ \dots \\ c_n \end{matrix} = \begin{matrix} a_{j1} c_1 + \dots + a_{jn} c_n \\ \dots \\ \dots \end{matrix}$$

e.g., $Av_i: \begin{matrix} a_{1i} \\ a_{2i} \\ \dots \\ a_{mi} \end{matrix}$ (ith matrix col)

matrix \times matrix rule for $B \circ A$:

$$\begin{matrix} b_{11} & \dots & b_{1m} \\ b_{21} & \dots & b_{2m} \\ \dots & & \dots \\ b_{l1} & \dots & b_{lm} \end{matrix} \begin{matrix} a_{11} & \dots & a_{1n} \\ a_{21} & \dots & a_{2n} \\ \dots & & \dots \\ a_{m1} & \dots & a_{mn} \end{matrix}$$

$$= \begin{matrix} \dots & \dots & \dots \\ \dots & \sum_j a_{\{j,i\}} b_{\{kj\}} & \dots \\ \dots & \dots & \dots \\ \dots & \dots & \dots \end{matrix}$$

so vectors become columns,
linear maps become matrices,
 \circ matrix multiplication

if A iso, then $\dim V = \dim W$: so matrix is square