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

Pf Outline

II) ker(T) + U = V

pick basis {w_1, ..., w_r} for im(T) pick u_1, ..., u_r s.t. T(v_i) = w_i for all i let U = span(u_1, ..., u_r) pick basis v_1, ..., v_k for ker(T) l) {u 1, ..., u_r} is a basis for U

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

II) ker(T) + U is a direct sum

Pf of I) if sum_i a_iu_i = 0, then $sum_i a_i T(u_i) = 0 \text{ so a}_i = 0 \text{ for all } i$

Pf of II) pick v in V

know $T(v) = sum_i b_iw_i$ for some b_i so $T(v) = sum_i b_iT(u_i) = T(sum_i b_iu_i)$ so $T(v - sum_i b_iu_i) = \mathbf{0}_{\mathbf{W}}$ so $v - sum_i b_iu_i$ in ker(T)so v in ker(T) + U

Pf of III) recall: W + U is a direct sum iff W cap $U = \{0\}$ so want ker(T) cap $U = \{0_V\}$

pick v in ker(T) cap U since v in U, have v = sum_i a_iu_i for some a_i since v in ker(T), have sum_i a_iw_i = T(v) = **0_W** but {w_i}_i lin. indep. so a_i = 0 for all i

assuming	V, W	are both	finite-dimensional:
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<u>Cor</u> if T: V to W is linear and dim V > dim W, then [?] T is not injective

 $\dim \operatorname{im}(T) \le \dim W < \dim V$, so

 $\dim \ker(T) = \dim V - \dim \operatorname{im}(T) > 0$

<u>Cor</u> if T: V to W is linear and dim V < dim W, then T is not surjective

exercise

<u>Pf</u>

Pf

<u>Cor</u> if T: V to W is linear and bijective, then [?] dim V = dim W [converse false!] (Axler §3D)

<u>Thm</u> 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 <u>linear</u> map S: V to W s.t.

S(T(v)) = v and T(S(w)) = w

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

Pf of Thm not hard to show that2) implies 3) implies 4) implies 1)remains to show 1) implies 2)

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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 ia ie i for some a i
     so w = sum_i a_iT(e_i)
claim that {T(e_i)}_i is lin. indep.
     suppose sum_i b_i T(e_i) = 0_W
     then T(sum_i b_ie_i) = 0_W
     so sum_i b_ie_i in ker(T)
     so sum_i b_ie_i = 0_V
          by injectivity of T
     so b i = 0 for all i
          by lin. independence of {e_i}_i \( \sigma\)
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(Axler §3C) [recap:]
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Slogan #1 [if we know a basis for V, then]
a linear map V to W is det by
where it sends [the] a basis
and any choices will do

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

 \underline{Ex} there's a linear map from F^4 to F[x] sending the ith std basis vec to x^{i - 1}

it <u>restricts</u> to a linear iso from F⁴ to P₃ where P₃ = { $p \mid p = 0$ or deg $p \le 3$ }

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[what else can we do with linear maps?]
                                                        B(A(v_i)) = B(sum_i a_{i,i})
                                                                  = sum_j a_{j,i}B(w_j)
a composition of linear maps is linear: given linear
                                                                  = sum_j a_{j,i} sum_k b_{k,j} u_k
    A: V to W.
                                                                  = sum_{k, j} a_{j, i}b_{k, j} u_k
     B: W to U,
                                                                  = sum_k c_{k,i} u_k
                                                                       where c \{k,i\} = sum i a \{i,i\}b \{k,i\}
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))
                                                        matrix multiplication = shorthand for these calc's
so B o A : V to U is also linear
                                                        Df
                                                                  matrix of A wrt the ordered bases
                                                                  (v_i)_{i = 1}^n, (w_j)_{j = 1}^m:
suppose V with ordered basis (v 1, ..., v n),
         W ...
                            (w 1, ..., w m),
                                                       a11 a12 ... a1n
                                                                                 # rows is n (dim V),
          U ...
                                  (u 1, ..., u {)
                                                        a21 a22 ... a2n
                                                                                 # cols is m (dim W)
         A(v_i) = sum_i a_{i,i}w_i,
         B(w_i) = sum_k b_{k,i}u_k
                                                        am1 am2 ...
                                                                       amn
```

henceforth write F^n, F^m, etc in column notation:			matrix × matrix rule for B ○ A:					
			c1	b11		b1m	a11	aın
	v = sum	_i c_iv_i:	c2	b21		b2m	a21	a2n
								•••
			cn	bl1		blm	am1	amn
matriy v	vector ru	lo for Ave						
a11	a1n	c1	•••		=	sum	n_j a_{j,i}b	_{kj}
a21	a2n	c2 =	aj1 c1 + + ajn cn					•••
			•••					
am1	amn	cn	•••					
				SO	vect	tors	become	columns,
			a1i		linea	ar maps		matrices,
e.g.,		Av_i:	a2i (ith matrix col)		0	•		matrix multiplication
			•••					
			ami					

Warning

if dim V = dim W, then matrix is square

e.g., if V = W

but even if V = W, the ordered bases (v_i)_i, (w_j)_j can still differ! i.e., rows and cols not indexed the same

Convention

if $V = F^n$ and $W = F^m$, then take the std bases

e.g., if V = W, then $(v_i)_i = std = (w_j)_j$