

BASES, COORDINATES,
INVERSES, SIMILARITY TRANSFORMS

vector space V $v_i \in V$

$\{v_i\}_{i=1}^n$ is basis $\{v_i\}_{i=1}^n$ is a basis for V

- lin ind
- span all of V

FUNDAMENTAL FACT:

any lin ind set of vectors

has fewer (or equal) vectors than

any spanning set

$\{v_i\}_{i=1}^m$ $v_i \in V$
lin ind.

$$\boxed{m \leq n}$$

$\{w_i\}_{i=1}^n$ $w_i \in V$
spanning

Completing a basis

can always add elements w_i to $\{v_i\}_{i=1}^m$

until $\{v_i\}_{i=1}^m \cup \{w_i\} \subset \{w_i\}_{i=1}^n$
basis

Steinitz
Exchange
LEMMA

BASIS:

- all bases have the same # of elements

Note: # of elements in a set = cardinality

$$\left[\begin{array}{l} \text{cardinality} \\ \text{of a basis} \\ \text{for } V \end{array} \right] = \underline{\text{dimension}} \text{ of } V$$

Standard basis for \mathbb{R}^n

$$\text{cols of } I = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

- "largest" (cardinality) lin ind. set
- "smallest" (cardinality) spanning set

Coordinates: representations of vectors wrt a basis.

$$P = [P_1 \dots P_n]$$

form a basis
for \mathbb{R}^n

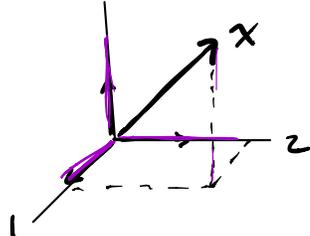
$$\begin{aligned} \vec{x} &= P \vec{x}' = [P_1 \dots P_n] \begin{bmatrix} x'_1 \\ \vdots \\ x'_n \end{bmatrix} \\ &= P_1 x'_1 + \dots + P_n x'_n \end{aligned}$$

"coords are coeffs"

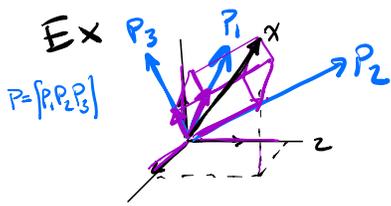
x' is the coordinates of x wrt (the cols of) P

Ex. standard basis

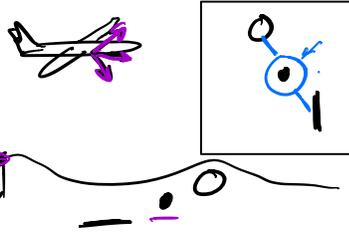
$$\mathbb{R}^3: P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \begin{aligned} x &= P x' \\ &= I x' \end{aligned}$$



$$x = x' = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix}$$



Ex. Aero camera on a drone



$$X = [P_1 P_2 P_3] \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \quad x' = \begin{bmatrix} 1/2 \\ 1/2 \\ 1/4 \end{bmatrix}$$

Construct \bar{x} as lin comb P_i 's

Ex $x = \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix}$ $P = \begin{bmatrix} P_1 & P_2 & P_3 \\ -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = [P_1 P_2 P_3]$

$$x = Px' \quad \checkmark \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} = \begin{bmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ -2 \\ 2 \end{bmatrix}$$

Basis cols of P :

- cols span the whole space "enough directions"
- lin ind. \rightarrow "no redundant directions"

\Rightarrow matrix P is invertible

"switch back & forth between x' coords & x (coords) without losing information"

$$\begin{aligned} x &= Px' \\ P^{-1}x &= P^{-1}Px' \\ P^{-1}x &= x' \end{aligned}$$

cols of P

- span $\mathbb{R}^n \rightarrow$ "every x has an x' "
- lin ind. \rightarrow ea. x has a unique x'

P is onto P is surjective
 P is one-to-one P is injective

can reach anywhere in the space

no redundant coords

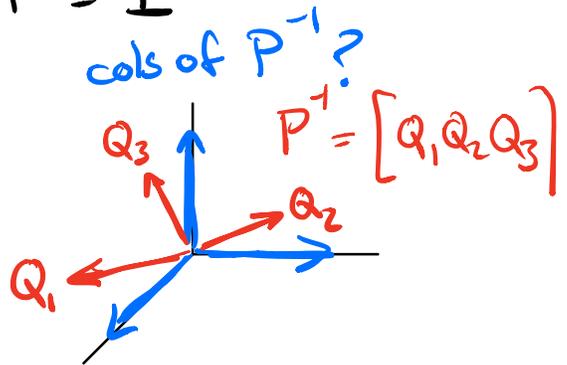
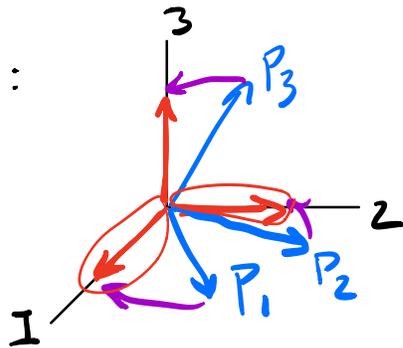
if P is both injective & surjective

$$= \text{one-to-one onto} = \text{bijective} = \text{invertible}$$

"reaches anywhere w/ unique coords"

For matrices $P^{-1}P = I = PP^{-1} = I$

BRIEF PICTURE:



FACTS ABOUT INVERSES

$P \in \mathbb{R}^{n \times n}$ P invertible
or $\mathbb{C}^{n \times n}$

Properties

- $(P^{-1})^{-1} = P$
- $(kP)^{-1} = \frac{1}{k}P^{-1}$
- $(PQ)^{-1} = Q^{-1}P^{-1}$ $Q \in \mathbb{C}^{n \times n}$
- $\det(P^{-1}) = \frac{1}{\det(P)}$
- $P^{-1} = \frac{1}{\det(P)} \text{Adj}(P)$ \rightarrow matrix of cofactors
"det of sub matrices" messy

Equivalent Properties P square

- P is invertible i.e. P^{-1} exists
- row reduce P to I
- col reduce P to I
- P is a product of elementary matrices
- P (square) and full row rank
- P (square) and full col rank
- cols of P are lin ind. (P square)
- rows of P are lin ind. (P square)
- $y = Px$ has a unique soln for eay

All of these statements are equivalent

\rightarrow Gaussian elimination (solve a system of eqns)

Wikipedia

- P has a trivial nullspace, $\text{null}(P) = \{0\}$
- $Px = 0 \Rightarrow x = 0$
- cols form a basis
- P^T is invertible
- 0 is not an eigenvalue of P
- $\det(P) \neq 0 \leftarrow$ "no dim collapse"
- $\exists Q$ s.t. $PQ = QP = I$ ($P^{-1} = Q$)
- P has a left & right inverse

Computational Inverse Facts

- 2x2 inverse

$$P = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \quad P^{-1} = \frac{1}{\det P} \text{Adj}(P) = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

only true
in 2x2 case

$$P^{-1} = \frac{1}{\det(P)} [\text{tr}(P)I - P]$$

- 3x3 $P^{-1} = \frac{1}{\det(P)} \left[\frac{1}{2} [\text{tr}(P)^2 - \text{tr}(P^2)]I - P \text{tr}(P) + P^2 \right]$

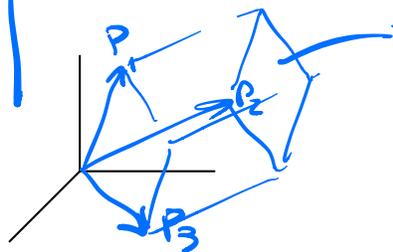
- $n \times n$ similar formulas ...

Side Note:

$\text{tr}(\cdot)$: sum of diagonal

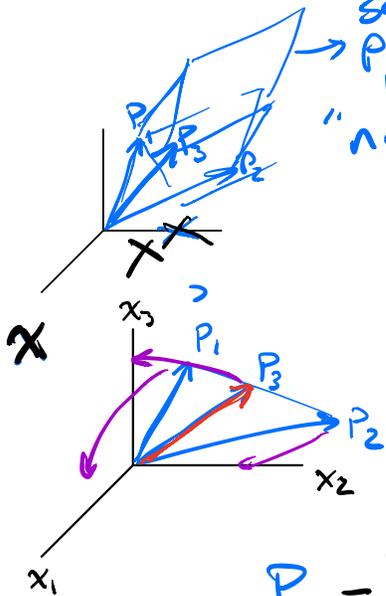
\det : (signed) volume of the parallel pippeel given by the cols of matrix

$$P = [P_1 P_2 P_3]$$



signed volume = \det
 "determinant becomes negative if you flip the volume inside out"

if $\det(P) = 0$

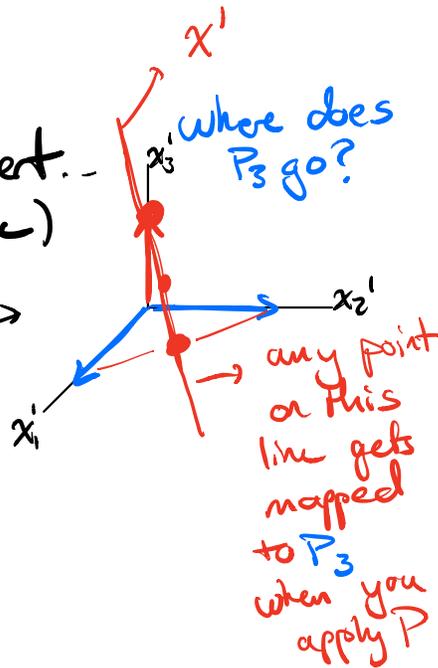


squashed parallel pippeel "no volume"

$$x = Px'$$

P
 try to invert... (impossible)

~~$$P^{-1}$$~~



where does P_3 go?

any point on this line gets mapped to P_3 when you apply P

$$P_3 = Px'$$

$$x' = P^{-1}P_3$$

Block Matrix Inversion:

$$\begin{aligned}
 \underline{P}^{-1} \underline{\begin{bmatrix} A & B \\ C & D \end{bmatrix}}^{-1} &= \underline{\begin{bmatrix} (A - BD^{-1}C)^{-1} & -(A - BD^{-1}C)^{-1}BD^{-1} \\ -D^{-1}C(A - BD^{-1}C)^{-1} & D^{-1} + D^{-1}C(A - BD^{-1}C)^{-1}BD^{-1} \end{bmatrix}}^* \\
 &= \underline{\begin{bmatrix} A^{-1} + A^{-1}B(D - CA^{-1}B)^{-1}CA^{-1} & -A^{-1}B(D - CA^{-1}B)^{-1} \\ -(D - CA^{-1}B)^{-1}CA^{-1} & (D - CA^{-1}B)^{-1} \end{bmatrix}}^*
 \end{aligned}$$

Caveat:

- D^{-1} exist and $(A - BD^{-1}C)^{-1}$ exist
- A^{-1} exist and $(D - CA^{-1}B)^{-1}$ exist OR

$\left. \begin{array}{l} (A - BD^{-1}C)^{-1} \\ (D - CA^{-1}B)^{-1} \end{array} \right\} \rightarrow$ Schur complements

SOURCE: $\begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} I & BD^{-1} \\ 0 & I \end{bmatrix} \begin{bmatrix} A - BD^{-1}C & 0 \\ 0 & D \end{bmatrix} \begin{bmatrix} I & 0 \\ D^{-1}C & I \end{bmatrix}^*$

$$\begin{aligned}
 P = \begin{matrix} n_1 & n_2 \\ \mathbb{R}^{n_1 \times n_1} & \mathbb{R}^{n_1 \times n_2} \\ \mathbb{R}^{n_2 \times n_1} & \mathbb{R}^{n_2 \times n_2} \end{matrix} \begin{bmatrix} A & B \\ C & D \end{bmatrix} &= \begin{bmatrix} I & 0 \\ CA^{-1} & I \end{bmatrix} \begin{bmatrix} A & 0 \\ 0 & D - CA^{-1}B \end{bmatrix} \begin{bmatrix} I & A^{-1}B \\ 0 & I \end{bmatrix}^* \\
 n_1 = n_2 \text{ not necessary} & \quad \text{easy to invert}
 \end{aligned}$$

$$\begin{bmatrix} A & B \\ \hline C & D \end{bmatrix} \quad \begin{bmatrix} A & B \\ \hline C & D \end{bmatrix}$$

Woodbury Matrix Identity

(specific case Sherman Morrison Formula)

$A \in \mathbb{R}^{n \times n}$

in general $(A+B)^{-1} \neq A^{-1} + B^{-1}$
 scalar $\frac{1}{x+y} \neq \frac{1}{x} + \frac{1}{y}$

$$(A + UCV)^{-1} = A^{-1} - \underbrace{A^{-1}U}_{\text{matrix}} \underbrace{(C^{-1} + VA^{-1}U)^{-1}}_{\text{add}} \underbrace{VA^{-1}}_{\text{inverse}} \star$$

U, C, V dimensions to match

UCV needs to be $n \times n$

A^{-1} needs to exist and $(C^{-1} + VA^{-1}U)^{-1}$ exists

$$\left(\begin{bmatrix} A \end{bmatrix} + \begin{bmatrix} u \end{bmatrix} \begin{bmatrix} c \end{bmatrix} \begin{bmatrix} v \end{bmatrix} \right)^{-1} = A^{-1} - \begin{bmatrix} \uparrow \\ A^{-1}u \end{bmatrix} \begin{bmatrix} \uparrow \\ (C^{-1} + VA^{-1}U)^{-1} \end{bmatrix} \begin{bmatrix} \uparrow \\ VA^{-1} \end{bmatrix}$$

addition is low rank

Sherman Morrison Formula { extreme case
 U col vector
 V row vector
 C scalar

$$\rightarrow \left(\begin{bmatrix} c \end{bmatrix} + \begin{bmatrix} v \end{bmatrix} \begin{bmatrix} A^{-1} \end{bmatrix} \begin{bmatrix} u \end{bmatrix} \right)^{-1}$$

trick behind

computationally efficient

Kalman filter

$(A + UCV)^{-1}$
 new covariance
 covariance C
 Kalman gain
 new measurement

Neumann series

$$\text{if } \lim_{n \rightarrow \infty} (I-A)^n = 0 \implies A^{-1} = \sum_{n=0}^{\infty} (I-A)^n$$

not on a test

matrix version of harmonic series

Derivative of Inverse:

$$P(t) \quad \boxed{\frac{dP^{-1}}{dt} = -P^{-1} \frac{dP}{dt} P^{-1}} \quad \leftarrow$$

$$\frac{d}{dt} P^{-1} P = \frac{d}{dt} \underbrace{I}_0 \implies \frac{dP^{-1}}{dt} P + P^{-1} \frac{dP}{dt} = 0$$

$$\implies \frac{dP^{-1}}{dt} = -P^{-1} \frac{dP}{dt} P^{-1}$$

Elementary Matrices & computing inverses

next time.

Bases for functions:

standard basis: $\underline{S(t)}$ t index. coordinates: $\underline{f(t)}$

Fourier basis: $\cos(\omega t)$
 $\sin(\omega t)$ coordinates: $F(\omega)$

$$\begin{matrix} P & P^{-1} \\ \rightarrow & \tilde{F}^{-1}(\cdot) \\ & \tilde{F}(\cdot) \end{matrix}$$