

Projections

1-dimensional
(onto a line)

$$P = C \underline{a}, \quad C = \frac{\underline{x}^T \underline{a}}{\underline{a}^T \underline{a}}$$

↑
projection
of \underline{x} onto \underline{a}

$$P = \frac{\underline{a} \underline{a}^T}{\underline{a}^T \underline{a}}$$

$$\underline{x}^T \underline{a}$$

$$\underline{a}$$

$$\underline{x} \text{ onto } \underline{a}$$

$$P = \frac{\underline{x}^T \underline{a}}{\underline{a}^T \underline{a}} \underline{a}$$

Onto Column
Space

Proj \underline{x} onto $\text{col}(A)$

$$P \underline{x} = A(A^T A)^{-1} A^T \underline{x}$$

① Find $A^T \underline{x}$

② Find $A^T A$

③ Find $(A^T A)^{-1} A^T \underline{x}$

Constructing
Orthonormal Basis

Vectors

1) Make $\underline{q}_1, \underline{q}_2, \dots$
using Gram-Schmidt

2) check $\underline{q}_1^T \underline{q}_2 = 0$

Solve linear system or
construct inverse directly

i) $(A^T A)^{-1} A^T \underline{x} = \underline{y}$

ii) $A^T \underline{x} = A^T A \underline{y}$

iii) $[A^T A] \underline{y} = \underline{x}$

iv) $\begin{array}{c|c} A^T A & A^T \underline{x} \end{array}$

ii) if $A^T A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$

$$= \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

Factorize
 $A = QR$

1) given

$$A = [a_1, a_2, \dots]$$

$$Q = [q_1, q_2, \dots] \leftarrow \text{from G-S}$$

2) find R s.e.

$$A = QR$$

3) $R = \begin{bmatrix} q_1^T a_1 & q_1^T a_2 \\ 0 & q_2^T a_2 \end{bmatrix}$

↑ Invertible
upper triangular

Gram-Schmidt

1) given a_1, a_2, a_3

2) $b_1 = a_1$

$b_2 = a_2 - \text{Proj}(a_2 \text{ onto } b_1)$

$b_3 = a_3 - \text{Proj}(a_3 \text{ onto } b_2) - \text{Proj}(a_3 \text{ onto } b_1)$

Make
orthogonal

normalize

$$3) \vec{q}_1 = \frac{\vec{b}_1}{\|\vec{b}_1\|}$$

$$\vec{q}_3 = \frac{\vec{b}_3}{\|\vec{b}_3\|}$$

$$\vec{q}_2 = \frac{\vec{b}_2}{\|\vec{b}_2\|}$$

Checking Linearity of a Transformation

Transformations
and their
Matrices

To disprove,

► show $T(kx) \neq kT(x)$

► show $T(x+y) \neq T(x)+T(y)$

To Prove,

AND ► show $T(kx) = kT(x)$ (Scalar multiplication)

► show $T(x+y) = T(x)+T(y)$ (Vector addition)

OR ► show $T(ax+by) = aT(x)+bT(y)$

Standard Mapping

of a
Transformation

$$A = [b_1 \ b_2 \ b_3 \dots]$$

where $\{b_1, \dots, b_n\}$

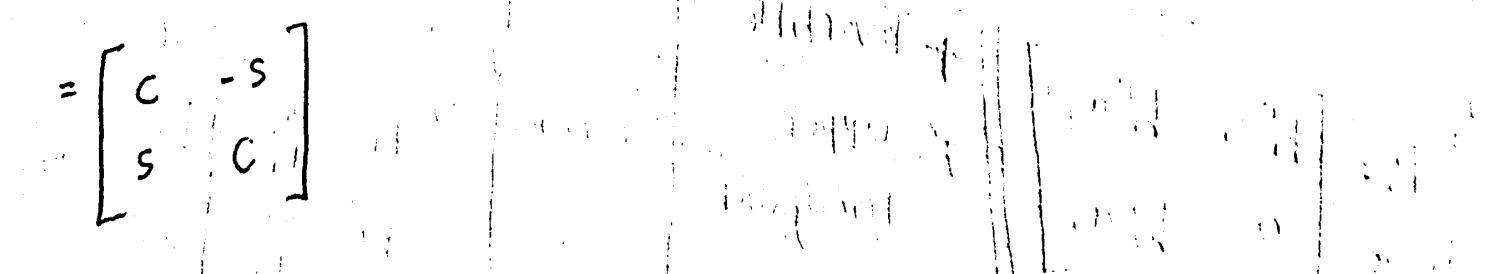
are the standard basis
vectors fed through the
transformation

Rotation Matrices

to rotate θ ,

$$Q_\theta = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$$

$$= \begin{bmatrix} c & -s \\ s & c \end{bmatrix}$$



Change of Basis Matrix

To change from basis V

to basis W , coB matrix
is

$$WV^{-1}$$

Order of Transformation Matrices

[third] [second] [first] target

Normal Equations

$$A^T A \hat{x} = A^T B$$

wooo

LLG (1-D)

$$\hat{x} = \frac{a^T b}{a^T a}$$

Props, II

onto

transformation matrix
A has a pivot in every column

one-to-one

transformation matrix
A has a pivot in every row

Superposition of orthogonal projections

if $\underline{a} \perp \underline{b}$, then $P_1 \underline{x} = \text{proj}_{\underline{a}} \underline{x}$

$$P_2 \underline{x} = \text{proj}_{\underline{b}} \underline{x}$$

and $\text{proj}_{\text{onto } \text{span}\{\underline{a}, \underline{b}\}} \underline{x} = P_1 \underline{x} + P_2 \underline{x}$

$$= P_1 \underline{x} + P_2 \underline{x}$$

Mutual orthogonality implies Linear Independence

LLS

(multivar)

① Construct A:

$$A = \begin{bmatrix} 1 & t_{\min} \\ 1 & \vdots \\ 1 & t_{\max} \end{bmatrix}$$

or

$$B = \begin{bmatrix} x(t_{\min}) \\ \vdots \\ x(t_{\max}) \end{bmatrix}$$

① Construct B:

$$A = \begin{bmatrix} x_1 & | & \\ \vdots & | & \\ x_n & | & \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} y(x_1) \\ \vdots \\ y(x_n) \end{bmatrix}$$

$$\hat{x} = \begin{bmatrix} m \\ b \end{bmatrix}$$

$$B = \begin{bmatrix} y(x_1) \\ \vdots \\ y(x_n) \end{bmatrix}$$

② Plug into

$$\hat{x} = (A^T A)^{-1} A^T B$$

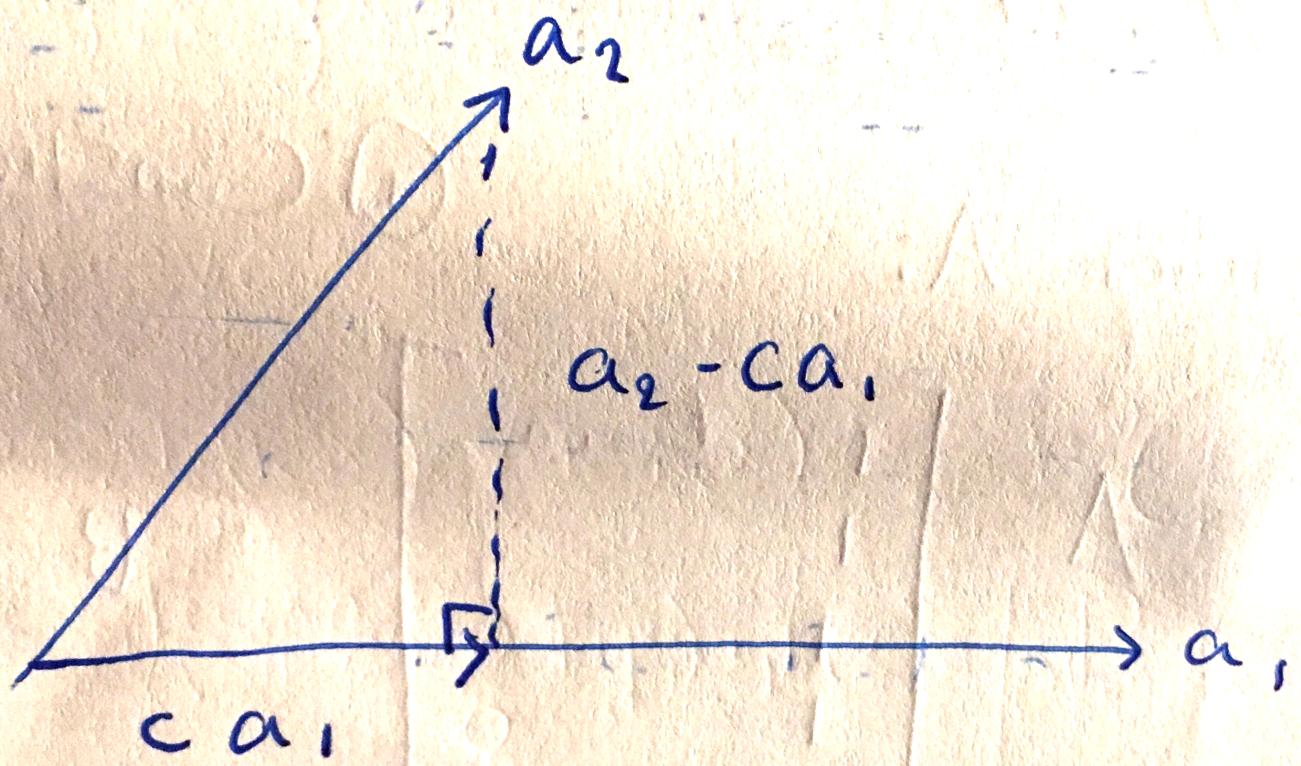
i) find $A^T B$

ii) find $A^T A$

iii) find $(A^T A)^{-1}$

③ Projection (optional)

$$\text{Proj} = A \hat{x}$$



PROPS

✓ ↗

Orthogonal Subspaces

Two subspaces $V + W$ are orthogonal if

$v^T w = 0$ for all $v \in V$ and all $w \in W$

Proj Matrix Props

$$\textcircled{1} P^T = P$$

$$\textcircled{2} P^2 = P$$

Fundamental Thm of Linear Algebra

$$\text{Null}(A) \perp \text{Col}(A^T)$$

$$\text{Null}(A^T) \perp \text{Col}(A)$$

where $\frac{1}{c} = \text{ortho complement}$

Types of Linear Transformations

① stretches

$$A = \begin{bmatrix} c & 0 \\ 0 & c \end{bmatrix}$$

② rotation

$$A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

e.g.

$$(x, y) \rightarrow (-y, x)$$

③ reflection

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

e.g.
over $y=x$

④ projection

$$P = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

e.g.

(covered extensively elsewhere)

$T(\underline{b}_i)$ is known for all vectors \underline{b}_i :

if Ax is known for every vector in a basis, then we know Ax for all x .

If $\underline{x} = c_1 \underline{x}_1 + \dots + c_n \underline{x}_n$ then $A\underline{x} = c_1(A\underline{x}_1) + \dots + c_n(A\underline{x}_n)$

basis vectors

basis vectors

Orthogonality

v, w are orthogonal if

$$v^T w = 0$$

Orthogonal Complement

given $V \subseteq \mathbb{R}^n$

the

Orthogonal Complement

V is the space of all vectors orthogonal to V .

Linear Transformation

any transformation whose transformation matrix A satisfies

$$A(cx + dy)$$

$$= cAx + dAy$$

is linear.

Linear T's when

known for all

vectors \underline{b}_i :

if Ax is known for every vector in a basis,

$$(T(x))$$

then we know Ax for all x .

1D proj

$$P = \frac{aa^T}{a^T a} \quad \square$$

Col(A) Proj

$$P = A(A^T A)^{-1} A^T B$$

$$P = A(A^T A)^{-1} A^T X$$

① find $A^T X$

$$② A^T A$$

$$③ (A^T A)^{-1}$$

1D
 $P = \frac{aa^T}{a^T a}$

Col(A)

$$P = A(A^T A)^{-1} A^T$$

$$P = A(A^T A)^{-1} A^T$$

$$P = P X$$

$$① A^T X$$

$$② A^T A$$

$$③ (A^T A)^{-1}$$

$\rightarrow G-S$

$$\begin{bmatrix} A^T A & | & A^T X \end{bmatrix}$$

$$A = QR$$

$$Q = \begin{bmatrix} q_1 \\ \vdots \\ q_n \end{bmatrix}$$

R

$R^T = R$

R^{-1} exists

$R^{-1} = R^T$

Symm, $\pm nv$

Linearity

$$T(ax + bx) = aT(x) + bT(y)$$

$$P = A(A^T A)^{-1} A^T$$

$$P = A(A^T A)^{-1} A^T X$$

\square

onto

T_{mat} has
pivs in every
col

1-to-1

T_{mat} has
pivs in
every row

Spaces V, W are
orthogonal, if
all $v^T w = 0$

orthogonal
complement

all vcs $\perp V$

(or basis
vecs)

$$\text{Null}(A) \subset \text{Col}(A^T)$$

$$\text{Null}(A^T) \subset \text{Col}(A)$$

$$x = c_1 x_1 + \dots + c_n x_n, A x = c_1 (A x_1) + \dots + c_n (A x_n)$$

If you know T
of all basis vcs
— columns of standard mat — you
know T of all vcs

Orthogonality
implies
independence

LLS

$$\hat{x} = (A^T A)^{-1} A^T B$$

$$\begin{bmatrix} \vdots & \vdots \\ \vdots & \vdots \\ \vdots & \vdots \end{bmatrix} \begin{bmatrix} m \\ b \end{bmatrix} = \begin{bmatrix} \vdots \\ \vdots \\ \vdots \end{bmatrix}$$

invert,
 U^T

$$R = \begin{bmatrix} q_1^T a_1 & q_1^T a_2 \\ 0 & q_2^T a_2 \end{bmatrix}$$

if $b \perp$

proj b onto $\text{Col}(A)$
also $b - \text{proj}$ \perp onto $\text{Null}(A^T)$

$$\hat{x} = (A^T A)^{-1} A^T B$$

$$P^T = P$$

$$P_{11} = \frac{a a^T}{a^T a}$$

onto a

onto A

$$P_{\text{cols}} = A (A^T A)^{-1} A^T$$

$$P = P x$$

use $A = \begin{bmatrix} x_1 & | & \vdots \\ | & & \vdots \\ | & & \vdots \end{bmatrix} [m]$
 $B = [y_1]$