

UNITARY OPERATORS, POSITIVE OPERATORS

THE QR DECOMPOSITION:

Ch7C, LADR

An operator $T: \mathcal{L}(V)$ is **isometric**
iff it preserves the norm.

$$\|Tx\| = \|x\| \quad x \in V //$$

Thm: $T: X \rightarrow Y$ is isometric iff it preserves
the inner product

$$\langle x, y \rangle = \langle Tx, Ty \rangle \quad \forall x, y \in V //$$

Lemma: An operator $U: X \rightarrow Y$ is an isometry
iff $U^*U = I$,

$$\begin{aligned} \text{Pf: } \langle x, x \rangle &= \langle U^*U x, x \rangle = \langle Ux, Ux \rangle \\ \parallel x \parallel^2 &= \parallel Ux \parallel^2 // \end{aligned}$$

A unitary matrix with real entries is an orthogonal matrix.

Properties of Unitary Operators

- ① $U^* = U^{-1}$
- ② if U is unitary, so is U^{-1}
- ③ A product of unitary operators is unitary.
- ④ $\det U = \pm 1$ i.e. e'values of U have $|x|=1$.
- ⑤ if U is isometric & $\{v_i\}_{i=1}^n$ are orthonormal vectors $\Rightarrow Uv_1, Uv_2, Uv_3, \dots, Uv_n$ is an orthonormal set of vectors.
If U is unitary then Uv_i $i=1, \dots, n$ form an orthonormal basis.

The QR Factorization $A = QR$
 $A \in \mathbb{F}^{m \times n}$

Return to G.S. Recall that

$$Qx = b = x_1 q_1 + x_2 q_2 + \dots + x_n q_n$$

here $\langle q_i, q_j \rangle = \delta_{ij}$ and $Q = \begin{bmatrix} | & | & & | \\ q_1 & q_2 & \dots & q_n \\ | & | & & | \end{bmatrix} \in \mathbb{R}^{m \times n}$

so $x_i = (q_i^T b) q_i$ are the coordinates //

Let's find that P in LSQ is $P = QQ^T$: to show this, let

$Qx = b$ (this may be true, but not generally true).

Project b onto cols of Q :

$$Pb = Q\tilde{x} \quad (\text{projective of } b \text{ onto } \text{col}(Q))$$

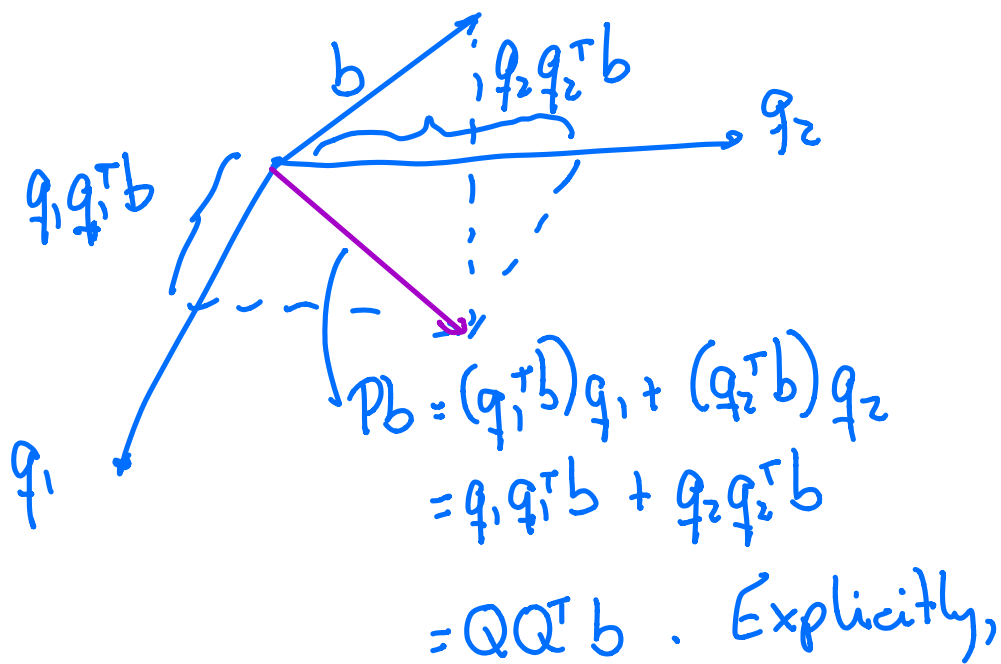
$$Q^T Q \tilde{x} = Q^T b \quad \therefore$$

$\underbrace{Pb}_{Pb}$

$$Pb = QQ^T b \Rightarrow \boxed{\tilde{P} = QQ^T}$$

So, geometrically, in $\mathbb{R}^3$:

let $b \in \mathbb{R}^3$ and q_1, q_2 standard basis of x - y plane:



ex) $q_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ $q_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ $b = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$

$Pb = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix}$ $q_1^T b q_1 = \begin{bmatrix} x \\ 0 \\ 0 \end{bmatrix}$, $q_2^T b q_2 = \begin{bmatrix} 0 \\ y \\ 0 \end{bmatrix}$

used G.S. to find q_1, q_2, q_3 :

$q_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$, $q_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$. By inspection, $q_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$.

$P = q_1 q_1^T + q_2 q_2^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \Rightarrow Pb = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix}$

i.e. $P = QQ^T$ projects b onto xy plane. //

QR Factorization:

Since $A = \begin{bmatrix} | & | & | \end{bmatrix} \in \mathbb{R}^{3 \times 3}$, so it is easy to follow:

Here a, b, c are LI

Use G.S to find q_1, q_2, q_3 :

$$a = (q_1^T a) q_1$$

$$b = (q_1^T b) q_1 + (q_2^T b) q_2$$

$$c = (q_1^T c) q_1 + (q_2^T c) q_2 + (q_3^T c) q_3$$

$$A = \begin{bmatrix} | & | & | \end{bmatrix} = \begin{bmatrix} | & | & | \\ q_1 & q_2 & q_3 \end{bmatrix} \begin{bmatrix} q_1^T a & q_1^T b & q_1^T c \\ 0 & q_2^T b & q_2^T c \\ 0 & 0 & q_3^T c \end{bmatrix}$$

$$A = Q$$

R
(Right of Diagonal)

Rule: $\text{diag}(R)$ are lengths a', b', c'

off diag (r_{ij}) lengths of q_1, q_2, q_3

Ex) $A = \begin{bmatrix} 1 & 1 & 2 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 0 & 0 & 1 \\ 1/\sqrt{2} & -1/\sqrt{2} & 0 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 1/\sqrt{2} & \sqrt{2} \\ 0 & 1/\sqrt{2} & \sqrt{2} \\ 0 & 0 & 1 \end{bmatrix}$

Confirm this yourself numerically $\begin{bmatrix} a & b & c \\ \vdots & \vdots & \vdots \end{bmatrix} = \begin{bmatrix} q_1 & q_2 & q_3 \end{bmatrix} R = QR$

Every matrix $A \in \mathbb{R}^{m \times n}$ with $L1$ columns can be factored $A = QR$. Q is orthogonal & R is upper triangular and invertible.

When $m=n$ Q becomes an orthogonal matrix.

Application of QR to LSQ:

$$Ax = b \text{ (may not have a solution)}$$

settle for LSQ solution $\tilde{x}$. Given by

$$\tilde{x} = (A^T A)^{-1} A^T b, \text{ assume } A = QR$$

Note: $A^T A = (QR)^T (QR) = R^T Q^T Q R = R^T R$

if $A^T A \tilde{x} = A^T b$

$$R^T R \tilde{x} = R^T Q^T b \Rightarrow \boxed{R \tilde{x} = Q^T b}$$

$$\begin{aligned}
 \text{Look at } Pb &= QR [(QR)^T QR]^{-1} (QR)^T b \\
 &= QR (R^T R)^{-1} (QR)^T b \\
 &= \underbrace{QR R^{-1}}_I \underbrace{(R^T)^{-1} R^T}_I Q^T b = QQ^T b
 \end{aligned}$$

$$\therefore \boxed{Pb = QQ^T b}$$

$\Rightarrow$ if $A = QR \Rightarrow$ LSG can be computed very efficiently, i.e. $O(N)$ where N is the "size" of matrix

POSITIVE OPERATORS

An operator $T \in \mathcal{L}(V)$ is called **positive** if T is S.A. and

$$\langle Tv, v \rangle \geq 0 \quad \forall v \in V //$$

Positive definite if $\langle Tv, v \rangle > 0$

Positive indefinite if $\langle Tv, v \rangle \geq 0$

The following are equivalent:

- (a) T is positive
- (b) T is SA & all e'ches of T are non-negative.
- (c) T has a positive "Square root"
i.e. $R^2 = T$
- (d) T has a S.D. square root
- (e) $\exists R \in \mathcal{L}(V)$ s.t. $T = R^*R //$

Singular Values of T : Since $T \in \mathcal{L}(V)$,

The singular values of T are eigenvalues of $\sqrt{T^*T}$. i.e. the singular values are the non-negative square roots of the e'ches of T^*T .