

THE INNER PRODUCT: for $x, y \in \mathbb{C}^m$

$$x^* y = \sum_{i=1}^m \bar{x}_i y_i$$

ex) Euclidean length: $x \in \mathbb{C}^m$

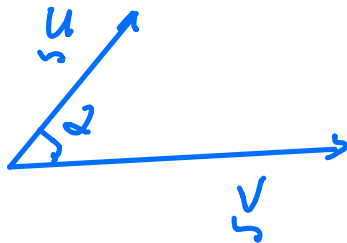
$$\|x\| = \sqrt{x^* x} = \sqrt{\sum_{j=1}^m |x_j|^2}$$

Note that: $x_j = \alpha_j + i\beta_j$ then

$$|x_j|^2 = (\alpha_j + i\beta_j)(\alpha_j - i\beta_j) = \alpha_j^2 + \beta_j^2$$

The Dot Product

$$\underline{v} \cdot \underline{u} = \underline{u} \cdot \underline{v} = |\underline{u}| |\underline{v}| \cos \alpha$$



if $\underline{u}, \underline{v}$ are real then $\underline{u} \cdot \underline{v} = \underline{u}^T \underline{v}$

Some Important Vector/matrix Properties:

Vectors:

$$(\alpha x)^* (\beta y) = \bar{\alpha} \beta x^* y$$

$$\alpha, \beta \in \mathbb{C}$$

$$(x y)^* = y^* x^*$$

Matrices, provided A & B are compatible

$$(AB)^* = B^* A^*$$

$$(AB)^{-1} = B^{-1} A^{-1}$$

$$A^{-*} \equiv (A^{-1})^* = (A^*)^{-1}$$

ORTHOGONAL VECTORS

2 vectors are **orthogonal** $\perp$ if $\underline{x}^* \underline{y} = 0$

if $\underline{x} \neq \underline{0}$ $\underline{y} \neq \underline{0}$

$$\underline{x}^* \underline{y} = |\underline{x}| |\underline{y}| \cos \alpha \Rightarrow \alpha = 90^\circ$$

2 sets X & Y are said to be $\perp$ if
all elements of X are $\perp$ to all elements of Y .

ORTHONORMAL VECTORS: $\perp$ have norm = 1

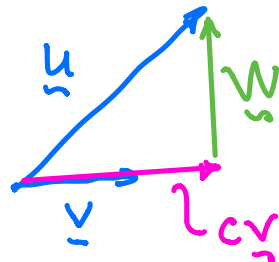
Thm 2.1 Orthogonal vectors are $\perp$ (linearly independent)

Thm 2.2 If an orthogonal set $S \subseteq \mathbb{C}^m$
contains m vectors $\Rightarrow$ the set S spans $\mathbb{C}^m$

$$S = \{s_1, s_2, \dots, s_m\}$$

vector space $\mathbb{C}^m = \{s_1\} \oplus \{s_2\} \oplus \dots \oplus \{s_m\}$
is a DIRECT SUM OF SUBSPACES $\{s_i\}_{i=1}^m$

Projection: consider a simple example: the projection
of $\underline{u}$ onto the vector $\underline{v}$:



$$\underline{u} = c\underline{v} + \underline{w}, \quad c = \frac{\underline{u}^* \underline{v}}{\|\underline{v}\|^2}$$

$\underline{u}$ has been written in terms
of an orthogonal basis

where $\underline{w} = \underline{u} - \frac{\underline{u}^* \underline{v}}{\|\underline{v}\|^2} \underline{v}$.

3D Case Suppose we know

3 **L**I vectors $\underline{a}, \underline{b}, \underline{c}$

Want to produce 3 vectors

$\underline{q}_1, \underline{q}_2, \underline{q}_3 \perp$ and normalized (orthonormal)

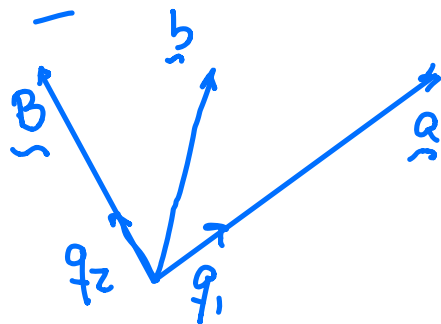
① Pick $\underline{q}_1 \parallel \underline{a}$ so $\underline{q}_1 = \frac{\underline{a}}{\|\underline{a}\|}$

② Find $\underline{q}_2$: if $\underline{b}$ has components in the $\underline{q}_1$ direction, these have to be subtracted

$$\underline{B} = \underline{b} - (\underline{q}_1^* \underline{b}) \underline{q}_1$$

$$\underline{q}_2 = \frac{\underline{B}}{\|\underline{B}\|} \quad \text{note that}$$

$$\underline{q}_2 \perp \underline{q}_1 \quad \& \quad \|\underline{q}_1\| = \|\underline{q}_2\| = 1.$$



So $\underline{q}_3$ cannot be in the plane

spanned by $\underline{q}_1$ & $\underline{q}_2$. So $\underline{c}$ should not have any components in $\text{Span}\{\underline{q}_1, \underline{q}_2\}$

$$\underline{c} = \underline{c} - (\underline{q}_1^* \underline{c}) \underline{q}_1 - (\underline{q}_2^* \underline{c}) \underline{q}_2$$

$$\underline{q}_3 = \frac{\underline{c}}{\|\underline{c}\|} //$$

Generally: Suppose $\underline{v} \in \mathbb{C}^m$ $m \geq n$:

Spse $\{\underline{q}_1, \underline{q}_2, \dots, \underline{q}_n\}$ are $\perp$, normal. Let

$\underline{v}$ be an arbitrary vector. Then

$\underline{q}_j(\underline{q}_j^* \underline{v})$ is a scalar projection of $\underline{v}$ onto $\underline{q}_j$. So

$$\underline{v} - (\underline{q}_1^* \underline{v}) \underline{q}_1 - (\underline{q}_2^* \underline{v}) \dots (\underline{q}_n^* \underline{v}) \underline{q}_n \equiv \underline{r}$$

$\underline{r} \perp \{\underline{q}_1, \underline{q}_2, \dots, \underline{q}_n\}$, since

$$\underline{q}_i^* \underline{r} = \underline{q}_i^* \underline{v} - \underline{q}_i^* (\underline{q}_1^* \underline{v}) \underline{q}_1 - \dots - \underline{q}_i^* (\underline{q}_n^* \underline{v}) \underline{q}_n$$

$$q_i^* r = q_i^* v - (q_i^* v) (q_i^* q_i) = 0$$

Hence the generalization of ~~(*)~~ is

$$v = r + \sum_{i=1}^n (q_i^* v) q_i$$

UNITARY MATRICES

Spse matrix $Q \in \mathbb{C}^{n \times n}$ is **Unitary**
(if Q is real we say Q is **orthonormal**)

$$\text{Then } Q^* = Q^{-1} \quad Q^* Q = I$$

$$q_i^* q_j = \delta_{ij}$$

$$\delta_{ij} = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{otherwise} \end{cases}$$

The cols of Q form an orthonormal basis for $\mathbb{C}^n$.

Multiplication by Unitary Matrix

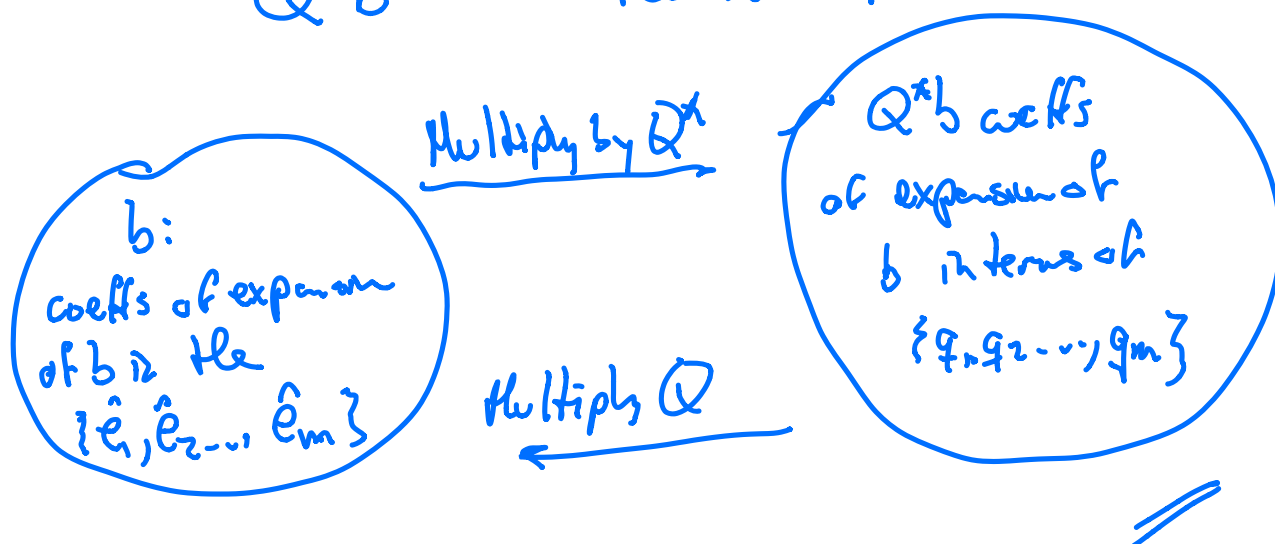
Recall that $b = Ax$ is interpreted as
 x , the set of coeff's that uniquely represent
 b in terms of cols of A

Take $A=Q$. Then

$$b = Qx$$

is the representation of b as a direct sum of subspaces of $\mathbb{C}^m$:

Q^*b finds the x coefficients.



Property of Q : preserves the inner product:

$$(Qx)^*(Qy) = x^* Q^* Q y = x^* y$$

(invariance of angles)

Also

$$\|Qx\| = \|x\|$$

In real case, multiplication by orthogonal matrix Q corresponds to rigid rotation ($\det Q = 1$) or a reflection ($\det Q = -1$) of vector space. //

Lecture 3 Vector Norms

L_p norm of a function $f(x)$

$$\left[\int_{\Omega} |f(x)|^p dx \right]^{1/p} \quad p=1, 2, \dots, \infty$$

Ω is the domain of $f(x)$

Note: $\|f\|_{\infty} = \max_{x \in \Omega} |f(x)|$

The "discrete" version is the l_p (p-norm)

$$l_p \quad \|x\|_p = \left(\sum_{i=1}^m |x_i|^p \right)^{1/p} \quad p=1, 2, \dots, \infty$$

for $x \in \mathbb{C}^m$

It gets confusing to have an L_p norm for the space of L_p functions
So Trefethen calls the norms p -norms.

Vector Norm : is the non-negative function $\|\cdot\|$

$\|\cdot\| : \mathbb{C}^m \rightarrow \{\mathbb{R}^+, 0\}$. It assigns a length to each vector.

We use it to describe the size of vector quantities and/or distance between vector constructs.

Norm Properties

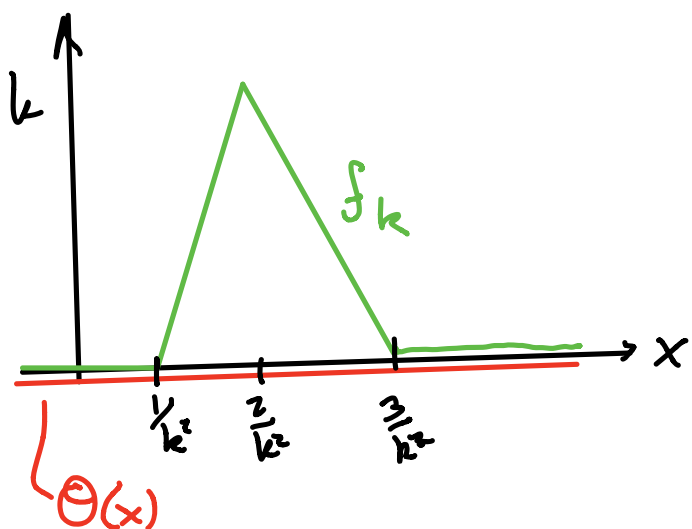
(1) $\|x\| \geq 0$ and $\|x\| = 0$ if $x = \emptyset$
(positivity)

(2) $\|x+y\| \leq \|x\| + \|y\|$ triangle inequality

(3) $\|\alpha x\| = |\alpha| \|x\|$ $\alpha \in \mathbb{C}$ scaling //

ex) let $\Theta(x) = 0$

$$\text{let } f_k = \begin{cases} k(k^2x-1) & \frac{1}{k^2} \leq x < \frac{2}{k^2} \\ -k(k^2x-1) & \frac{2}{k^2} \leq x < \frac{3}{k^2} \\ 0 & \text{otherwise} \end{cases}$$



$$\text{if } \|\theta - f_k\|_1 = \frac{1}{k}$$

$$\text{if } \|\theta - f_k\|_2 = \frac{\sqrt{2}}{\sqrt{3}}$$

$$\text{if } \|\theta - f_k\|_\infty = k$$

The L_1 norm decreases as $k \rightarrow \infty$

L_2 " constant

L_∞ " increase as $k \rightarrow \infty$

Matrices & Induced Norms

$$A \in \mathbb{C}^{m \times n} \quad x \in \mathbb{C}^n$$

$$Ax \in \mathbb{C}^m$$

Want a notion of length or size for A :

$$\|Ax\|_{(m)} \leq C \|x\|_{(n)}$$

$$\|A\|_{(m,n)} = \sup_{\substack{x \in \mathbb{C}^n \\ x \neq 0}} \frac{\|Ax\|_{(m)}}{\|x\|_{(n)}}$$

$$= \sup_{\substack{x \in \mathbb{C}^n \\ \|x\|_{(n)}=1}} \|Ax\|_{(m)}$$

$$x \in \mathbb{C}^n$$

$$\|x\|_{(n)}=1$$

