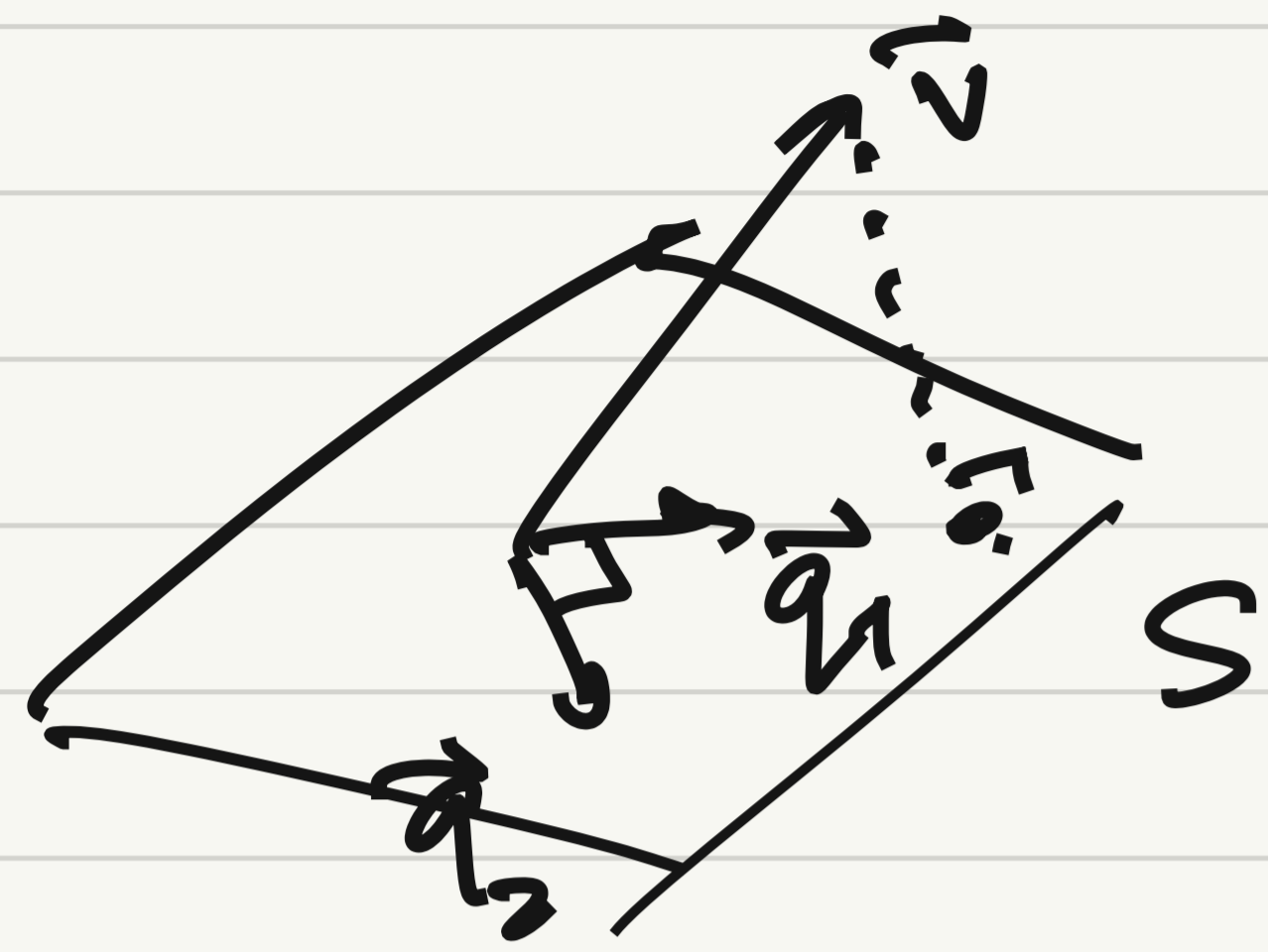


COL 726: QR factorization



$$\langle \vec{q}_1, \dots, \vec{q}_n \rangle$$

$$P \vec{v} = \underbrace{Q Q^*}_{P} \vec{v}$$

$$Q = \begin{bmatrix} \vec{q}_1 & \dots & \vec{q}_n \end{bmatrix}$$

$$S = \text{range}(A) = \langle \vec{a}_1, \dots, \vec{a}_n \rangle \xrightarrow{?} \langle \vec{q}_1, \dots, \vec{q}_n \rangle$$

$A \in \mathbb{R}^{m \times n}$
full rank

Gram-Schmidt

P_S : ortho.
proj.
onto S

$$\vec{a}_1 \rightarrow \vec{a}_1 / \|\vec{a}_1\|_2 = \vec{q}_1$$

$$\vec{a}_2 \rightarrow P_{\langle \vec{q}_1 \rangle} \vec{a}_2 = \vec{a}_2 - \underbrace{\vec{q}_1 \vec{q}_1^*}_{\text{proj. onto } \vec{q}_1} \vec{a}_2 = \vec{v}_2 \rightarrow \frac{\vec{v}_2}{\|\vec{v}_2\|_2} = \vec{q}_2$$

$$\vec{a}_3 \rightarrow P_{\langle \vec{q}_1, \vec{q}_2 \rangle} \vec{a}_3 = \vec{a}_3 - \vec{q}_1 \vec{q}_1^* \vec{a}_3 - \vec{q}_2 \vec{q}_2^* \vec{a}_3 = \vec{v}_3 \rightarrow \dots$$

$$\vec{a}_j \rightarrow \vec{v}_j = \vec{a}_j - \sum_{i < j} \vec{q}_i \vec{q}_i^* \vec{a}_j \rightarrow \vec{q}_j = \vec{v}_j / \|\vec{v}_j\|_2$$

for each $j = 1, \dots, n$

$$\vec{v}_j = \vec{a}_j$$

for each $i = 1, \dots, j-1$

$$\vec{v}_j := \vec{q}_i \vec{q}_i^* \vec{a}_j$$

$$\vec{q}_j = \vec{v}_j / \|\vec{v}_j\|$$

classical
Gram-Schmidt
orthogonalization

$$\vec{v}_3 = P_{\langle \vec{q}_2 \rangle} + P_{\langle \vec{q}_1 \rangle} \vec{a}_3$$

Modified Gram-Schmidt

$$\vec{v}_j = P_{\langle \vec{q}_1, \dots, \vec{q}_{j-1} \rangle} \vec{a}_j = P_{\langle \vec{q}_{j-1} \rangle} \cdot P_{\langle \vec{q}_2 \rangle} \cdot P_{\langle \vec{q}_1 \rangle} \vec{a}_j$$

for each $i = 1, \dots, j-1$

$$\vec{v}_j := \vec{q}_i \vec{q}_i^* \vec{v}_j$$

mathematically equivalent,
numerically more stable!

$$a_1 \rightarrow \vec{v}_1 = a_1 \rightarrow \vec{q}_1 = \vec{v}_1 / \|\vec{v}_1\|$$

$$a_2 \rightarrow P_{\langle \vec{q}_1 \rangle^\perp} a_2 = \vec{v}_2 \rightarrow \vec{q}_2 = \vec{v}_2 / \|\vec{v}_2\|$$

$$a_3 \rightarrow P_{\langle \vec{q}_1 \rangle^\perp} a_3 \rightarrow P_{\langle \vec{q}_2 \rangle^\perp} P_{\langle \vec{q}_1 \rangle^\perp} a_3 = \vec{v}_3 \rightarrow \vec{q}_3$$

⋮

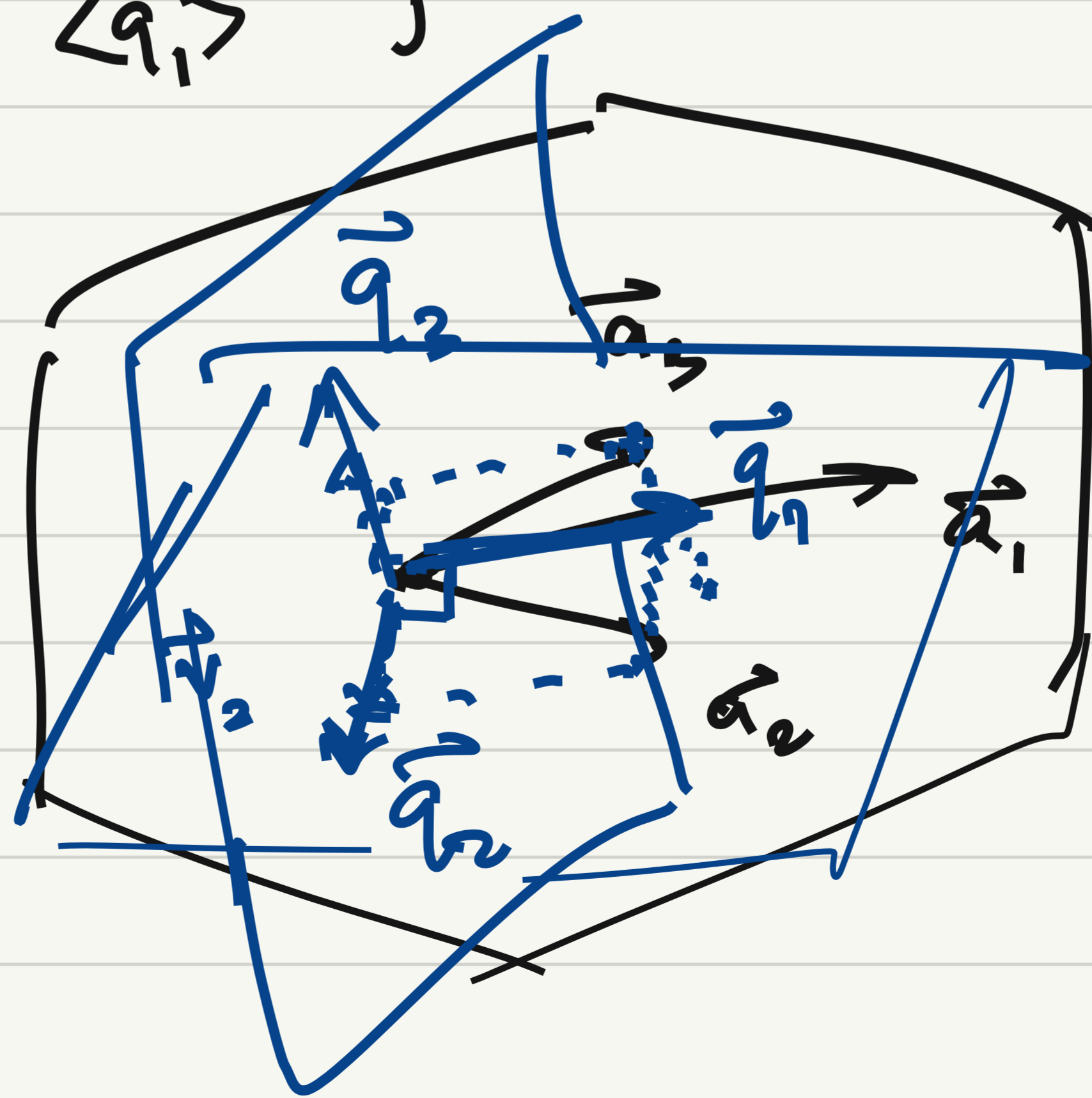
$$a_j \rightarrow P_{\langle \vec{q}_1 \rangle^\perp} a_j \rightarrow P_{\langle \vec{q}_2 \rangle^\perp} P_{\langle \vec{q}_1 \rangle^\perp} a_j \rightarrow \dots \rightarrow P_{\langle \vec{q}_{j-1} \rangle^\perp} \dots P_{\langle \vec{q}_1 \rangle^\perp} a_j = \vec{v}_j$$

for $i = 1, \dots, n$

$$\vec{q}_i = \vec{v}_i / \|\vec{v}_i\|$$

for $j = i+1, \dots, n$

$$\vec{v}_j = \vec{q}_i \vec{q}_i^* \vec{v}_j$$



↓
 \vec{v}_j

$\underbrace{\vec{a}_1, \dots, \vec{a}_n}_{\text{columns of } A} \rightarrow \underbrace{\vec{q}_1, \dots, \vec{q}_n}_{\text{orthonormal}}$

$$A = \begin{bmatrix} \vec{a}_1 & \dots & \vec{a}_n \end{bmatrix} \quad \hat{Q} = \begin{bmatrix} \vec{q}_1 & \dots & \vec{q}_n \end{bmatrix}$$

$A \in \mathbb{C}^{m \times n}$, $\vec{b} \in \mathbb{C}^m$, $\vec{b} \stackrel{?}{\in} \text{range}(A)$, what is \vec{x} s.t. $A\vec{x} \approx \vec{b}$?

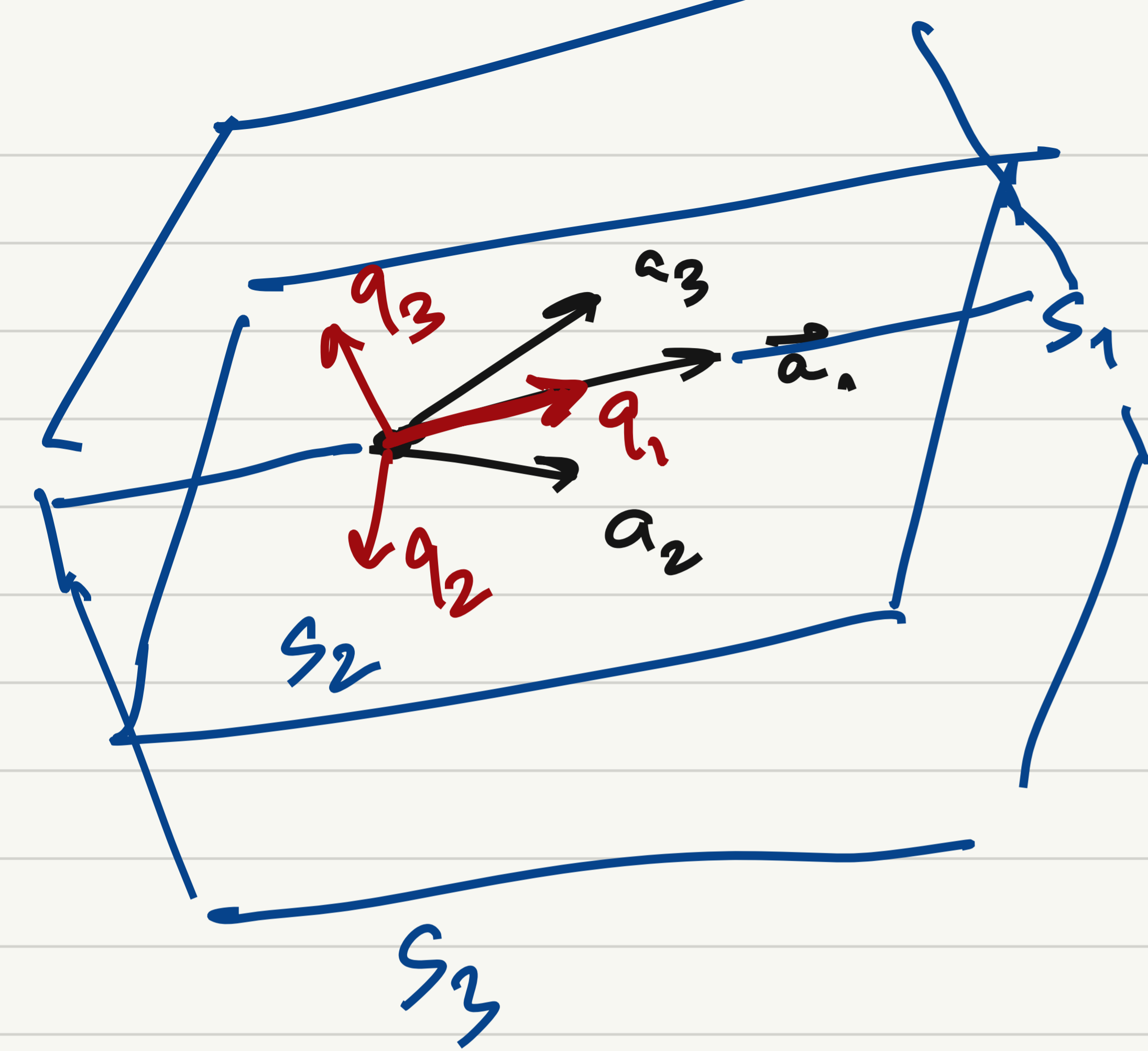
$$\Rightarrow \vec{b} = \alpha_1 \vec{a}_1 + \alpha_2 \vec{a}_2 + \dots$$

$$\langle \vec{a}_1, \dots, \vec{a}_n \rangle = \langle \vec{q}_1, \dots, \vec{q}_n \rangle$$

$$\vec{a}_j \in \langle \vec{q}_1, \dots, \vec{q}_n \rangle \Rightarrow \vec{a}_j = r_{j1} \vec{q}_1 + \dots + r_{jn} \vec{q}_n = \hat{Q} \vec{r}_j$$

$$\vec{r}_j = \hat{Q}^* \vec{a}_j$$

$$\vec{r}_j = \begin{bmatrix} r_{j1} \\ \vdots \\ r_{jn} \end{bmatrix}$$



$$\langle \vec{a}_1 \rangle \subseteq \langle \vec{a}_1, \vec{a}_2 \rangle \subseteq \langle \vec{a}_1, \dots, \vec{a}_3 \rangle \subseteq \dots$$

$$\begin{aligned} &= \\ &= \\ &= \end{aligned} \langle \vec{q}_1 \rangle \subseteq \langle \vec{q}_1, \vec{q}_2 \rangle \subseteq \langle \vec{q}_1, \dots, \vec{q}_3 \rangle \subseteq$$

$$\langle \vec{q}_1, \dots, \vec{q}_j \rangle = \langle \vec{a}_1, \dots, \vec{a}_j \rangle$$

$$\vec{a}_j = r_{1j} \vec{q}_1 + \dots + r_{jj} \vec{q}_j = \hat{Q} \begin{bmatrix} r_{1j} \\ \vdots \\ r_{jj} \\ \vdots \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$A = \begin{bmatrix} \vec{a}_1 & \dots & \vec{a}_j & \dots & \vec{a}_n \end{bmatrix}$$

$$= \hat{Q} \begin{bmatrix} r_{11} & \dots & r_{1j} & \dots & r_{1n} \\ \vdots & & \vdots & & \vdots \\ & & r_{jj} & & \vdots \\ & & & \ddots & \\ & & & & r_{nn} \end{bmatrix}$$

\hat{R}

reduced QR factorization

Reduced QR factorization: $A \in \mathbb{C}^{m \times n}$

$$A = \hat{Q} \hat{R}, \quad \hat{Q} \in \mathbb{C}^{m \times n}, \text{ orthonormal columns}$$

$$\hat{R} \in \mathbb{C}^{n \times n}, \text{ upper triangular}$$

$$\begin{bmatrix} A \end{bmatrix} = \begin{bmatrix} \hat{Q} \end{bmatrix} \begin{bmatrix} \hat{R} \end{bmatrix}$$

full QR factorization: $A = QR$, $Q \in \mathbb{C}^{m \times m}$ unitary

$R \in \mathbb{C}^{m \times n}$ upper tri.

$$\begin{bmatrix} A \end{bmatrix} = \underbrace{\begin{bmatrix} \hat{Q} \end{bmatrix}}_Q \underbrace{\begin{bmatrix} \hat{R} \\ 0 \end{bmatrix}}_R$$

$$Q = \left[\begin{array}{c|c} \vec{q}_1 & \dots & \vec{q}_n & \vec{q}_{n+1} & \dots & \vec{q}_m \end{array} \right]$$

$\vec{q}_{n+1}, \dots, \vec{q}_m \perp \text{range}(A)$.

If A is full rank,
 $\{\vec{q}_1, \dots, \vec{q}_n\}$ are ^{orthonormal} basis for $\text{range}(A)$.
 $\{\vec{q}_{n+1}, \dots, \vec{q}_m\}$ are orthonormal basis
 for $\text{range}(A)^\perp$

$$A\vec{x} = \vec{b}$$

$$A \in \mathbb{C}^{m \times m}$$

$$A = QR$$

$$\Rightarrow QR\vec{x} = \vec{b}$$

$$\Rightarrow R\vec{x} = Q^*\vec{b}$$

$$\begin{bmatrix} \times & \times & \times \\ \times & \times & \times \\ \times & \times & \times \end{bmatrix} \begin{bmatrix} \vec{x} \\ \vdots \\ 0 \end{bmatrix} = \begin{bmatrix} Q^*\vec{b} \\ \vdots \\ 0 \end{bmatrix}$$

What are r_{ij} ?

$$\vec{v}_j = \vec{a}_j - \vec{q}_1 \vec{q}_1^* \vec{a}_j - \dots - \vec{q}_{j-1} \vec{q}_{j-1}^* \vec{a}_j, \quad \vec{q}_j = \frac{\vec{v}_j}{\|\vec{v}_j\|}$$

back substitution
 $\mathcal{O}(n^2)$ time

$$\vec{a}_j = r_{1j} \vec{q}_1 + \dots + r_{j-1,j} \vec{q}_{j-1} + r_{jj} \vec{q}_j$$

$$\Rightarrow \vec{q}_j = \frac{\vec{a}_j - r_{1j} \vec{q}_1 - \dots - r_{j-1,j} \vec{q}_{j-1}}{r_{jj}}$$

$$\Rightarrow r_{jj} = \|\vec{v}_j\|$$

$$\left. \begin{aligned} r_{1j} &= \vec{q}_1^* \vec{a}_j \\ &\vdots \\ r_{ij} &= \vec{q}_i^* \vec{a}_j \end{aligned} \right\}$$

Operation count : flops (floating-point operations) : add, sub, mul, div, sqrt

Both CGS and MGS require $\sim 2mn^2$ flops

Def: $f(x) \sim g(x)$ if $\lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = 1$

$$2mn^2 + 10mn + n^2 + 500 \sim 2mn^2$$

$$3mn^2 \not\sim 2mn^2$$

MGS:

for $i = 1 \dots n$

$$r_{ii} = \|\vec{v}_i\| \leftarrow \sim 2m$$

$$\vec{q}_i = \vec{v}_i / r_{ii}$$

for $j = i+1 \dots n$

$$r_{ij} = \vec{q}_i \cdot \vec{v}_j$$

$$\vec{v}_j = r_{ij} \vec{q}_i$$

} $\sim m$ mult,
 $\sim m$ add
 $\sim m$ mult
 $\sim m$ sub
 $\sim 4m$ flops

$$\sum_{i=1}^n \sum_{j=i+1}^n 4m \sim 2mn^2$$

Hilbert matrix =

$$\begin{bmatrix} 1 & 1/2 & 1/3 & 1/4 & \dots \\ 1/2 & 1/3 & 1/4 & & \\ 1/3 & 1/4 & & & \\ 1/4 & & & & \\ \vdots & & & & \end{bmatrix}$$

If you have to do Gram-Schmidt: don't use CGS

If you just want QR: don't use GS!

$$A = \begin{bmatrix} \vec{a}_1 & \dots & \vec{a}_n \end{bmatrix}$$

$$V = \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix}$$

initially $V = A$

$$V^{(0)} \rightarrow V^{(1)} \rightarrow V^{(2)} \rightarrow \dots \rightarrow \hat{Q}$$

$$V = \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix}$$

then $\vec{v}_1 \rightarrow \vec{v}_1 / \|\vec{v}_1\| = \vec{v}_1 / r_{11}$

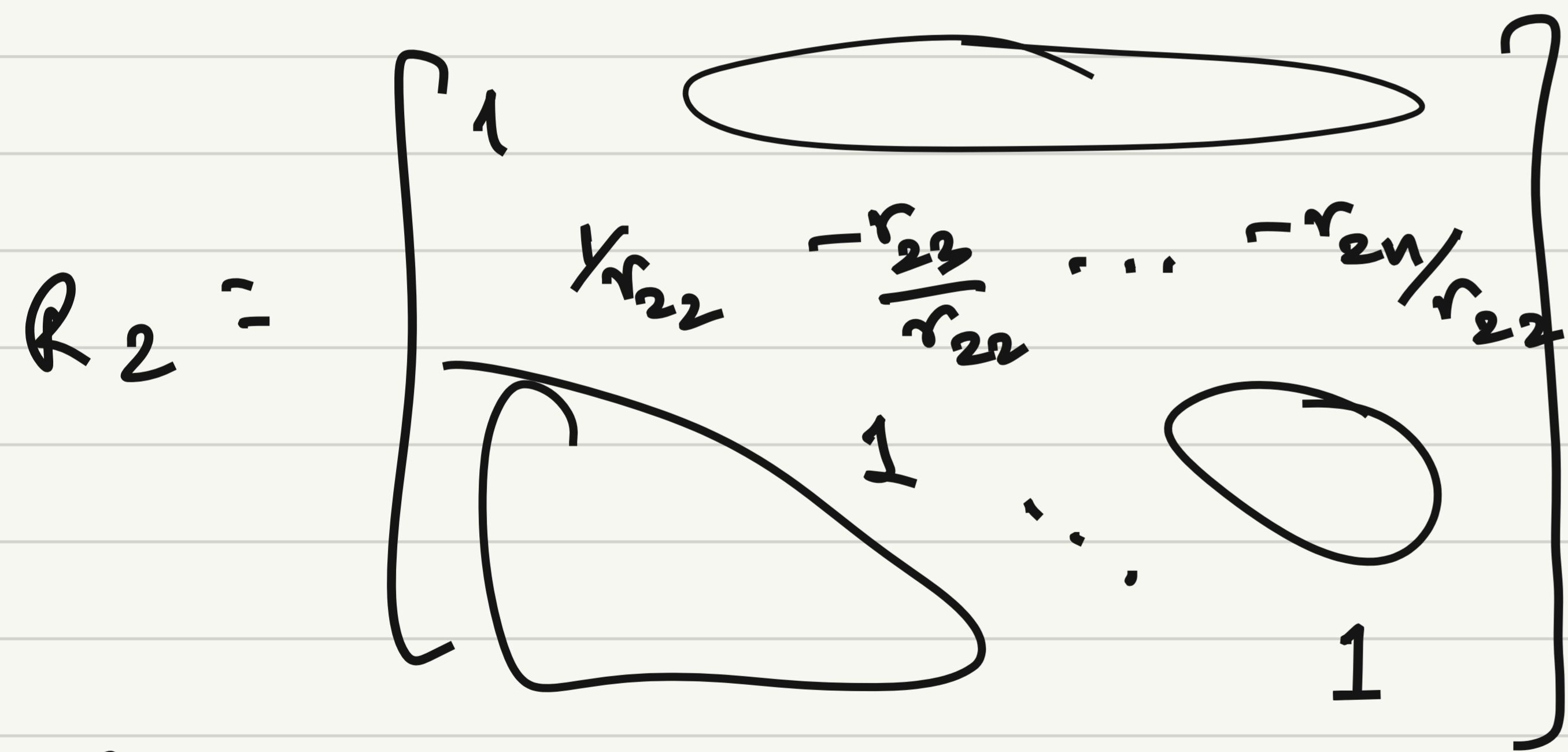
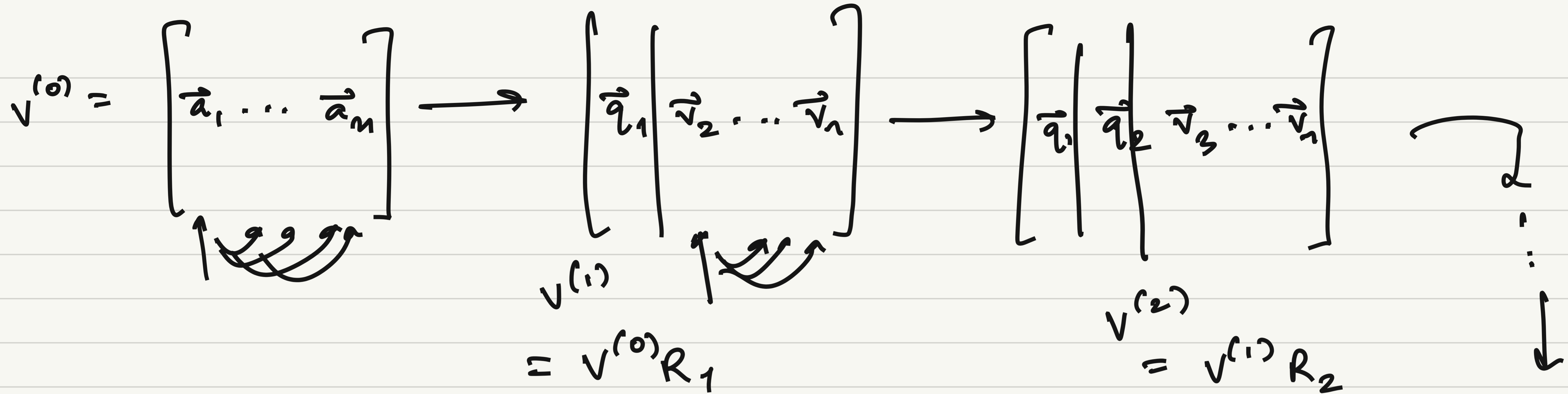
$$\vec{v}_2 \rightarrow \vec{v}_2 - \frac{\vec{v}_1^* \vec{v}_2}{\|\vec{v}_1\|} \vec{v}_1 = \vec{v}_2 - \frac{r_{12}}{r_{11}} \vec{v}_1$$

$$\begin{matrix} \vec{v}^{(0)} \\ \parallel \end{matrix} \left[\begin{matrix} \vec{z}_1 \\ \dots \\ \vec{z}_n \end{matrix} \right] \rightarrow \left[\begin{matrix} \vec{z}_1 \\ \vec{z}_2 \\ \vec{z}_3 \\ \dots \\ \vec{z}_n \end{matrix} \right] = \vec{v}^{(0)} R_1$$

$$\begin{matrix} \parallel \\ \vec{v}^{(0)} \end{matrix} \left[\begin{matrix} \vec{z}_1 \\ \dots \\ \vec{z}_n \end{matrix} \right] = \left[\begin{matrix} 0 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{matrix} \right] = \vec{v}^{(0)} \left[\begin{matrix} \vec{q}_1 \\ \vec{q}_2 \\ \dots \\ \vec{q}_n \end{matrix} \right]$$

R_1

orthogonal to \vec{q}_1



(Exercise)

$V^{(0)} = A$
 $V^{(1)} = AR_1$

$V^{(2)} = AR_1 R_2$

\vdots

$\hat{Q} = V^{(n)} = AR_1 R_2 \dots R_n$

upper triangular

$V^{(n)} = \begin{bmatrix} \vec{q}_1 & \vec{q}_2 & \dots & \vec{q}_n \end{bmatrix}$

$$\hat{Q} = A(R_1 R_2 \dots R_n)$$

GS = "triangular orthogonalization"

$$A = \hat{Q} \underbrace{(R_1 R_2 \dots R_n)^{-1}}_{\hat{R}}$$

At each step, coeffs are $\frac{1}{r_{jj}}$, $-\frac{r_{ij}}{r_{jj}}$, ...

Dividing by small number

produced by subtraction is dangerous.

Next class: Householder algorithm for QR

"orthogonal triangularization"