

5.3 Least-squares problems

Complex vector space \mathbb{C}^n :

Real

$$\langle \vec{x}, \vec{y} \rangle = \sum_{i=1}^n x_i y_i$$

$$\|\vec{x}\|^2 = \sum_{i=1}^n x_i^2$$

Symmetric: $A^T = A$

$$\langle \vec{x}, A\vec{y} \rangle = \langle A\vec{x}, \vec{y} \rangle$$

orthogonal: $A^T A = A A^T = I$

Complex

$$\langle \vec{x}, \vec{y} \rangle = \sum_{i=1}^n \bar{x}_i \bar{y}_i$$

$$\|\vec{x}\|^2 = \sum_{i=1}^n |x_i|^2$$

Hermitian: $A^* = A$

$$(A^*)_{ij} = \overline{A_{ji}}$$

Unitary: $A^* A = A A^* = I$

Let A be an $m \times n$ (possibly complex) matrix, $\vec{b} \in \mathbb{C}^m$. We want to find $\vec{x} \in \mathbb{C}^n$ s.t. $\|A\vec{x} - \vec{b}\|$ is minimized.

$$\begin{array}{c|c} m & \begin{array}{|c|} \hline n \\ \hline \end{array} \\ \hline A & \vec{x} \end{array} = \begin{array}{|c|} \hline \vec{b} \\ \hline \end{array}$$

Lemma \vec{x} solves the least squares problem $\Leftrightarrow \underbrace{A^* A \vec{x} = A^* \vec{b}}_{\text{the normal equation}}$

Proof ① If $A^* A \vec{x} = A^* \vec{b}$ then

$$\begin{aligned} \|A(\vec{x} + \vec{g}) - \vec{b}\|^2 &= \|(A\vec{x} - \vec{b}) + A\vec{g}\|^2 & \langle \vec{x} - \vec{g}, \vec{g} \rangle &= \overline{\langle \vec{g}, \vec{x} \rangle} \\ &= \langle (A\vec{x} - \vec{b}) + A\vec{g}, (A\vec{x} - \vec{b}) + A\vec{g} \rangle & \langle \vec{x} - \vec{g}, A\vec{g} \rangle &= \langle A^* \vec{x}, \vec{g} \rangle \\ &= \|A\vec{x} - \vec{b}\|^2 + \|A\vec{g}\|^2 + 2\operatorname{Re} \langle A\vec{x} - \vec{b}, A\vec{g} \rangle \\ &= \|A\vec{x} - \vec{b}\|^2 + \|A\vec{g}\|^2 + 2\operatorname{Re} \langle \underline{A^*(A\vec{x} - \vec{b})}, \vec{g} \rangle \\ &\quad = \underline{0} \end{aligned}$$

$$= \|A\hat{x} - \vec{b}\|^2 + \|A\vec{y}\|^2 \geq \|A\hat{x} - \vec{b}\|^2$$

② If \hat{x} solves least squares, suppose $\tilde{z} := A^*(A\hat{x} - \vec{b}) \neq \vec{0}$

$$\|A(\hat{x} + \vec{y}) - \vec{b}\|^2 = \|A\hat{x} - \vec{b}\|^2 + \|A\vec{y}\|^2 + 2\operatorname{Re}\langle \tilde{z}, \vec{y} \rangle$$

Take $\vec{y} = -\varepsilon \tilde{z}$ w/ $\varepsilon > 0$ small. Then

$$\|A(\hat{x} + \vec{y}) - \vec{b}\|^2 = \|A\hat{x} - \vec{b}\|^2 + \varepsilon^2 \|A\tilde{z}\|^2 - 2\varepsilon \underline{\|\tilde{z}\|^2} > 0$$

\Rightarrow when ε is sufficiently small, we get

$$\|A(\hat{x} + \vec{y}) - \vec{b}\|^2 < \|A\hat{x} - \vec{b}\|^2 \quad \text{contradiction.}$$

QR factorization

Def A QR-factorization of an $m \times n$ matrix A is

$$A = QR \quad \Leftrightarrow Q^* A = R$$

where Q is $m \times m$ unitary and R is $m \times n$ upper-triangular.

$$\begin{matrix} n \\ m \end{matrix} \begin{array}{|c|} \hline \square \\ \hline \end{array} = \begin{matrix} m \\ m \end{matrix} \begin{array}{|c|c|c|} \hline \square & | & \dots & | \\ \hline \end{array} \begin{matrix} n \\ m \end{matrix} \begin{array}{|c|} \hline \triangle \\ \hline \end{array}$$

A Q R

If A has QR-factorization, then least squares problem can be solved

$$A^* A \hat{x} = A^* \vec{b}$$

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

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

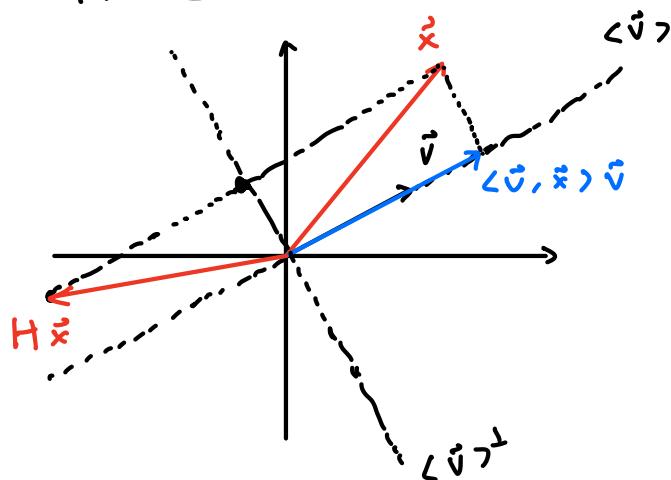
\hookrightarrow upper-triangular
 \hookrightarrow lower-triangular.

We obtain QR-factorization by building Q^* as a product of Householder transformations, i.e., a matrix of the form

$$\begin{pmatrix} I_k & 0 \\ 0 & I_{m-k} - 2\vec{v}\vec{v}^* \end{pmatrix} \quad w' \quad \vec{v} \in \mathbb{C}^{m-k}, \quad \| \vec{v} \| = 1$$

- Geometrically, a Householder transf. is a reflection about $\langle \vec{v} \rangle^\perp$

$$H = I - 2\vec{v}\vec{v}^T \text{ in } \mathbb{R}^2$$



$$\begin{aligned} H\vec{x} &= \vec{x} - 2\vec{v}\vec{v}^T\vec{x} \\ &= \vec{x} - 2\langle \vec{v}, \vec{x} \rangle \vec{v} \end{aligned}$$

Starting from $A = A^{(1)}$

Step 1 : find $\vec{v}^{(1)}$ s.t.

$$\vec{e}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} I_m - 2\vec{v}^{(1)}\vec{v}^{(1)*} \\ U^{(1)} \end{pmatrix} \left(\begin{array}{c|c|c} & & \\ \hline & & \dots \\ \hline & & \end{array} \right) \begin{array}{c|c} m \\ \hline A^{(1)} \end{array} = \begin{pmatrix} * & \\ \hline \begin{matrix} 0 \\ \vdots \\ 0 \end{matrix} & \dots \end{pmatrix} \begin{array}{c|c} \\ \hline A^{(1)} \end{array}$$

Denote \vec{x} as the first column of $A^{(1)}$. We need

$$(I_m - 2\vec{v}\vec{v}^*)\vec{x} = \alpha \vec{e}_1 \quad ; \quad |\alpha| = \|\vec{x}\|$$

$$\vec{x} - 2\langle \vec{x}, \vec{v} \rangle \vec{v} = \alpha \vec{e}_1 \quad ; \quad \alpha = -e^{i\phi} \|\vec{x}\|$$

$$\vec{x} - \alpha \vec{e}_1 = 2\langle \vec{x}, \vec{v} \rangle \vec{v}$$

\hookrightarrow choose $\phi = \arg x_1$
to avoid subtracting close
numbers in $x_1 - \alpha$.

$$\Rightarrow \vec{v} = \frac{\vec{x} - \alpha \vec{e}_1}{\|\vec{x} - \alpha \vec{e}_1\|} \quad (\text{since } \|\vec{v}\| = 1)$$

⋮

Step k :

$$\begin{pmatrix} I_{k-1} \\ I_{m-k+1} - 2\vec{v}^{(k)} \vec{v}^{(k)*} \end{pmatrix} \begin{pmatrix} \square & \square \\ \square & A^{(k)} \\ \vdots & \end{pmatrix} = \begin{pmatrix} \square & \square \\ \square & A^{(k+1)} \\ \vdots & \end{pmatrix}$$

$\cdots \hat{x} \cdots$

After $n-1$ steps

$$\underbrace{U^{(n-1)} \cdots U^{(1)} U^{(1)*}}_{Q^*} A = A^{(n)} := R$$

$$Q = U^{(1)} U^{(2)} \cdots U^{(n-1)}$$

Total cost of QR-factorization for full matrix: $O(mn^2)$ (for $m \geq n$)