Numerical Analysis Hilary Term 2018 Lecture 6: QR Factorization

Definition: a square real matrix Q is **orthogonal** if $Q^{T} = Q^{-1}$. This is true if, and only if, $Q^{T}Q = I = QQ^{T}$.

Example: the permutation matrices P in LU factorization with partial pivoting are orthogonal.

Proposition. The product of orthogonal matrices is an orthogonal matrix.

Proof. If S and T are orthogonal, $(ST)^{\mathrm{T}} = T^{\mathrm{T}}S^{\mathrm{T}}$ so

$$(ST)^{\mathrm{T}}(ST) = T^{\mathrm{T}}S^{\mathrm{T}}ST = T^{\mathrm{T}}(S^{\mathrm{T}}S)T = T^{\mathrm{T}}T = I.$$

Definition: The scalar (dot)(inner) product of two vectors

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \text{ and } y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

in \mathbb{R}^n is

$$x^{\mathrm{T}}y = y^{\mathrm{T}}x = \sum_{i=1}^{n} x_{i}y_{i} \in \mathbb{R}$$

Definition: Two vectors $x, y \in \mathbb{R}^n$ are **orthogonal** if $x^T y = 0$. A set of vectors $\{u_1, u_2, \ldots, u_r\}$ is an **orthogonal set** if $u_i^T u_j = 0$ for all $i, j \in \{1, 2, \ldots, r\}$ such that $i \neq j$.

Lemma. The columns of an orthogonal matrix Q form an orthogonal set, which is moreover an orthonormal basis for \mathbb{R}^n .

Proof. Suppose that $Q = [q_1 \ q_2 \ \vdots \ q_n]$, i.e., q_j is the *j*th column of Q. Then

$$Q^{\mathrm{T}}Q = I = \begin{bmatrix} q_1^{\mathrm{T}} \\ q_2^{\mathrm{T}} \\ \vdots \\ q_n^{\mathrm{T}} \end{bmatrix} [q_1 \ q_2 \ \vdots \ q_n] = \begin{bmatrix} 1 \ 0 \ \cdots \ 0 \\ 0 \ 1 \ \cdots \ 0 \\ \vdots \ \vdots \ \ddots \ \vdots \\ 0 \ 0 \ \cdots \ 1 \end{bmatrix}.$$

Comparing the (i, j)th entries yields

$$q_i^{\mathrm{T}} q_j = \begin{cases} 0 & i \neq j \\ 1 & i = j. \end{cases}$$

Note that the columns of an orthogonal matrix are of length 1 as $q_i^{\mathrm{T}}q_i = 1$, so they form an orthonormal set \iff they are linearly independent (check!) \implies they form an

orthonormal basis for \mathbb{R}^n as there are n of them.

Lemma. If $u \in \mathbb{R}^n$, P is n-by-n orthogonal and v = Pu, then $u^{\mathrm{T}}u = v^{\mathrm{T}}v$. **Proof.** See problem sheet.

Definition: The **outer product** of two vectors x and $y \in \mathbb{R}^n$ is

$$xy^{\mathrm{T}} = \begin{bmatrix} x_1y_1 & x_1y_2 & \cdots & x_1y_n \\ x_2y_1 & x_2y_2 & \cdots & x_2y_n \\ \vdots & \vdots & \ddots & \vdots \\ x_ny_1 & x_ny_2 & \cdots & x_ny_n \end{bmatrix},$$

an *n*-by-*n* matrix (notation: $xy^{\mathrm{T}} \in \mathbb{R}^{n \times n}$). More usefully, if $z \in \mathbb{R}^n$, then

$$(xy^{\mathrm{T}})z = xy^{\mathrm{T}}z = x(y^{\mathrm{T}}z) = \left(\sum_{i=1}^{n} y_i z_i\right)x.$$

Definition: For $w \in \mathbb{R}^n$, $w \neq 0$, the **Householder** matrix $H(w) \in \mathbb{R}^{n \times n}$ is the matrix

$$H(w) = I - \frac{2}{w^{\mathrm{T}}w}ww^{\mathrm{T}}.$$

Proposition. H(w) is an orthogonal matrix. **Proof.**

$$H(w)H(w)^{\mathrm{T}} = \left(I - \frac{2}{w^{\mathrm{T}}w}ww^{\mathrm{T}}\right)\left(I - \frac{2}{w^{\mathrm{T}}w}ww^{\mathrm{T}}\right)$$
$$= I - \frac{4}{w^{\mathrm{T}}w}ww^{\mathrm{T}} + \frac{4}{(w^{\mathrm{T}}w)^{2}}w(w^{\mathrm{T}}w)w^{\mathrm{T}}$$
$$= I.$$

Lemma. Given $u \in \mathbb{R}^n$, there exists a $w \in \mathbb{R}^n$ such that

$$H(w)u = \begin{bmatrix} \alpha \\ 0 \\ \vdots \\ 0 \end{bmatrix} \equiv v,$$

say, where $\alpha = \pm \sqrt{u^{\mathrm{T}} u}$.

Remark: Since H(w) is an orthogonal matrix for any $w \in \mathbb{R}$, $w \neq 0$, it is necessary for the validity of the equality H(w)u = v that $v^{\mathrm{T}}v = u^{\mathrm{T}}u$, i.e., $\alpha^2 = u^{\mathrm{T}}u$; hence our choice of $\alpha = \pm \sqrt{u^{\mathrm{T}}u}$.

Proof. Take $w = \gamma(u - v)$, where $\gamma \neq 0$. Recall that $u^{\mathrm{T}}u = v^{\mathrm{T}}v$. Thus,

$$w^{\mathrm{T}}w = \gamma^{2}(u-v)^{\mathrm{T}}(u-v) = \gamma^{2}(u^{\mathrm{T}}u - 2u^{\mathrm{T}}v + v^{\mathrm{T}}v) = \gamma^{2}(u^{\mathrm{T}}u - 2u^{\mathrm{T}}v + u^{\mathrm{T}}u) = 2\gamma u^{\mathrm{T}}(\gamma(u-v)) = 2\gamma w^{\mathrm{T}}u.$$

 So

$$H(w)u = \left(I - \frac{2}{w^{\mathrm{T}}w}ww^{\mathrm{T}}\right)u = u - \frac{2w^{\mathrm{T}}u}{w^{\mathrm{T}}w}w = u - \frac{1}{\gamma}w = u - (u - v) = v.$$

Now if u is the first column of the *n*-by-*n* matrix A,

$$H(w)A = \begin{bmatrix} \alpha & \times & \cdots & \times \\ \hline 0 & & & \\ \vdots & B & \\ 0 & & & \end{bmatrix}, \text{ where } \times = \text{general entry.}$$

Similarly for B, we can find $\hat{w} \in \mathbb{R}^{n-1}$ such that

$$H(\hat{w})B = \begin{bmatrix} \beta & \times & \cdots & \times \\ \hline 0 & & & \\ \vdots & C & \\ 0 & & & \end{bmatrix}$$

and then

$$\begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & & \\ \vdots & & H(\hat{w}) \\ 0 & & \end{bmatrix} H(w)A = \begin{bmatrix} \alpha & \times & \times & \cdots & \times \\ 0 & \beta & \times & \cdots & \times \\ 0 & 0 & & \\ 0 & 0 & & \\ \vdots & \vdots & & C \\ 0 & 0 & & \\ \end{bmatrix}$$

Note

$$\begin{bmatrix} 1 & 0 \\ 0 & H(\hat{w}) \end{bmatrix} = H(w_2), \text{ where } w_2 = \begin{bmatrix} 0 \\ \hat{w} \end{bmatrix}.$$

Thus if we continue in this manner for the n-1 steps, we obtain

$$\underbrace{H(w_{n-1})\cdots H(w_3)H(w_2)H(w)}_{Q^{\mathrm{T}}}A = \begin{bmatrix} \alpha & \times & \cdots & \times \\ 0 & \beta & \cdots & \times \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \gamma \end{bmatrix} = (\neg)$$

The matrix Q^{T} is orthogonal as it is the product of orthogonal (Householder) matrices, so we have constructively proved that

Theorem. Given any square matrix A, there exists an orthogonal matrix Q and an upper triangular matrix R such that

$$A = QR$$

Notes: 1. This could also be established using the Gram–Schmidt Process. 2. If u is already of the form $(\alpha, 0, \dots, 0)^{\mathrm{T}}$, we just take H = I. 3. It is not necessary that A is square: if $A \in \mathbb{R}^{m \times n}$, then we need the product of (a) m-1 Householder matrices if $m \leq n \Longrightarrow$

$$(\square) = A = QR = (\square) (\frown)$$

or (b) n Householder matrices if $m > n \Longrightarrow$

$$\left(\square \right) = A = QR = \left(\square \right) \left(\square \right).$$

Another useful family of orthogonal matrices are the **Givens rotation** matrices:

$$J(i, j, \theta) = \begin{bmatrix} 1 & & & \\ & \cdot & & \\ & c & s & \\ & & \cdot & \\ & & -s & c & \\ & & & \cdot & \\ & & & & 1 \end{bmatrix} \leftarrow i \text{th row}$$
$$\leftarrow j \text{th row}$$
$$\uparrow & \uparrow \\ i & j$$

where $c = \cos \theta$ and $s = \sin \theta$.

Exercise: Prove that $J(i, j, \theta)J(i, j, \theta)^{\mathrm{T}} = I$ — obvious though, since the columns form an orthonormal basis.

Note that if $x = (x_1, x_2, \ldots, x_n)^T$ and $y = J(i, j, \theta)x$, then

$$y_k = x_k \text{ for } k \neq i, j$$

$$y_i = cx_i + sx_j$$

$$y_i = -sx_i + cx_j$$

and so we can ensure that $y_j = 0$ by choosing $x_i \sin \theta = x_j \cos \theta$, i.e.,

$$\tan \theta = \frac{x_j}{x_i} \text{ or equivalently } s = \frac{x_j}{\sqrt{x_i^2 + x_j^2}} \text{ and } c = \frac{x_i}{\sqrt{x_i^2 + x_j^2}}.$$
 (1)

Thus, unlike the Householder matrices, which introduce lots of zeros by pre-multiplication, the Givens matrices introduce a single zero in a chosen position by pre-multiplication. Since (1) can always be satisfied, we only ever think of Givens matrices J(i, j) for a specific vector or column with the angle chosen to make a zero in the *j*th position, e.g., J(1, 2)xtacitly implies that we choose $\theta = \tan^{-1} x_2/x_1$ so that the second entry of J(1, 2)x is zero. Similarly, for a matrix $A \in \mathbb{R}^{m \times n}$, $J(i, j)A := J(i, j, \theta)A$, where $\theta = \tan^{-1} a_{ji}/a_{ii}$, i.e., it is the *i*th column of A that is used to define θ so that $(J(i, j)A)_{ji} = 0$.

We shall return to these in a later lecture.