

# Assignment 3

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## 1 Workout 2.3

**Show that orthogonal matrices preserve the 2-norm and the Frobenius norm of matrices.**

Let  $Q$  be a orthogonal matrix  $Q \in \mathbb{R}^{m \times n}$  and an arbitrary matrix  $A \in \mathbb{R}^{n \times m}$

$$\|QA\|_2^2 = (QAu)^T QA = A^T Q^T QA = A^T A = \|A\|_2^2 \quad (1)$$

We know that  $\|A\|_F^2 = \sum_{i=1}^m \sum_{j=1}^n a_{ij}^2 = \text{trace}(AA^T)$

$$\|QA\|_F^2 = \sum_{i=1}^m \sum_{j=1}^n q_{ij}^2 a_{ij}^2 = \text{trace}(QA(QA)^T) = \text{trace}(QAA^T Q^T) \quad (2)$$

The trace is invariant to cyclic permutations and so we got:

$$\text{trace}(QAA^T Q^T) = \text{trace}(Q^T Q A^T A) = \text{trace}(A^T A) = \|A\|_F^2 \quad (3)$$

## 2 Workout 2.8

**Show that it requires  $2n^2(m - \frac{n}{3})$  flops to compute  $R$  in the QR-factorization of  $A \in \mathbb{R}^{m \times n}, m \geq n$  using Householder transformations. This cost does not include the explicit construction of  $Q$ . Show that it is required  $4(m^2n - mn^2 + \frac{n^3}{3})$**

The Householder matrix multiplication in QR factorization:

$$HA = A - \frac{2}{u^T u} u u^T A = A - \beta u (u^T A) = A - \beta u w^T \quad (4)$$
$$\beta = \frac{2}{u^T u} \quad w = A^T u$$

We get approximately  $mn$  flops for each of these calculations. We then proceed to calculate the sum when we reduce the size of  $H$ .

$$\begin{aligned} \sum_{k=1}^n 4(m-k)(n-k) &= 4 \sum_{i=1}^{n-1} (m-n+i)i = 4(m-n) \sum_{i=0}^{n-1} i + 4 \sum_{i=0}^{n-1} i^2 = \\ &= 2(m-n)n(n-1) + 4 \sum_{i=0}^{n-1} i^2 \approx 2(m-n)n(n-1) + 4 \int_0^n x^2 dx = \\ &= 2(m-n)n(n-1) + \frac{4}{3}n^3 = 2mn^2 - \frac{2}{3}n^3 \end{aligned} \quad (5)$$

### 3 Workout 3.3

Let  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n \geq 0$  be the singular values of  $A \in \mathbb{R}^{m \times n}$  with  $m \geq n$  and prove:

**3.1**  $\|A\|_2 = \sigma_1$

$L_2$  norm is defined as:  $\max \frac{\|Ax\|_2}{\|x\|_2}$

$$\|A\|_2^2 = \max \frac{\|Ax\|_2^2}{\|x\|_2^2} = \max \frac{x^T A^T A x}{x^T x} = \lambda_1(A^T A) = \lambda_1((U \Sigma V^T)^T U \Sigma V^T) \quad (6)$$

$$\lambda_1(\Sigma^T \Sigma) = \sigma_1^2$$

The max is the largest vector in A

**3.2**  $\|A\|_F = \sqrt{\sigma_1^2 + \dots + \sigma_n^2}$

$$\|A\|_F^2 = \text{trace}(A^T A) = \text{trace}((U \Sigma V^T)^T U \Sigma V^T) = \text{trace}(V \Sigma^T U^T U \Sigma V^T) \quad (7)$$

$$\text{trace}(V \Sigma^T \Sigma V^T) = \text{trace}(\Sigma^T \Sigma) = \sigma_1^2 + \sigma_2^2 + \dots + \sigma_n^2$$

**3.3**  $\|A^{-1}\|_2 = \frac{1}{\sigma_n}$

Using that  $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n \geq 0$

$$\|A^{-1}\|_2^2 = \max \frac{\|x\|_2^2}{\|Ax\|_2^2} = \max \frac{x^T x}{Ax^T A^T Ax} = \frac{1}{\sigma_n^2} \quad (8)$$

**3.4**  $\text{cond}(A)_2 = \frac{\sigma_1}{\sigma_n}$

$$\text{Cond}_2(A) = \|A\| \|A^{-1}\| = [\text{from 1 and 3 we get}] = \frac{\sigma_1}{\sigma_n} \quad (9)$$

### 3.5 Rank(A) = number of nonzeros singular values

If we decompose A such that  $A = U \Sigma V^T$  and U and V are orthogonal matrices. The multiplication of these invertible/orthogonal matrices with  $\Sigma$  doesn't change the rank of  $\Sigma$ . The rank of  $\Sigma$  is therefore the rank of A. The rank of  $\Sigma$  is equal the number of non-singular values.

## 4 Workout 3.5

Let  $A$  be an  $m \times n$  matrix of full rank  $r = \min(m, n)$ . If  $C$  is another  $m \times n$  matrix such that  $\|C - A\|_2 < \sigma_r$ , then show  $C$  also has full rank.

Using SVD let  $D$  be the nearest singular matrix to  $A$ :

$$A = \begin{bmatrix} \sigma_1 & 0 & \dots & \dots & 0 \\ 0 & \sigma_2 & \dots & \dots & 0 \\ \vdots & \dots & \ddots & \vdots & \vdots \\ 0 & \dots & \dots & \sigma_{r-1} & 0 \\ 0 & \dots & \dots & \dots & \sigma_r \\ 0 & \dots & \dots & \dots & 0 \end{bmatrix}, D = \begin{bmatrix} \sigma_1 & 0 & \dots & \dots & 0 \\ 0 & \sigma_2 & \dots & \dots & 0 \\ \vdots & \dots & \ddots & \vdots & \vdots \\ 0 & \dots & \dots & \sigma_{r-1} & 0 \\ 0 & \dots & \dots & \dots & 0 \\ 0 & \dots & \dots & \dots & 0 \end{bmatrix} \quad (10)$$

The L2 norm  $\|A - D\|_2 = \|\Sigma_A - \Sigma_D\|_2 = \sigma_r$  if  $\|A - C\|_2 < \sigma_r$  then the rank of  $C$  must be of the same  $A$ .

## 5 WO 2.9 programming Python

*#Implement a function for QR factorization*

```
def grot(a, b):
    if abs(a) > abs(b):
        t = b/a; c = 1/np.sqrt(1+t**2); s = c*t
    else:
        t = a/b; s = 1/np.sqrt(1+t**2); c = s*t

    return np.array([c, s])

def G(R, c, s, i):
    R[i][i] = s
    R[i][i - 1] = c
    R[i - 1][i] = -c
    R[i - 1][i - 1] = s
    return R

def qrfac(A, mode = "R"):
    m, n = np.shape(A)
    if mode == "R":
        R = A
        for j in range(1, n + 1, 1):
            for i in range(m, j, -1):
                Q = np.eye(m)
                [a, b] = grot(R[i - 1][j - 1],
                               R[i - 2][j - 1])
                R = np.transpose(G(Q, a, b, i - 1)) @ R
        return np.triu(R)
```

```

elif mode == "RQ":
    Qsave = np.eye(m)
    R = A
    for j in range(1, n + 1, 1):
        for i in range(m, j, -1):
            Q = np.eye(m)
            [a, b] = grot(R[i - 1][j - 1],
                          R[i - 2][j - 1])
            R = np.transpose(G(Q, a, b, i - 1)) @ R
            Qsave = Qsave @ Q
    return np.triu(R), Qsave[0:m][0:m]

else:
    print('Input_mode_types_ "R" _or_ "QR" _')

A = np.array([[0.8147, 0.0975, 0.1576],
              [0.9058, 0.2785, 0.9706],
              [0.1270, 0.5469, 0.9572],
              [0.9134, 0.9575, 0.4854],
              [0.6324, 0.9649, 0.8003]])

B = np.array([[2, 3, 5],
              [1, 2, -1],
              [2, 5, 3],
              [1, -1, 0]])

C = np.array([[6, 5, 0],
              [5, 1, 4],
              [0, 4, 3]])

r, q = qrfac(A, "RQ")
print(q@r-A)

r, q = qrfac(B, "RQ")
print(q@r-B)

r, q = qrfac(C, "RQ")
print(q@r-C)

#OUTPUT

[[ 0.00000000e+00  0.00000000e+00  2.77555756e-17]
 [-2.22044605e-16  1.11022302e-16 -3.33066907e-16]
 [-2.77555756e-17 -2.22044605e-16 -5.55111512e-16]
 [-4.44089210e-16 -4.44089210e-16  0.00000000e+00]
 [-3.33066907e-16 -4.44089210e-16 -3.33066907e-16]]

[[-2.22044605e-16 -4.44089210e-16  0.00000000e+00]]

```

$$\begin{bmatrix}
0.00000000e+00 & 0.00000000e+00 & 3.33066907e-16 \\
0.00000000e+00 & -8.88178420e-16 & 0.00000000e+00 \\
0.00000000e+00 & 1.55431223e-15 & 2.50130237e-16
\end{bmatrix}$$

$$\begin{bmatrix}
-1.77635684e-15 & -8.88178420e-16 & 2.84775512e-17 \\
-8.88178420e-16 & 2.22044605e-16 & -8.88178420e-16 \\
0.00000000e+00 & 0.00000000e+00 & 0.00000000e+00
\end{bmatrix}$$