

ETH Lecture 401-0663-00L Numerical Methods for CSE

# Mid-term Exam

Autumn Term 2020

Oct 31, 2020, 10:00, HG E1.2 / HG E7 / HG F1

**Don't  
panic!**

Family Name		%
First Name		
Department		
Legi Nr.		
Date	Oct 31, 2020	

Points:

	1	2	3	Total
max	10	9	8	
achvd				

- This is a **closed-book exam**.
- Keep only writing material and your ETH ID card on the table.
- Keep mobile phones, tablets, smartwatches, etc. turned off in your bag.
- Fill in this cover sheet first.
- Turn the cover sheet only when instructed to do so.
- Then write your name and ETH ID number on every page.
- **Write your answers in the appropriate fields on these problem sheets.**
- **Wrong ticks in multiple-choice boxes will lead to points being subtracted.** Hence, mere guessing is really dangerous! If you have no clue, leave all tickboxes empty.
- If you change your mind about an answer to an MC-question, write a clear NO next to the old answer, draw fresh tickboxes and fill them.
- **Anything written outside the answer boxes will not be taken into account.**
- Do not write with red/green color or with pencil.
- Make sure to hand in every sheet.
- Two blank pages handed out with the exam: space for notes
- **Duration: 30 minutes.**

Throughout the exam use the notations introduced in class, in particular [Lecture → Section 1.1.1]:

- $(\mathbf{A})_{i,j}$  to refer to the entry of the matrix  $\mathbf{A} \in \mathbb{K}^{m,n}$  at position  $(i,j)$ .

- $(\mathbf{A})_{:,i}$  to designate the  $i$ -th-column of the matrix  $\mathbf{A}$ ,
- $(\mathbf{A})_{i,:}$  to denote the  $i$ -th row of the matrix  $\mathbf{A}$ ,
- $(\mathbf{A})_{i:j,k:\ell}$  to single out the sub-matrix  $\left[(\mathbf{A})_{r,s}\right]_{\substack{i \leq r \leq j \\ k \leq s \leq \ell}}$  of the matrix  $\mathbf{A}$ ,
- $(\mathbf{x})_k$  to reference the  $k$ -th entry of the vector  $\mathbf{x}$ ,
- $\mathbf{e}_j \in \mathbb{R}^n$  to write the  $j$ -th Cartesian coordinate vector,
- $\mathbf{I}$  to denote the identity matrix,
- $\mathbf{O}$  to write a zero matrix,
- $\mathcal{P}_n$  for the space of (univariate polynomials of degree  $\leq n$ ),
- and superscript indices in brackets to denote iterates:  $\mathbf{x}^{(k)}$ , etc.

By default, vectors are regarded as column vectors.

**Problem 0-1: Economical QR-decomposition**

The economical QR-decomposition is a fundamental matrix factorization in numerical linear algebra. This problem reviews some of its properties.

This problem is based on the contents of [Lecture → Section 3.3.3].


Given a matrix  $\mathbf{A} \in \mathbb{R}^{m,n}$ ,  $m > n$ ,  $\mathbf{A} \neq \mathbf{O}$ , we write  $\mathbf{A} = \mathbf{Q}\mathbf{R}$  for its **economical (thin) QR decomposition**.

(0-1.a)  (2 pts.) Fill in the correct matrix sizes

$\mathbf{Q} \in \mathbb{R}$   ,  ,  $\mathbf{R} \in \mathbb{R}$   , .

SOLUTION of (0-1.a):

$$\mathbf{Q} \in \mathbb{R}^{m,n} , \mathbf{R} \in \mathbb{R}^{n,n} .$$

(0-1.b)  (4 pts.) Decide, whether the following statements are true for any  $\mathbf{A} \in \mathbb{R}^{m,n} \setminus \{\mathbf{O}\}$ ,  $m > n$ .

1. The matrix  $\mathbf{Q}$  is an orthogonal matrix.

☐ YES

☐ NO

2. The columns of  $\mathbf{Q}$  provide an orthonormal basis of  $\mathcal{R}(\mathbf{A})$ .

☐ YES

☐ NO

3.  $\text{rank}(\mathbf{R}) = n$ .

☐ YES

☐ NO

4.  $\text{rank}(\mathbf{Q}) = n$ .

☐ YES


☐ NO

SOLUTION of (0-1.b):

1. NO: because  $\mathbf{Q}$  is not even square in general.

2. NO: Consider a rank-1 matrix  $\mathbf{A}$  with  $n \geq 2$ . Then  $\mathbf{Q}$  has more columns than  $\dim \mathcal{R}(\mathbf{A})$ . As a consequence, the columns of  $\mathbf{Q}$  cannot be a basis of  $\mathcal{R}(\mathbf{A})$ .
3. NO: because  $\text{rank}(\mathbf{R}) = \text{rank}(\mathbf{A})$ , but  $\text{rank}(\mathbf{A}) < n$  is not excluded.
4. YES: because the  $n$  orthonormal columns of  $\mathbf{Q} \in \mathbb{R}^{m,n}$  are linearly independent.



(0-1.c)  (2 pts.)

How many left-multiplications by Householder matrices does it take to compute  $\mathbf{R}$ ?

☐  $mn$ 
☐  $n$ 
☐  $\frac{1}{2}n(n-1)$ 
☐  $m-1$ 

SOLUTION of (0-1.c):

We need  $n$  left-multiplications with Householder matrices in order to annihilate the bottom entries of the  $n$  columns of  $\mathbf{A}$ .

Note that  $m > n$  was assumed. In the case  $m \leq n$  the computation of  $\mathbf{R}$  requires only  $m-1$  left-multiplications with Householder matrices.



(0-1.d)  (2 pts.)

What is the asymptotic complexity of EIGEN's `HouseholderQR()` methods applied to dense  $m \times n$  matrices,  $m > n$ , for  $m, n \rightarrow \infty$ ?

☐  $O(mn)$ 
☐  $O(m^2n)$ 
☐  $O(m(m+n^2))$ 
☐  $O(mn^2)$ 

SOLUTION of (0-1.d):

$O(mn^2)$  according to [Lecture  $\rightarrow$  § 3.3.3.37].



**End Problem 0-1 , 10 pts.**

**Problem 0-2: Evaluation of a quadratic form in COO format**

This problem examines the algorithmic realization of the evaluation of a quadratic form described by a sparse matrix in triplet/COO storage format.

Familiarity with C++/EIGEN and [Lecture → Section 2.7.1] is required.

In an EIGEN-based numerical code a sparse matrix in COO format is represented by the data type

```
using COOMat = std::vector<Eigen::Triplet<double>>;
```

This function converts a matrix in triplet/COO format into a dense matrix format:

```
Eigen::MatrixXd cooToDense(const COOMat &A) {
    Eigen::Index m = 0;
    Eigen::Index n = 0;
    for (const Eigen::Triplet<double> &t : A) {
        m = (m > t.row()) ? m : t.row() + 1;
        n = (n > t.col()) ? n : t.col() + 1;
    }
    Eigen::MatrixXd M = Eigen::MatrixXd::Zero(m, n);
    for (const Eigen::Triplet<double> &t : A) {
        M(t.row(), t.col()) += t.value();
    }
    return M;
}
```

(0-2.a) (9 pts.) The following function is supposed to compute  $(\mathbf{1} - \mathbf{x})^\top \mathbf{A} \mathbf{x}$ ,  $\mathbf{1} \in \mathbb{R}^n$  a vector with components all = 1, for a large sparse matrix  $\mathbf{A} \in \mathbb{R}^{n,n}$  given in triplet/COO format and a vector  $\mathbf{x} \in \mathbb{R}^n$ :

```
double evalQuadFormCOO(const COOMat &A, const Eigen::VectorXd &x) {
    double s = 0.0;
    for (const Eigen::Triplet<double> &t : A) {
        assert((t.row() < x.size()) && (t.col() < x.size()));
        s += (1.0 - x[t.row()]) * x[t.col()] * t.value();
    }
    return s;
}
```

Supplement the missing parts of C++ code in the boxes.

SOLUTION of (0-2.a):

**C++ code 0.2.1: Implementation of evalQuadFormCOO()**

```
2 double evalQuadFormCOO(const COOMat &A, const Eigen::VectorXd &x) {
3     double s = 0.0;
4     for (const Eigen::Triplet<double> &t : A) {
5         assert((t.row() < x.size()) && (t.col() < x.size()));
6         s += (1.0 - x[t.row()]) * x[t.col()] * t.value();
7     }
8     return s;
}
```

Name:

ETH ID No.:

9 }



**End Problem 0-2 , 9 pts.**

**Problem 0-3: Analysis of an EIGEN-based C++ code**

In this problem you are asked to analyze a complex code and determine the computational cost of certain operations and the size of matrices stored in certain variables.

This problem assumes familiarity with EIGEN.

In a project you are expected to understand the following undocumented C++ function, which takes two arguments, a matrix  $\mathbf{A} \in \mathbb{R}^{m,n}$ ,  $m \geq n$ , and an integer  $d < n$ .

**C++ code 0.3.1: A complex EIGEN-based C++ code**

```

2  std::pair<Eigen::VectorXd, Eigen::VectorXd> clsq2(const Eigen::MatrixXd &A,
3                                                    const unsigned d) {
4      unsigned int n = A.cols(), m = A.rows();
5      assert(m >= n);
6      const Eigen::MatrixXd R =
7          A.householderQr().matrixQR().template triangularView<Eigen::Upper>(); //
8      const auto V = R.block(d, d, m-d, n-d) //
9          .jacobiSvd(Eigen::ComputeFullV) //
10         .matrixV(); //
11      const auto y = V.col(n-d-1); //
12      const auto b = R.block(0, d, d, n-d) * y; //
13      const auto S = R.topLeftCorner(d, d); //
14      const Eigen::VectorXd D = S.diagonal().cwiseAbs(); //
15      if (D.minCoeff() < (numeric_limits<double>::epsilon()) * D.maxCoeff()) //
16          throw runtime_error("Upper left block of R not regular"); //
17      const auto z = -(S.template triangularView<Eigen::Upper>()).solve(b); //
18      return {z, y};
19  }

```

(0-3.a) (4 pts.)

Find out the type of matrix/vector stored in various variables

(i)  $\mathbf{R}$  defined in Line 7 stores matrix/vector  $\in \mathbb{R}$

 , 

(ii)  $\mathbf{V}$  defined in Line 8 stores matrix/vector  $\in \mathbb{R}$

 , 

(iii)  $\mathbf{b}$  defined in Line 12 stores matrix/vector  $\in \mathbb{R}$

 , 

(iv)  $\mathbf{z}$  defined in Line 17 stores matrix/vector  $\in \mathbb{R}$

 , 


HINT 1 for (0-3.a): The method `block(int i, int j, int p, int q)` of a matrix type in EIGEN returns a reference to the submatrix with upper left corner at position  $(i, j)$  and size  $p \times q$ .

SOLUTION of (0-3.a):

- (i)  $\mathbf{R}$  is the upper triangular part of the  $\mathbf{R}$ -factor of the QR-decomposition of the matrix  $\mathbf{A} \in \mathbb{R}^{m,n}$ . It has the same size as  $\mathbf{R}$  and stores an  $m \times n$ -matrix.

- (ii)  $\mathbf{v}$  is the  $\mathbf{V}$ -factor of the SVD of a  $(m-d) \times (n-d)$ -matrix, hence of size  $(n-d) \times (n-d)$ .
- (iii)  $\mathbf{b}$  stores a column vector of length  $d$  obtained by the multiplication of a  $d \times (n-d)$ -matrix (in  $\mathbf{v}$ ) with and  $n-d$ -vector (in  $\mathbf{y}$ ).
- (iv)  $\mathbf{S}$  contains a  $d \times d$ -matrix  $\mathbf{S} \in \mathbb{R}^{d,d}$ . Hence  $\mathbf{z}$  holds the vector  $\mathbf{S}^{-1}\mathbf{b} \in \mathbb{R}^d$ .



**(0-3.b)**  (4 pts.) Assuming that  $d$  is small and fixed, determine the asymptotic complexity in terms of  $m, n \rightarrow \infty$  of the statements in different lines of Code 0.3.1

- |       |                |                                       |
|-------|----------------|---------------------------------------|
| (i)   | Line 7         | Cost = $O(\$ <input type="text"/> $)$ |
| (ii)  | Line 8–Line 10 | Cost = $O(\$ <input type="text"/> $)$ |
| (iii) | Line 14        | Cost = $O(\$ <input type="text"/> $)$ |
| (iv)  | Line 17        | Cost = $O(\$ <input type="text"/> $)$ |

SOLUTION of (0-3.b):

- (i) Line 7: Computation of the QR-decomposition of  $\mathbf{A} \in \mathbb{R}^{m,n}$ : cost =  $O(mn^2)$ , see [Lecture  $\rightarrow$  § 3.3.3.37].
- (ii) Line 8–Line 10: Computation of the SVD of an  $(m-d) \times (n-d)$ -matrix,  $m \geq n$ , which is an economical SVD, because the full  $\mathbf{Q}$ -factor is not requested: Cost =  $O(mn^2)$ , see [Lecture  $\rightarrow$  § 3.4.2.2].
- (iii) Line 14: Extraction of a  $d$ -vector by taking the absolute values of the entries of the diagonal of a  $d \times d$ -matrix. Cost =  $O(1)$ , because  $d$  was supposed to be fixed and small.
- (iv) Line 17: Solution of a small  $d \times d$  linear system of equations. Cost =  $O(1)$ , because  $d$  was supposed to be fixed and small.



**End Problem 0-3 , 8 pts.**