ETH Zürich D-MATH

Numerical Methods for CSE

Problem Sheet 0

These problems are meant as an introduction to EIGEN in the first tutorial classes of the new semester.

Problem 1. Gram-Schmidt orthogonalization with EIGEN

- [1, Code 1.5.3] presents a MATLAB code that effects the Gram-Schmidt orthogonalization of the columns of an argument matrix.
- (1a) Based on the C++ linear algebra library EIGEN implement a function

```
template <class Matrix>
Matrix gramschmidt(const Matrix &A);
```

that performs the same computations as [1, Code 1.5.3].

Solution: See gramschmidt.cpp.

(1b) Test your implementation by applying it to a small random matrix and checking the orthonormality of the columns of the output matrix.

Solution: See gramschmidt.cpp.

Problem 2. Fast matrix multiplication

- [1, Rem. 1.4.9] presents Strassen's algorithm that can achieve the multiplication of two dense square matrices of size $n = 2^k$, $k \in \mathbb{N}$, with an asymptotic complexity better than $O(n^3)$.
- (2a) Using EIGEN implement a function

```
MatrixXd strassenMatMult(const MatrixXd & A, const
    MatrixXd & B)
```

that uses Strassen's algorithm to multiply the two matrices A and B and return the result as output.

Solution: See Listing 1.

(2b) Validate the correctness of your code by comparing the result with EIGEN's built-in matrix multiplication.

Solution: See Listing 1.

(2c) • Measure the runtime of your function strassenMatMult for random matrices of sizes 2^k , $k=4,\ldots,10$, and compare with the matrix multiplication offered by the *-operator of EIGEN.

Solution: See Listing 1.

Listing 1: EIGEN Implementation of the Strassen's algorithm and runtime comparisons.

```
#include < Eigen / Dense>
#include <iostream>
#include <vector>
   #include "timer.h"
7 using namespace Eigen;
8 using namespace std;
10 //! \brief Compute the Matrix product A \times B using
     Strassen's algorithm.
11 //! \param[in] A Matrix 2^k \times 2^k
12 //! \param[in] B Matrix 2^k \times 2^k
13 //! \param[out] Matrix product of A and B of dim 2^k \times 2^k
15 MatrixXd strassenMatMult(const MatrixXd & A, const
     MatrixXd & B)
16 {
      int n=A.rows();
     MatrixXd C(n,n);
```

```
19
         (n==2)
      i f
20
21
          C < A(0,0) *B(0,0) + A(0,1) *B(1,0)
22
               A(0,0)*B(0,1) + A(0,1)*B(1,1)
               A(1,0)*B(0,0) + A(1,1)*B(1,0)
               A(1,0)*B(0,1) + A(1,1)*B(1,1);
25
           return C:
26
      }
      else
29
           MatrixXd
         Q0(n/2, n/2), Q1(n/2, n/2), Q2(n/2, n/2), Q3(n/2, n/2),
          Q4(n/2, n/2), Q5(n/2, n/2), Q6(n/2, n/2);
32
           MatrixXd A11=A.topLeftCorner(n/2,n/2);
           MatrixXd A12=A.topRightCorner(n/2,n/2);
           MatrixXd A21=A.bottomLeftCorner(n/2,n/2);
           MatrixXd A22=A.bottomRightCorner(n/2, n/2);
36
           MatrixXd B11=B.topLeftCorner(n/2,n/2);
           MatrixXd B12=B.topRightCorner(n/2,n/2);
39
           MatrixXd B21=B.bottomLeftCorner(n/2,n/2);
           MatrixXd B22=B.bottomRightCorner(n/2, n/2);
          Q0=strassenMatMult(A11+A22, B11+B22);
43
          Q1=strassenMatMult(A21+A22, B11);
          Q2=strassenMatMult(A11,B12-B22);
          Q3=strassenMatMult(A22,B21-B11);
          Q4=strassenMatMult(A11+A12, B22);
47
          Q5=strassenMatMult(A21-A11, B11+B12);
          Q6=strassenMatMult(A12-A22, B21+B22);
          C<< Q0+Q3-Q4+Q6
51
          Q2+Q4,
          Q1+Q3,
          Q0+Q2-Q1+Q5;
           return C;
55
```

```
}
57
  int main(void)
59
  {
60
      srand((unsigned int) time(0));
61
62
      //check if strassenMatMult works
      int k=2;
      int n=pow(2,k);
      MatrixXd A=MatrixXd::Random(n,n);
66
      MatrixXd B=MatrixXd::Random(n,n);
      MatrixXd AB(n,n), AxB(n,n);
      AB=strassenMatMult(A,B);
      AxB=A*B:
70
      cout << "Using Strassen's method, A*B="<<AB<<endl;
71
      cout << "Using standard method, A*B="<<AxB<<endl;
      cout << "The norm of the error is
73
          " << (AB-AxB) . norm() << endl;
      //compare runtimes of strassenMatMult and of direct
75
         multiplication
      unsigned int repeats = 10;
      timer <> tm x, tm strassen;
      std::vector<int> times_x, times_strassen;
79
      for (unsigned int k = 4; k \le 10; k++) {
           tm x.reset();
           tm strassen.reset();
83
           for (unsigned int r = 0; r < repeats; ++r) {</pre>
               unsigned int n = pow(2,k);
               A = MatrixXd::Random(n,n);
               B = MatrixXd::Random(n,n);
87
               MatrixXd AB(n,n);
               tm x.start();
90
               AB=A*B;
91
```

```
tm_x.stop();
92
93
                tm strassen.start();
94
                AB=strassenMatMult(A,B);
95
                tm strassen.stop();
           std::cout << "The standard matrix multiplication</pre>
                            " << tm x.avg().count() /
              1000000. << " ms" << std::endl;
           std::cout << "The Strassen's algorithm took:
                     " << tm_strassen.avg().count() /</pre>
              1000000. << " ms" << std::endl;
100
           times_x.push_back( tm_x.avg().count() );
101
           times strassen.push back(
102
              tm strassen.avg().count() );
       }
103
104
       for(auto it = times x.begin(); it != times x.end();
105
          ++it) {
           std::cout << *it << " ";
106
107
       std::cout << std::endl;
108
       for (auto it = times strassen.begin(); it !=
109
          times strassen.end(); ++it) {
           std::cout << *it << " ";
110
111
       std::cout << std::endl;
112
113
114
```

Problem 3. Householder reflections

This problem is a supplement to [1, Section 1.5.1] and related to Gram-Schmidt orthogonalization, see [1, Code 1.5.3]. Before you tackle this problem, please make sure that you remember and understand the notion of a QR-decomposition of a matrix, see [1, Thm. 1.5.7]. This problem will put to the test your advanced linear algebra skills.

Listing 2: MATLAB implementation for Problem 3. in file houserefl.m

Write a C++ function with declaration:

```
void houserefl(const VectorXd &v, MatrixXd &Z);
```

that is equivalent to the MATLAB function houserefl(). Use data types from EIGEN.

Solution:

Listing 3: C++implementation for Problem 3. in file houserefl.cpp

```
g Z = X.rightCols(n-1);
10 }
```

(3b) Show that the matrix X, defined at line 10 in Listing 2, satisfies:

$$\mathbf{X}^{\mathsf{T}}\mathbf{X} = \mathbf{I}_n$$

HINT: $\|\mathbf{q}\|^2 = 1$.

Solution:

$$\mathbf{X}^{\mathsf{T}}\mathbf{X} = (\mathbf{I}_{n} - 2\mathbf{q}\mathbf{q}^{\mathsf{T}})(\mathbf{I}_{n} - 2\mathbf{q}\mathbf{q}^{\mathsf{T}})$$

$$= \mathbf{I}_{n} - 4\mathbf{q}\mathbf{q}^{\mathsf{T}} + 4\mathbf{q} \underbrace{\mathbf{q}^{\mathsf{T}}\mathbf{q}}_{=\|\mathbf{q}\|=1} \mathbf{q}^{\mathsf{T}}$$

$$= \mathbf{I}_{n} - 4\mathbf{q}\mathbf{q}^{\mathsf{T}} + 4\mathbf{q}\mathbf{q}^{\mathsf{T}}$$

$$= \mathbf{I}_{n}$$

(3c) \square Show that the first column of X, after line 9 of the function houserefl, is a multiple of the vector \mathbf{v} .

HINT: Use the previous hint, and the facts that $\mathbf{u} = \mathbf{w} + \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ and $\mathbf{w} = 1$.

Solution: Let $X = [X_1, \dots, X_n]$ be the matrix of line 9 in Listing 2, then:

$$\mathbf{X}_{1} = \mathbf{e}^{(1)} - 2q_{1}\mathbf{q}$$

$$= \begin{bmatrix} 1 - 2q_{1}^{2} \\ -2q_{1}q_{2} \\ \vdots \\ -2q_{1}q_{n} \end{bmatrix}$$

$$= \begin{bmatrix} 1 - 2\frac{u_{1}^{2}}{\sum_{i=1}^{n}u_{i}^{2}} \\ -2\frac{u_{1}u_{2}}{\sum_{i=1}^{n}u_{i}^{2}} \\ \vdots \\ -2\frac{u_{1}u_{n}}{\sum_{i=1}^{n}u_{i}^{2}} \end{bmatrix}$$

$$Hint = \begin{bmatrix} \frac{(w_{1}+1)^{2}+w_{2}^{2}+\cdots+w_{n}^{2}-2(w_{1}+1)^{2}}{(w_{1}+1)^{2}+w_{2}^{2}+\cdots+w_{n}^{2}} \\ -\frac{2(w_{1}+1)w_{2}}{(w_{1}+1)^{2}+w_{2}^{2}+\cdots+w_{n}^{2}} \\ \cdots \\ -\frac{2(w_{1}+1)w_{n}}{(w_{1}+1)^{2}+w_{2}^{2}+\cdots+w_{n}^{2}} \end{bmatrix}$$

$$\|w\| = 1 \begin{bmatrix} \frac{2w_{1}(w_{1}+1)}{2(w_{1}+1)} \\ \frac{2(w_{1}+1)w_{2}}{2(w_{1}+1)} \\ \cdots \\ \frac{2(w_{1}+1)w_{n}}{2(w_{1}+1)} \end{bmatrix}$$

$$= -\mathbf{W} = -\frac{\mathbf{V}}{\|\mathbf{V}\|},$$

which is a multiple of v.

(3d) ☑ What property does the set of columns of the matrix **Z** have? What is the purpose of the function houserefl?

HINT: Use (3b) and (3c).

Solution: The columns of $\mathbf{X} = [\mathbf{X}_1, \dots, \mathbf{X}_n]$ are an orthonormal basis (ONB) of \mathbb{R}^n (cf. (3b)). Thus, the columns of $\mathbf{Z} = [\mathbf{X}_2, \dots, \mathbf{X}_n]$ are an ONB of the complement of $\mathrm{Span}(\mathbf{X}_1) \stackrel{(3c)}{=} \mathrm{Span}(\mathbf{v})$. The function houserefl computes an ONB of the complement of \mathbf{v} .

(3e) \odot What is the asymptotic complexity of the function houserefl as the length n of the input vector \mathbf{v} goes to ∞ ?

Solution: $O(n^2)$: this is the asymptotic complexity of the construction of the tensor prod-

uct at line 9 of Listing 3.

(3f) Rewrite the function as MATLAB function and use a *standard function* of MATLAB to achieve the same result of lines 5-9 with a single call to this function.

HINT: It is worth reading [1, Rem. 1.5.10] before mulling over this problem.

Solution: Check the code in Listing 2 for the porting to MATLAB code. Using the QR-decomposition qr one can rewrite (cf. Listing 4) the C++ code in MATLAB with a few lines.

Listing 4: MATLAB implementation for Problem 3. in file qr_houserefl.m using QR decomposition.

```
function Z = qr_houserefl(v)

with two distributions are provided in the second complement of span(v)

| X,R] = qr(v);
|
```

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Hand-in: — (in the boxes in front of HG G 53/54).

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