#### AS 2015

ETH Zürich D-MATH

# Numerical Methods for CSE

### Problem Sheet 2

You should try to your best to do the core problems. If time permits, please try to do the rest as well.

## **Problem 1.** Lyapunov Equation (core problem)

Any linear system of equations with a finite number of unknowns can be written in the "canonical form" Ax = b with a system matrix A and a right hand side vector b. However, the LSE may be given in a different form and it may not be obvious how to extract the system matrix. This task gives an intriguing example and also presents an important *matrix* equation, the so-called Lyapunov Equation.

Given  $\mathbf{A} \in \mathbb{R}^{n \times n}$ , consider the equation

$$\mathbf{AX} + \mathbf{XA}^T = \mathbf{I} \tag{5}$$

with unknown  $\mathbf{X} \in \mathbb{R}^{n \times n}$ .

(1a)  $\odot$  Show that for a fixed matrix  $\mathbf{A} \in \mathbb{R}^{n,n}$  the mapping

$$L: \left\{ \begin{array}{ccc} \mathbb{R}^{n,n} & \rightarrow & \mathbb{R}^{n,n} \\ \mathbf{X} & \mapsto & \mathbf{A}\mathbf{X} + \mathbf{X}\mathbf{A}^T \end{array} \right.$$

is linear.

HINT: Recall from linear algebra the definition of a linear mapping between two vector spaces.

In the sequel let  $\operatorname{vec}(\mathbf{M}) \in \mathbb{R}^{n^2}$  denote the column vector obtained by reinterpreting the internal coefficient array of a matrix  $M \in \mathbb{R}^{n,n}$  stored in column major format as the data array of a vector with  $n^2$  components. In MATLAB,  $\operatorname{vec}(\mathbf{M})$  would be the column vector obtained by  $\operatorname{reshape}(M, n*n, 1)$  or by M(:). See [1, Rem. 1.2.18] for the implementation with Eigen.

Problem (5) is equivalent to a linear system of equations

$$Cvec(X) = b (6)$$

with system matrix  $\mathbf{C} \in \mathbb{R}^{n^2,n^2}$  and right hand side vector  $\mathbf{b} \in \mathbb{R}^{n^2}$ .

- (1b) ⊡ Refresh yourself on the notion of "sparse matrix", see [1, Section 1.7] and, in particular, [1, Notion 1.7.1], [1, Def. 1.7.3].
- (1c)  $\odot$  Determine C and b from (6) for n = 2 and

$$\mathbf{A} = \begin{bmatrix} 2 & 1 \\ -1 & 3 \end{bmatrix}.$$

- (1d)  $\odot$  Use the Kronecker product to find a general expression for C in terms of a general A.
- (1e) Urite a MATLAB function

function 
$$C = buildC$$
 (A)

that returns the matrix C from (6) when given a square matrix A. (The function kron may be used.)

(1f)  $\square$  Give an upper bound (as sharp as possible) for  $nnz(\mathbb{C})$  in terms of  $nnz(\mathbb{A})$ . Can  $\mathbb{C}$  be legitimately regarded as a sparse matrix for large n even if  $\mathbb{A}$  is dense?

HINT: Run the following MATLAB code:

```
n=4;
A=sym('A',[n,n]);
I=eye(n);
C=kron(A,I)+kron(I,A)
```

```
Eigen::SparseMatrix<double> buildC(const MatrixXd &A)
```

that builds the Eigen matrix C from A. Make sure that initialization is done efficiently using an intermediate triplet format. Read [1, Section 1.7.3] very carefully before starting.

(1h)  $\odot$  Validate the correctness of your C++ implementation of buildC by comparing with the equivalent Matlab function for n=5 and

$$A = \begin{bmatrix} 10 & 2 & 3 & 4 & 5 \\ 6 & 20 & 8 & 9 & 1 \\ 1 & 2 & 30 & 4 & 5 \\ 6 & 7 & 8 & 20 & 0 \\ 1 & 2 & 3 & 4 & 10 \end{bmatrix}.$$

(1i) Write a C++ function

void solveLyapunov (const MatrixXd & A, MatrixXd & X) that returns the solution of (5) in the  $n \times n$ -matrix  $\mathbf{X}$ , if  $A \in \mathbb{R}^{n,n}$ .

(1j) ☐ Test your C++ implementation of solveLyapunov by comparing with Matlab for the test case proposed in (1h).

## **Problem 2.** Partitioned Matrix (core problem)

Based on the block view of matrix multiplication presented in [1, § 1.3.13], we looked a *block elimination* for the solution of block partitioned linear systems of equations in [1, § 1.6.91]. Also of interest are [1, Rem. 1.6.45] and [1, Rem. 1.6.43] where LU-factorization is viewed from a block perspective. Closely related to this problem is [1, Ex. 1.6.94], which you should study again as warm-up to this problem.

Let the matrix  $\mathbf{A} \in \mathbb{R}^{n+1,n+1}$  be partitioned according to

$$\mathbf{A} = \begin{bmatrix} \mathbf{R} & \mathbf{v} \\ \mathbf{u}^T & 0 \end{bmatrix}, \tag{7}$$

where  $\mathbf{v} \in \mathbb{R}^n$ ,  $\mathbf{u} \in \mathbb{R}^n$ , and  $\mathbf{R} \in \mathbb{R}^{n \times n}$  is upper triangular and regular.

- (2a)  $\odot$  Give a necessary and sufficient condition for the triangular matrix R to be invertible.
- (2b)  $\Box$  Determine expressions for the subvectors  $\mathbf{z} \in \mathbb{R}^n, \xi \in \mathbb{R}$  of the solution vector of the linear system of equations

$$\begin{bmatrix} \mathbf{R} & \mathbf{v} \\ \mathbf{u}^T & 0 \end{bmatrix} \begin{bmatrix} \mathbf{z} \\ \xi \end{bmatrix} = \begin{bmatrix} \mathbf{b} \\ \beta \end{bmatrix}$$

for arbitrary  $\mathbf{b} \in \mathbb{R}^n$ ,  $\beta \in \mathbb{R}$ .

HINT: Use blockwise Gaussian elimination as presented in [1, § 1.6.91].

- (2c) Show that A is regular if and only if  $\mathbf{u}^T \mathbf{R}^{-1} \mathbf{v} \neq 0$ .
- (2d) ⊡ Implement the C++ function

```
template <class Matrix, class Vector>
void solvelse(const Matrix & R, const Vector & v, const
Vector & u, const Vector & b, Vector & x);
```

for computing the solution of Ax = b (with A as in (7)) efficiently. Perform size check on input matrices and vectors.

HINT: Use the decomposition from (2b).

HINT: you can rely on the triangularView() function to instruct EIGEN of the triangular structure of  $\mathbf{R}$ , see [1, Code 1.2.12].

HINT: using the construct:

```
typedef typename Matrix::Scalar Scalar;
```

you can obtain the scalar type of the Matrix type (e.g. double for MatrixXd). This can then be used as:

```
Scalar a = 5;
```

HINT: using triangularView and templates you may incur in weird compiling errors. If this happens to you, check http://eigen.tuxfamily.org/dox/TopicTemplateKeyword.html

HINT: sometimes the C++ keyword auto (only in std. C++11) can be used if you do not want to explicitly write the return type of a function, as in:

```
MatrixXd a;
auto b = 5*a;
```

HINT: Check the page http://eigen.tuxfamily.org/dox/group\_\_TutorialLinearAlgebra.html.

(2f)  $\Box$  What is the asymptotic complexity of your implementation of solvelse () in terms of problem size parameter  $n \to \infty$ ?

#### Problem 3. Banded matrix

For  $n \in \mathbb{N}$  we consider the matrix

$$\mathbf{A} := \begin{bmatrix} 2 & a_1 & 0 & \dots & \dots & 0 \\ 0 & 2 & a_2 & 0 & \dots & \dots & 0 \\ b_1 & 0 & \ddots & \ddots & \ddots & \ddots & \vdots \\ 0 & b_2 & \ddots & \ddots & \ddots & \ddots & \vdots \\ \vdots & 0 & \ddots & \ddots & \ddots & \ddots & 0 \\ \vdots & \vdots & \ddots & \ddots & \ddots & \ddots & a_{n-1} \\ 0 & 0 & \dots & 0 & b_{n-2} & 0 & 2 \end{bmatrix} \in \mathbb{R}^{n,n}$$

with  $a_i, b_i \in \mathbb{R}$ .

*Remark.* The matrix **A** is an instance of a banded matrix, see [1, Section 1.7.6] and, in particular, the examples after [1, Def. 1.7.53]. However, you need not know any of the content of this section for solving this problem.

(3a) Implement an *efficient* C++ function:

```
template <class Vector>
void multAx(const Vector & a, const Vector & b, const
Vector & x, Vector & y);
```

for the computation of y = Ax.

(3b)  $\square$  Show that **A** is invertible if  $a_i, b_i \in [0, 1]$ .

HINT: Give an indirect proof that  $\ker \mathbf{A}$  is trivial, by looking at the largest (in modulus) component of an  $\mathbf{x} \in \ker \mathbf{A}$ .

*Remark.* That **A** is invertible can immediately be concluded from the general fact that kernel vectors of irreducible, diagonally dominant matrices ( $\rightarrow$  [1, Def. 1.8.8]) must be multiples of  $[1,1,\ldots,1]^{\mathsf{T}}$ . Actually, the proof recommended in the hint shows this fact first before bumping into a contradiction.

(3c)  $\blacksquare$  Fix  $b_i = 0, \forall i = 1, ..., n-2$ . Implement an efficient C++ function

```
template <class Vector>
void solvelseAupper(const Vector & a, const Vector &
r, Vector & x);
```

solving Ax = r.

(3d)  $\odot$  For general  $a_i, b_i \in [0, 1]$  devise an efficient C++ function:

```
template <class Vector>
void solvelseA(const Vector & a, const Vector & b,
const Vector & r, Vector & x);
```

that computes the solution of Ax = r by means of Gaussian elimination. You cannot use any high level solver routines of EIGEN.

HINT: Thanks to the constraint  $a_i, b_i \in [0, 1]$ , pivoting is not required in order to ensure stability of Gaussian elimination. This is asserted in [1, Lemma 1.8.9], but you may just use this fact here. Thus, you can perform a straightforward Gaussian elimination from top to bottom as you have learned it in your linear algebra course.

- (3e)  $\odot$  What is the asymptotic complexity of your implementation of solvelseA for  $n \to \infty$ .
- (3f) Implement solvelseAEigen as in (3d), this time using EIGEN's sparse elimination solver.

HINT: The standard way of initializing a sparse EIGEN-matrix efficiently, is via the triplet format as discussed in [1, Section 1.7.3]. You may also use direct initialization of a sparse matrix, provided that you reserve() enough space for the non-zero entries of each column, see documentation.

# **Problem 4.** Sequential linear systems

This problem is about a sequence of linear systems, please see [1, Rem. 1.6.85]. The idea is that if we solve several linear systems with the same matrix A, that computational cost may be reduced by performing the LU decomposition only once.

Consider the following MATLAB function with input data  $\mathbf{A} \in \mathbb{R}^{n,n}$  and  $\mathbf{b} \in \mathbb{R}^n$ .

```
function X = solvepermb(A,b)
[n,m] = size(A);
if ((n + numel(b)) || (m + numel(b))), error('Size mismatch'); end

X = [];
for l=1:n
    X = [X,A\b];
    b = [b(end);b(1:end-1)];
end
```

- (4a)  $\odot$  What is the asymptotic complexity of this function as  $n \to \infty$ ?
- (4b) Port the MATLAB function solvepermb to C++ using EIGEN. (This means that the C++ code should perform exactly the same computations in exactly the same order.)
- (4c)  $\odot$  Design an efficient implementation of this function with asymptotic complexity  $O(n^3)$  in Eigen.

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#### References

[1] R. Hiptmair. *Lecture slides for course "Numerical Methods for CSE"*. http://www.sam.math.ethz.ch/~hiptmair/tmp/NumCSE/NumCSE15.pdf. 2015.