

Numerical Methods for Computational Science and Engineering

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URL: <http://www.sam.math.ethz.ch/~hiptmair/tmp/NumCSE/NumCSE15.pdf>

Introduction

- Use the study center HG E 41, Mon from 18:00
- A toolbox course \rightarrow many different topics, that are only loosely related

[toolbox contains hammer, screwdriver, duct tape]

I. Computing with Matrices and Vectors
 (some aspects of numerical linear algebra)
 foundation of all numerical codes

I.1. Fundamentals

I.1.1. Notations:

Vector $\underline{v} = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \in \mathbb{K}^n$, $\mathbb{K} = \mathbb{R}, \mathbb{C}$ (default: column vector)

Matrix $A = \begin{bmatrix} a_{1,1} & \dots & a_{1,n} \\ \vdots & \square & \vdots \\ a_{m,1} & \dots & a_{m,n} \end{bmatrix} \in \mathbb{K}^{m,n}$

row vector $\underline{v}^T = [v_1, \dots, v_n]$

↳ just special matrices $\in \mathbb{K}^{1,n}$

Component/entries: $(\underline{v})_i := v_i$, $(A)_{i,j} := a_{ij}$

sub-vector/matrix blocks: $(\underline{v})_{k:l} = \begin{bmatrix} v_k \\ \vdots \\ v_l \end{bmatrix}$, $1 \leq k \leq l \leq n$
 $(A)_{k:l, r:s} \in \mathbb{K}^{l-k+1, s-r+1}$

$$A := \begin{bmatrix} a_{11} & \dots & a_{1m} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nm} \end{bmatrix}$$

→ entry $(A)_{i,j} = a_{ij}$, $1 \leq i \leq n, 1 \leq j \leq m$,

→ i -th row, $1 \leq i \leq n$: $a_{i,:} = (A)_{i,:}$,

→ j -th column, $1 \leq j \leq m$: $a_{:,j} = (A)_{:,j}$,

→ **matrix block** $(a_{ij})_{\substack{i=k,\dots,l \\ j=r,\dots,s}} = (A)_{k:l, r:s}$, $1 \leq k \leq l \leq n$, $1 \leq r \leq s \leq m$.
(sub-matrix)

1.2. Software and Libraries

→ Always rely on them for numerical linear algebra

1.2.1. MATLAB

↳ IDE for numerical computation
(numerical engine, libraries, editor, debugger, profiler, help)

"In MATLAB everything is a matrix"

(Fundamental "data type" in MATLAB = **matrix** of complex numbers)

↳ non-typed language

$[m,n] = \text{size}(A) \rightarrow$ request dimensions of matrix
 $A(i,j), v(i) \rightarrow$ entry access

Initialization of matrix:

$;$ \leftrightarrow horizontal
 $;$ \leftrightarrow vertical
} concatenation

Example:

Output: M=

1	2	3	0	0	0
4	5	6	0	0	0
0	0	0	0	0	0
0	0	0	0	0	1

% Caution: matrices are dynamically expanded when
% out of range entries are accessed
M = [1,2,3;4,5,6]; M(4,6) = 1.0; M,

"loop index vectors" $v = (a:s:b)$

```
>> v = (3:-0.5:-0.3)
v = 3.0000 2.5000 2.0000 1.5000 1.0000 0.5000 0
>> v = (1:2.5:-13)
v = Empty matrix: 1-by-0
```

⇒ loop: for $i = (a:s:b)$

```
% MATLAB loop over columns of a matrix
```

```
M = [1,2,3;4,5,6];
```

```
for i = M; i, end
```

Output:

```
i = 1    i = 2    i = 3
    4      5      6
```

1.2.2. Eigen

= Header-only C++ library for numerical algebra,
→ template metaprogramming
→ expression templates

Fundamental data type : matrix

```
Matrix<typename Scalar, int RowsAtCompileTime, int  
      ColsAtCompileTime>
```

↑ ↑
for specifying small fixed size matrices

Special data types : MatrixX^* , $^* \in \{i, f, d, cd\}$

```
#include <Eigen/Dense >
```

```
template<typename Scalar>
```

```
void eigenTypeDemo(unsigned int dim)
```

```
{
```

```
    using dynMat_t =
```

```
        Eigen::Matrix<Scalar, Eigen::Dynamic, Eigen::Dynamic>;
```

```
    using dynColVec_t = Eigen::Matrix<Scalar, Eigen::Dynamic, 1>;
```

```
    using dynRowVec_t = Eigen::Matrix<Scalar, 1, Eigen::Dynamic>;
```

```
    using index_t = typename dynMat_t::Index;
```

```
    using entry_t = typename dynMat_t::Scalar;
```

```
    dynColVec_t colvec(dim);
```

```
    dynRowVec_t rowvec(dim);
```

```
    for(index_t i=0; i< colvec.size(); ++i) colvec(i) = (Scalar)i;
```

```
    for(index_t i=0; i< rowvec.size(); ++i) rowvec(i) =
```

```
        (Scalar)1/(i+1);
```

```
    dynMat_t vecprod = colvec*rowvec; ← matrix product
```

```
    const int nrows = vecprod.rows();
```

```
    const int ncols = vecprod.cols();
```

```
}
```

Initialization of :
matrices

```
#include <Eigen/Dense >
```

```
// Just allocate space for matrix, no initialisation
```

```
Eigen::MatrixXd A(rows,cols); ← reserve space
```

```
// Zero matrix. Similar to matlab command zeros(rows,cols);
```

```
Eigen::MatrixXd B = MatrixXd::Zero(rows, cols);
```

```
// Ones matrix. Similar to matlab command ones(rows,cols);
```

```
Eigen::MatrixXd C = MatrixXd::Ones(rows, cols);
```

```
// Matrix with all entries same as value.
```

```
Eigen::MatrixXd D = MatrixXd::Constant(rows, cols, value);
```

```
// Random matrix, entries uniformly distributed in [0,1]
```

```
Eigen::MatrixXd E = MatrixXd::Random(rows, cols);
```

```
// (Generalized) identity matrix, 1 on main diagonal
```

```
Eigen::MatrixXd I = MatrixXd::Identity(rows,cols);
```

```
std::cout << "size of A = (" << A.rows() << ',' << A.cols() << '))'
```

```
<< std::endl;
```

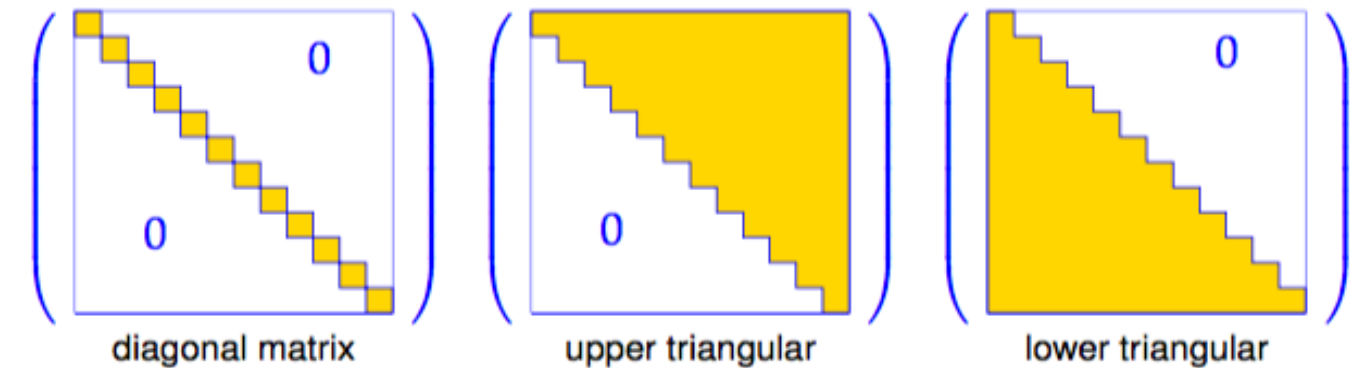
Addressing sub-matrices :

```
template<typename MatType> void
blockAccess(Eigen::MatrixBase<MatType> &M)
{
    using index_t = typename Eigen::MatrixBase<MatType>::Index;
    using entry_t = typename Eigen::MatrixBase<MatType>::Scalar;
    const index_t nrows(M.rows()); % No. of rows
    const index_t ncols(M.cols()); % No. of columns

    cout << "Matrix M = " endl << M << endl; // Print matrix
    // Block size half the size of the matrix
    index_t p = nrows/2, q = ncols/2;
    // Output submatrix with left upper entry at position (i,i)
    for(index_t i=0; i < min(p,q); i++)
        cout << "Block (" << i << ', ' << i << ', ' << p << ', ' << q
            << ") = " << M.block(i, i, p, q) << endl;
    // l-value access: modify sub-matrix by adding a constant
    M.block(1,1,p,q) += MatrixXd::Constant(p,q,1.0); ←
    cout << "M = " endl << M << endl;
    // r-value access: extract sub-matrix
    MatrixXd B(M.block(1,1,p,q));
    cout << "Isolated modified block = " endl << B << endl;
    // Special sub-matrices
    cout << p << " top rows of m = " << M.topRows(p) << endl;
    cout << p << " bottom rows of m = " << M.bottomRows(p) << endl;
    cout << q << " left cols of m = " << M.leftCols(q) << endl;
    cout << q << " right cols of m = " << M.rightCols(p) << endl;
    // r-value access to upper triangular part
    const MatrixXd T = M.template triangularView<Upper>(); //
    cout << "Upper triangular part = " << endl << T << endl;
    // l-value access to upper triangular part
    M.template triangularView<Lower>() *= -1.5; //
    cout << "Matrix M = " << endl << M << endl;
}
```

In Eigen : indexing from 0 !

Triangular matrices :



1.2.3. Matrix storage formats

→ Matrix stored in linearized form (as a vector)

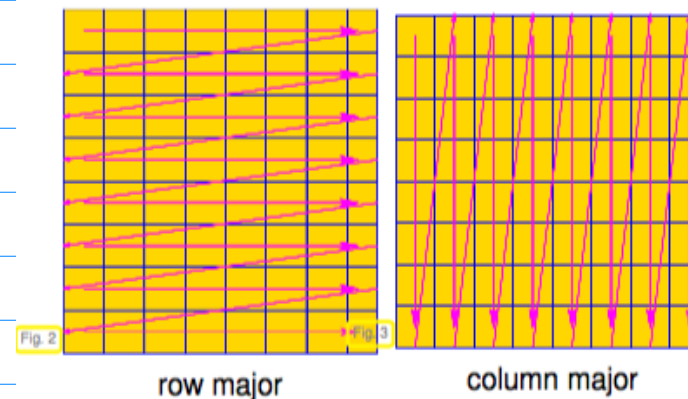
$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$

Row major (C-arrays, bitmaps, Python):

A_arr	1	2	3	4	5	6	7	8	9
-------	---	---	---	---	---	---	---	---	---

Column major (Fortran, MATLAB, EIGEN):

A_arr	1	4	7	2	5	8	3	6	9
-------	---	---	---	---	---	---	---	---	---



Index mappings $A \in \mathbb{K}^{n,m}$

r.m.: $(i,j) \leftrightarrow m(i-1)+j$


```
// Template parameter ColMajor selects column major data layout
Matrix<double, Dynamic, Dynamic, ColMajor> mcm(nrows, ncols); → default
// Template parameter RowMajor selects row major data layout
Matrix<double, Dynamic, Dynamic, RowMajor> mrm(nrows, ncols);
```

Why care? → Calling library functions

Exp.: Impact of storage format on runtime
1.2.20

Matlab: Column major

<pre>A = randn(n,n); for j = 1:n-1, A(:,j+1) = A(:,j+1) - A(:,j); end</pre> <p>column oriented access</p>	<pre>A = randn(n,n); for i = 1:n-1, A(i+1,:) = A(i+1,:) - A(i,:); end</pre> <p>row oriented access</p>
---	--

↓
faster, because contiguous
access to memory

↓
many cache misses

