

Quick reference guide

Dense matrix and array manipulation

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Modules and Header files

The **Eigen** library is divided in a Core module and several additional modules. Each module has a corresponding header file which has to be included in order to use the module. The `Dense` and **Eigen** header files are provided to conveniently gain access to several modules at once.

Module	Header file	Contents
Core	<code>#include <Eigen/Core></code>	Matrix and Array classes, basic linear algebra (including triangular and selfadjoint products), array manipulation
Geometry	<code>#include <Eigen/Geometry></code>	Transform , Translation , Scaling , Rotation2D and 3D rotations (Quaternion , AngleAxis)
LU	<code>#include <Eigen/LU></code>	Inverse, determinant, LU decompositions with solver (FullPivLU , PartialPivLU)
Cholesky	<code>#include <Eigen/Cholesky></code>	LLT and LDLT Cholesky factorization with solver
Householder	<code>#include <Eigen/Householder></code>	Householder transformations; this module is used by several linear algebra modules
SVD	<code>#include <Eigen/SVD></code>	SVD decomposition with least-squares solver (JacobiSVD)
QR	<code>#include <Eigen/QR></code>	QR decomposition with solver (HouseholderQR , ColPivHouseholderQR , FullPivHouseholderQR)
Eigenvalues	<code>#include <Eigen/Eigenvalues></code>	Eigenvalue, eigenvector decompositions (EigenSolver , SelfAdjointEigenSolver , ComplexEigenSolver)
Sparse	<code>#include <Eigen/Sparse></code>	Sparse matrix storage and related basic linear algebra (SparseMatrix , DynamicSparseMatrix , SparseVector)
	<code>#include <Eigen/Dense></code>	Includes Core, Geometry, LU, Cholesky, SVD, QR, and Eigenvalues header files
	<code>#include <Eigen/Eigen></code>	Includes Dense and Sparse header files (the whole Eigen library)

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Array, matrix and vector types

Recall: **Eigen** provides two kinds of dense objects: mathematical matrices and vectors which are both represented by the template class **Matrix**, and general 1D and 2D arrays represented by the template class **Array**:

```
typedef Matrix<Scalar, RowsAtCompileTime, ColsAtCompileTime, Options> MyMatrixType;
typedef Array<Scalar, RowsAtCompileTime, ColsAtCompileTime, Options> MyArrayType;
```

- `Scalar` is the scalar type of the coefficients (e.g., `float`, `double`, `bool`, `int`, etc.).
- `RowsAtCompileTime` and `ColsAtCompileTime` are the number of rows and columns of the matrix as known at compile-time or `Dynamic`.
- `Options` can be `ColMajor` or `RowMajor`, default is `ColMajor`. (see class **Matrix** for more options)

All combinations are allowed: you can have a matrix with a fixed number of rows and a dynamic number of columns, etc. The following are all valid:

```
Matrix<double, 6, Dynamic>           // Dynamic number of columns (heap allocation)
Matrix<double, Dynamic, 2>           // Dynamic number of rows (heap allocation)
Matrix<double, Dynamic, Dynamic, RowMajor> // Fully dynamic, row major (heap allocation)
Matrix<double, 13, 3>                 // Fully fixed (usually allocated on stack)
```

In most cases, you can simply use one of the convenience typedefs for **matrices** and **arrays**. Some examples:

Matrices

```
Matrix<float,Dynamic,Dynamic>  <=>  MatrixXf
Matrix<double,Dynamic,1>       <=>  VectorXd
Matrix<int,1,Dynamic>          <=>  RowVectorXi
Matrix<float,3,3>              <=>  Matrix3f
Matrix<float,4,1>              <=>  Vector4f
```

Arrays

```
Array<float,Dynamic,Dynamic>  <=>  ArrayXXf
Array<double,Dynamic,1>       <=>  ArrayXd
Array<int,1,Dynamic>          <=>  RowArrayXi
Array<float,3,3>              <=>  Array33f
Array<float,4,1>              <=>  Array4f
```

Conversion between the matrix and array worlds:

```
Array44f a1, a1;
Matrix4f m1, m2;
m1 = a1 * a2;           // coeffwise product, implicit conversion from array to matrix.
a1 = m1 * m2;           // matrix product, implicit conversion from matrix to array.
a2 = a1 + m1.array();   // mixing array and matrix is forbidden
m2 = a1.matrix() + m1;  // and explicit conversion is required.
ArrayWrapper<Matrix4f> mla(m1); // mla is an alias for m1.array(), they share the same coefficients
MatrixWrapper<Array44f> alm(a1);
```

In the rest of this document we will use the following symbols to emphasize the features which are specifics to a given kind of object:

- * linear algebra matrix and vector only
- * array objects only

Basic matrix manipulation

	1D objects	2D objects	Notes
Constructors	<pre>Vector4d v4; Vector2f v1(x, y); Array3i v2(x, y, z); Vector4d v3(x, y, z, w); VectorXf v5; // empty object ArrayXf v6(size);</pre>	<pre>Matrix4f m1; MatrixXf m5; // empty object MatrixXf m6(nb_rows, nb_columns);</pre>	By default, the coefficients are left uninitialized
Comma initializer	<pre>Vector3f v1; v1 << x, y, z; ArrayXf v2(4); v2 << 1, 2, 3, 4;</pre>	<pre>Matrix3f m1; m1 << 1, 2, 3, 4, 5, 6, 7, 8, 9;</pre>	
Comma initializer (bis)	<pre>int rows=5, cols=5; MatrixXf m(rows,cols); m << (Matrix3f() << 1, 2, 3, 4, 5, 6, 7, 8, 9).finished(), MatrixXf::Zero(3,cols-3), MatrixXf::Zero(rows-3,3), MatrixXf::Identity(rows-3,cols-3); cout << m;</pre>		<p>output:</p> <pre>1 2 3 0 0 4 5 6 0 0 7 8 9 0 0 0 0 0 1 0 0 0 0 0 1</pre>
Runtime info	<pre>vector.size(); vector.innerStride(); vector.data();</pre>	<pre>matrix.rows(); matrix.cols(); matrix.innerSize(); matrix.outerSize(); matrix.innerStride(); matrix.outerStride(); matrix.data();</pre>	Inner/Outer* are storage order dependent

Compile-time info	ObjectType::Scalar ObjectType::RealScalar ObjectType::Index	ObjectType::RowsAtCompileTime ObjectType::ColsAtCompileTime ObjectType::SizeAtCompileTime	
Resizing	<pre>vector.resize(size); vector.resizeLike(other_vector); vector.conservativeResize(size);</pre>	<pre>matrix.resize(nb_rows, nb_cols); matrix.resize(Eigen::NoChange, nb_cols); matrix.resize(nb_rows, Eigen::NoChange); matrix.resizeLike(other_matrix); matrix.conservativeResize(nb_rows, nb_cols);</pre>	no-op if the new sizes match, otherwise data are lost resizing with data preservation
Coeff access with range checking	<pre>vector(i) vector.x() vector[i] vector.y() vector.z() vector.w()</pre>	<code>matrix(i,j)</code>	Range checking is disabled if <code>NDEBUG</code> or <code>EIGEN_NO_DEBUG</code> is defined
Coeff access without range checking	<pre>vector.coeff(i) vector.coeffRef(i)</pre>	<pre>matrix.coeff(i,j) matrix.coeffRef(i,j)</pre>	
Assignment/copy	<pre>object = expression; object_of_float = expression_of_double.cast<float>();</pre>		the destination is automatically resized (if possible)

Predefined Matrices

Fixed-size matrix or vector

```
typedef {Matrix3f|Array33f}
        FixedXD;
FixedXD x;

x = FixedXD::Zero();
x = FixedXD::Ones();
x = FixedXD::Constant(value);
x = FixedXD::Random();
x = FixedXD::LinSpaced(size, low,
                      high);

x.setZero();
x.setOnes();
x.setConstant(value);
x.setRandom();
x.setLinSpaced(size, low, high);
```

Identity and basis vectors *

```
x = FixedXD::Identity();
x.setIdentity();

Vector3f::UnitX() // 1 0 0
Vector3f::UnitY() // 0 1 0
Vector3f::UnitZ() // 0 0 1
```

Dynamic-size matrix

```
typedef {MatrixXf|ArrayXXf}
        Dynamic2D;
Dynamic2D x;

x = Dynamic2D::Zero(rows, cols);
x = Dynamic2D::Ones(rows, cols);
x = Dynamic2D::Constant(rows, cols,
                        value);
x = Dynamic2D::Random(rows, cols);
N/A

x.setZero(rows, cols);
x.setOnes(rows, cols);
x.setConstant(rows, cols, value);
x.setRandom(rows, cols);
N/A
```

```
x = Dynamic2D::Identity(rows,
                       cols);
x.setIdentity(rows, cols);
```

N/A

Dynamic-size vector

```
typedef {VectorXf|ArrayXf}
        Dynamic1D;
Dynamic1D x;

x = Dynamic1D::Zero(size);
x = Dynamic1D::Ones(size);
x = Dynamic1D::Constant(size,
                        value);
x = Dynamic1D::Random(size);
x = Dynamic1D::LinSpaced(size,
                        low, high);

x.setZero(size);
x.setOnes(size);
x.setConstant(size, value);
x.setRandom(size);
x.setLinSpaced(size, low, high);
```

N/A

```
VectorXf::Unit(size,i)
VectorXf::Unit(4,1) ==
    Vector4f(0,1,0,0)
    ==
```

Vector4f::UnitY()

Mapping external arrays

Contiguous memory	<pre>float data[] = {1,2,3,4}; Map<Vector3f> v1(data); // uses v1 as a Vector3f object Map<ArrayXf> v2(data,3); // uses v2 as a ArrayXf object Map<Array22f> m1(data); // uses m1 as a Array22f object Map<MatrixXf> m2(data,2,2); // uses m2 as a MatrixXf object</pre>
Typical usage of strides	<pre>float data[] = {1,2,3,4,5,6,7,8,9}; Map<VectorXf,0,InnerStride<2>> > v1(data,3); // = [1,3,5] Map<VectorXf,0,InnerStride<>> > v2(data,3,InnerStride<>(3)); // = [1,4,7] Map<MatrixXf,0,OuterStride<3>> > m2(data,2,3); // both lines Map<MatrixXf,0,OuterStride<>> > m1(data,2,3,OuterStride<>(3)); // are equal to: 1,4,7 2,5,8 </pre>

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Arithmetic Operators

add	<code>mat3 = mat1 + mat2;</code>	<code>mat3 += mat1;</code>	
subtract	<code>mat3 = mat1 - mat2;</code>	<code>mat3 -= mat1;</code>	
scalar product	<pre>mat3 = mat1 * s1; mat3 = mat1 / s1;</pre>	<pre>mat3 *= s1; mat3 /= s1;</pre>	<code>mat3 = s1 * mat1;</code>
matrix/vector products *	<pre>col2 = mat1 * col1; row2 = row1 * mat1; mat3 = mat1 * mat2;</pre>	<pre>row1 *= mat1; mat3 *= mat1;</pre>	
transposition	<code>mat1 = mat2.transpose();</code>	<code>mat1.transposeInPlace();</code>	
adjoint *	<code>mat1 = mat2.adjoint();</code>	<code>mat1.adjointInPlace();</code>	
dot product	<code>scalar = vec1.dot(vec2);</code>		
inner product *	<pre>scalar = col1.adjoint() * col2; scalar = (col1.adjoint() * col2).value();</pre>		
outer product *	<code>mat = col1 * col2.transpose();</code>		
norm	<code>scalar = vec1.norm();</code>	<code>scalar = vec1.squaredNorm();</code>	
normalization *	<code>vec2 = vec1.normalized();</code>	<code>vec1.normalize(); // inplace</code>	
cross product *	<pre>#include <Eigen/Geometry> vec3 = vec1.cross(vec2);</pre>		

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Coefficient-wise & Array operators

Coefficient-wise operators for matrices and vectors:

Matrix API *	Via Array conversions
<code>mat1.cwiseMin(mat2)</code>	<code>mat1.array().min(mat2.array())</code>
<code>mat1.cwiseMax(mat2)</code>	<code>mat1.array().max(mat2.array())</code>
<code>mat1.cwiseAbs2()</code>	<code>mat1.array().abs2()</code>
<code>mat1.cwiseAbs()</code>	<code>mat1.array().abs()</code>
<code>mat1.cwiseSqrt()</code>	<code>mat1.array().sqrt()</code>
<code>mat1.cwiseProduct(mat2)</code>	<code>mat1.array() * mat2.array()</code>
<code>mat1.cwiseQuotient(mat2)</code>	<code>mat1.array() / mat2.array()</code>

It is also very simple to apply any user defined function `f` using `DenseBase::unaryExpr` together with `std::ptr_fun`:

```
mat1.unaryExpr(std::ptr_fun(foo))
```

Array operators:*

Arithmetic operators	array1 * array2 array1 + scalar	array1 / array2 array1 - scalar	array1 *= array2 array1 += scalar	array1 /= array2 array1 -= scalar
Comparisons	array1 < array2 array1 <= array2 array1 == array2	array1 > array2 array1 >= array2 array1 != array2	array1 < scalar array1 <= scalar array1 == scalar	array1 > scalar array1 >= scalar array1 != scalar
Trigo, power, and misc functions and the STL variants	<pre>array1.min(array2) array1.max(array2) array1.abs2() array1.abs() array1.sqrt() array1.log() array1.exp() array1.pow(exponent) array1.square() array1.cube() array1.inverse() array1.sin() array1.cos() array1.tan() array1.asin() array1.acos()</pre>			
		abs(array1) sqrt(array1) log(array1) exp(array1) pow(array1,exponent)		
		sin(array1) cos(array1) tan(array1) asin(array1) acos(array1)		

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Reductions

Eigen provides several reduction methods such as: **minCoeff()**, **maxCoeff()**, **sum()**, **prod()**, **trace()** *, **norm()** *, **squaredNorm()** *, **all()**, and **any()**. All reduction operations can be done matrix-wise, **column-wise** or **row-wise**. Usage example:

<pre>mat = 5 3 1 2 7 8 9 4 6</pre>	<code>mat.minCoeff();</code>	1
	<code>mat.colwise().minCoeff();</code>	2 3 1
	<code>mat.rowwise().minCoeff();</code>	1 2 4

Special versions of **minCoeff** and **maxCoeff** :

```
int i, j;
s = vector.minCoeff(&i);           // s == vector[i]
s = matrix.maxCoeff(&i, &j);       // s == matrix(i,j)
```

Typical use cases of **all()** and **any()**:

```
if((array1 > 0).all()) ... // if all coefficients of array1 are greater than 0 ...
if((array1 < array2).any()) ... // if there exist a pair i,j such that array1(i,j) < array2(i,j) ...
```

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Sub-matrices

Read-write access to a **column** or a **row** of a matrix (or array):

```
mat1.row(i) = mat2.col(j);
mat1.col(j1).swap(mat1.col(j2));
```

Read-write access to sub-vectors:

Optimized versions when the size

Default versions**is known at compile time**`vec1.head(n)``vec1.head<n>()`the first n coeffs`vec1.tail(n)``vec1.tail<n>()`the last n coeffs`vec1.segment(pos, n)``vec1.segment<n>(pos)`the n coeffs in the range $[pos : pos + n - 1]$

Read-write access to sub-matrices:

`mat1.block(i, j, rows, cols)``mat1.block<rows, cols>(i, j)`the $rows \times cols$ sub-matrix starting from position (i, j) **(more)****(more)**`mat1.topLeftCorner(rows, cols)
mat1.topRightCorner(rows, cols)
mat1.bottomLeftCorner(rows, cols)
mat1.bottomRightCorner(rows, cols)``mat1.topLeftCorner<rows, cols>()
mat1.topRightCorner<rows, cols>()
mat1.bottomLeftCorner<rows, cols>()
mat1.bottomRightCorner<rows, cols>()`the $rows \times cols$ sub-matrix taken in one of the four corners`mat1.topRows(rows)
mat1.bottomRows(rows)
mat1.leftCols(cols)
mat1.rightCols(cols)``mat1.topRows<rows>()
mat1.bottomRows<rows>()
mat1.leftCols<cols>()
mat1.rightCols<cols>()`specialized versions of `block()` when the block fit two corners[top](#)

Miscellaneous operations

Reverse

Vectors, rows, and/or columns of a matrix can be reversed (see [DenseBase::reverse\(\)](#), [DenseBase::reverseInPlace\(\)](#), [VectorwiseOp::reverse\(\)](#)).

```
vec.reverse()           mat.colwise().reverse()   mat.rowwise().reverse()  
vec.reverseInPlace()
```

Replicate

Vectors, matrices, rows, and/or columns can be replicated in any direction (see [DenseBase::replicate\(\)](#), [VectorwiseOp::replicate\(\)](#))

```
vec.replicate(times)           vec.replicate<Times>  
mat.replicate(vertical_times, horizontal_times)  mat.replicate<VerticalTimes, HorizontalTimes>()  
mat.colwise().replicate(vertical_times, horizontal_times)  mat.colwise().replicate<VerticalTimes,  
    HorizontalTimes>()                                     mat.rowwise().replicate<VerticalTimes,  
mat.rowwise().replicate(vertical_times, horizontal_times)  HorizontalTimes>()
```

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Diagonal, Triangular, and Self-adjoint matrices

(matrix world ^{*})

Diagonal matrices

Operation

view a vector **as a diagonal matrix**

Declare a diagonal matrix

Access the **diagonal** and **super/sub diagonals** of a matrix as a vector (read/write)

Optimized products and inverse

Code

```
mat1 = vec1.asDiagonal();
```

```
DiagonalMatrix<Scalar,SizeAtCompileTime> diag1(size);
diag1.diagonal() = vector;
```

```
vec1 = mat1.diagonal();          mat1.diagonal() = vec1;
// main diagonal
vec1 = mat1.diagonal(+n);       mat1.diagonal(+n) = vec1;
// n-th super diagonal
vec1 = mat1.diagonal(-n);       mat1.diagonal(-n) = vec1;
// n-th sub diagonal
vec1 = mat1.diagonal<1>();       mat1.diagonal<1>() = vec1;
// first super diagonal
vec1 = mat1.diagonal<-2>();      mat1.diagonal<-2>() = vec1;
// second sub diagonal
```

```
mat3 = scalar * diag1 * mat1;
mat3 += scalar * mat1 * vec1.asDiagonal();
mat3 = vec1.asDiagonal().inverse() * mat1
mat3 = mat1 * diag1.inverse()
```

Triangular views

TriangularView gives a view on a triangular part of a dense matrix and allows to perform optimized operations on it. The opposite triangular part is never referenced and can be used to store other information.

Note

The `.triangularView()` template member function requires the `template` keyword if it is used on an object of a type that depends on a template parameter; see [The template and typename keywords in C++](#) for details.

Operation

Reference to a triangular with optional unit or null diagonal (read/write):

Writing to a specific triangular part:
(only the referenced triangular part is evaluated)

Conversion to a dense matrix setting the opposite triangular part to zero:

Products:

Solving linear equations:

$$M_2 := L_1^{-1} M_2$$

$$M_3 := L_1^{*-1} M_3$$

$$M_4 := M_4 U_1^{-1}$$

Code

```
m.triangularView<Xxx>()
```

Xxx = Upper, Lower, StrictlyUpper, StrictlyLower, UnitUpper, UnitLower

```
m1.triangularView<Eigen::Lower>() = m2 + m3
```

```
m2 = m1.triangularView<Eigen::UnitUpper>()
```

```
m3 += s1 * m1.adjoint().triangularView<Eigen::UnitUpper>()
      * m2
m3 -= s1 * m2.conjugate() *
      m1.adjoint().triangularView<Eigen::Lower>()
```

```
L1.triangularView<Eigen::UnitLower>().solveInPlace(M2)
L1.triangularView<Eigen::Lower>
  ().adjoint().solveInPlace(M3)
U1.triangularView<Eigen::Upper>().solveInPlace<OnTheRight>
  (M4)
```

Symmetric/selfadjoint views

Just as for triangular matrix, you can reference any triangular part of a square matrix to see it as a selfadjoint matrix and perform special and optimized operations. Again the opposite triangular part is never referenced and can be used to store other information.

Note

The `.selfadjointView()` template member function requires the `template` keyword if it is used on an object of a type that depends on a template parameter; see [The template and typename keywords in C++](#) for details.

Operation

Code

Conversion to a dense matrix:

```
m2 = m.selfadjointView<Eigen::Lower>();
```

Product with another general matrix or vector:

```
m3 = s1 * m1.conjugate().selfadjointView<Eigen::Upper>() * m3;
m3 -= s1 * m3.adjoint() * m1.selfadjointView<Eigen::Lower>();
```

Rank 1 and rank K update:

$upper(M_1) += s_1 M_2 M_2^*$
 $lower(M_1) -= M_2^* M_2$

```
M1.selfadjointView<Eigen::Upper>().rankUpdate(M2,s1);
M1.selfadjointView<Eigen::Lower>().rankUpdate(M2.adjoint(),-1);
```

Rank 2 update: ($M += suv^* + svu^*$)

```
M.selfadjointView<Eigen::Upper>().rankUpdate(u,v,s);
```

Solving linear equations:

($M_2 := M_1^{-1} M_2$)

```
// via a standard Cholesky factorization
m2 = m1.selfadjointView<Eigen::Upper>().llt().solve(m2);
// via a Cholesky factorization with pivoting
m2 = m1.selfadjointView<Eigen::Lower>().ldlt().solve(m2);
```