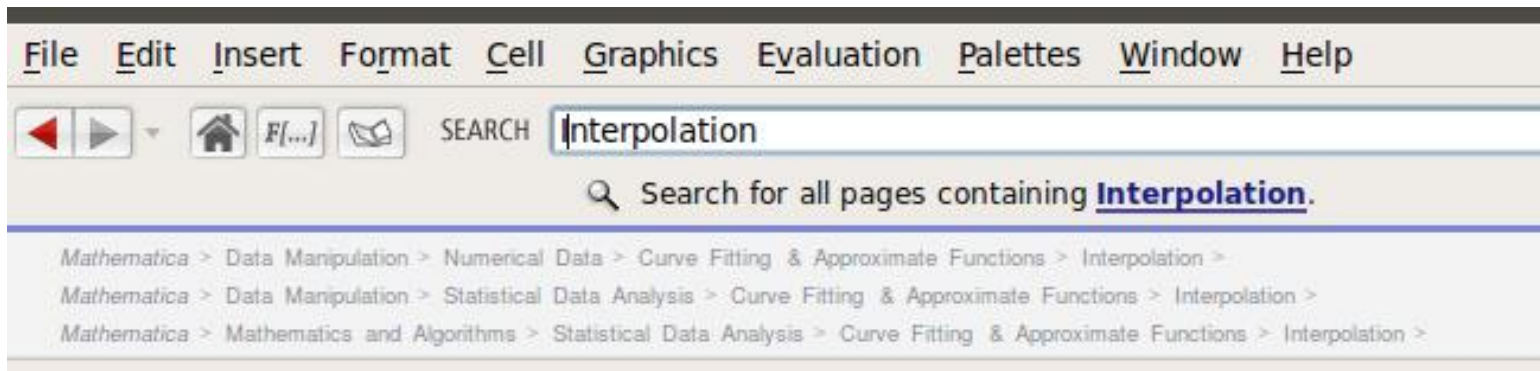


Mathematica (1)

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BUILT-IN MATHEMATICA SYMBOL

Interpolation

`Interpolation` $[\{f_1, f_2, \dots\}]$

constructs an interpolation of the function values f_i , assumed to correspond to x values $1, 2, \dots$.

`Interpolation` $[\{\{x_1, f_1\}, \{x_2, f_2\}, \dots\}]$

constructs an interpolation of the function values f_i corresponding to x values x_i .

`Interpolation` $[\{\{\{x_1, y_1, \dots\}, f_1\}, \{\{x_2, y_2, \dots\}, f_2\}, \dots\}]$

constructs an interpolation of multidimensional data.

`Interpolation` $[\{\{\{x_1, \dots\}, f_1, df_1, \dots\}, \dots\}]$

constructs an interpolation that reproduces derivatives as well as function values.

`Interpolation` $[data, x]$

find an interpolation of $data$ at the point x .

利用好help

找各种例子

多实践

```
In[1]:= InterpolatingPolynomial[{{-1, 4}, {0, 2}, {1, 6}}, x]
```

```
Out[1]= 4 + (1 + x) (-2 + 3 x)
```

```
Clear[x, x1, h]
```

```
Fx = InterpolatingPolynomial[{{x1, y1}, {x1 + h, y2}, {x1 + 2 h, y3}}, x]
```

```
Integrate[Fx, {x, x1, x1 + 2 h}]
```

Screenshot

$$y_1 + (x - x_1) \left(\frac{-y_1 + y_2}{h} + \frac{(-h + x - x_1) \left(-\frac{-y_1 + y_2}{h} + \frac{-y_2 + y_3}{h} \right)}{2h} \right)$$

$$\frac{h y_1}{3} + \frac{4 h y_2}{3} + \frac{h y_3}{3}$$

Simpson's Rule

Simpson's Rule was actually invented by Sir Isaac Newton but Thomas Simpson published it again in 1743.

It is more accurate than the Trapezium Rule because it fits a number of parabolas to the curve rather than straight lines.

To use Simpson's Rule, you must have an even number of intervals (or an odd number of ordinates).

Simpson's Rule :

$$\int_a^b y \, dx = \frac{h}{3} ((y_0 + y_n) + 4(y_1 + y_3 + \dots + y_{n-1}) + 2(y_2 + y_4 + \dots + y_{n-2}))$$

```
Clear[x, x1, h]
```

```
Fx = InterpolatingPolynomial[
```

```
  {{x1, y1}, {x1 + h, y2}, {x1 + 2 h, y3}, {x1 + 3 h, y4}}, x]
```

```
Integrate[Fx, {x, x1, x1 + 3 h}]
```

$$y_1 + (x - x_1) \left(\frac{-y_1 + y_2}{h} + (-h + x - x_1) \left(\frac{-\frac{-y_1 + y_2}{h} + \frac{-y_2 + y_3}{h}}{2h} + \right. \right. \\ \left. \left. \frac{1}{3h} (-2h + x - x_1) \left(-\frac{-\frac{-y_1 + y_2}{h} + \frac{-y_2 + y_3}{h}}{2h} + \frac{-\frac{-y_2 + y_3}{h} + \frac{-y_3 + y_4}{h}}{2h} \right) \right) \right)$$

$$\frac{3h y_1}{8} + \frac{9h y_2}{8} + \frac{9h y_3}{8} + \frac{3h y_4}{8}$$

Matrix Operations

The Wolfram Language's matrix operations handle both numeric and symbolic matrices, automatically accessing large numbers of highly efficient algorithms. The Wolfram Language uses state-of-the-art algorithms to work with both dense and sparse matrices, and incorporates a number of powerful original algorithms, especially for high-precision and symbolic matrices.

+, *****, **^**, ... — all automatically work element-wise

Dot (.) — products of matrices, automatically handling row and column vectors

Inverse — matrix inverse (use **LinearSolve** for linear systems)

Transpose — transpose (m^T , entered with `esc tr esc`)

ConjugateTranspose — conjugate transpose (m^\dagger , entered with `esc ct esc`)

Tr — trace

Det — determinant

Permanent — permanent

KroneckerProduct — matrix direct product (outer product)

MatrixPower — powers of numeric or symbolic matrices

MatrixExp — matrix exponential

MatrixLog — matrix logarithm

MatrixFunction — general matrix function

Eigenvalues, **Eigenvectors** — exact or approximate eigenvalues and eigenvectors

Eigensystem — eigenvalues and eigenvectors together

CharacteristicPolynomial — symbolic characteristic polynomial

Constructing Matrices

Mathematica offers several ways for constructing matrices:

`Table[f,{i,m},{j,n}]`

`Array[f,{m,n}]`

`ConstantArray[a,{m,n}]`

`DiagonalMatrix[list]`

`IdentityMatrix[n]`

`Normal[SparseArray[{{i1,j1}->v1,{i2,j2}->v2,...},{m,n}]]`

`RandomReal[1, {m,n}]`

IdentityMatrix[3]//MatrixForm

Out[3]//MatrixForm=

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

IdentityMatrix[3]//TraditionalForm

Out[4]//MatrixForm=

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

`A = {{1, 2, 3}, {-1, 3, 0}}`

`A // MatrixForm`

`B = {{1, 2.0, 3}, {-1, 3, 0}, {1, 0, -3}}`

`B // MatrixForm`

`Eigenvalues[B]`

`{{1, 2, 3}, {-1, 3, 0}}`

atrixForm=

$$\begin{pmatrix} 1 & 2 & 3 \\ -1 & 3 & 0 \end{pmatrix}$$

`{{1, 2., 3}, {-1, 3, 0}, {1, 0, -3}}`

atrixForm=

$$\begin{pmatrix} 1 & 2. & 3 \\ -1 & 3 & 0 \\ 1 & 0 & -3 \end{pmatrix}$$

`{-3.61061, 2.3053 + 1.15441 i, 2.3053 - 1.15441 i}`

```
B // MatrixForm
Eigenvalues[B]
Eigenvectors[B]
Eigensystem[B]
```

```
MatrixForm=
```

```
( 1  2.  3 )
(-1  3  0 )
( 1  0 -3 )
```

1. 掌握Eigenvalues, Eigenvectors, Eigensystem
2. 可以计算任意方阵

```
{-3.61061, 2.3053 + 1.15441 i, 2.3053 - 1.15441 i}
```

```
{{0.519524, 0.0785895, -0.850834},
```

```
{0.79435 + 0. i, 0.303996 + 0.505163 i, 0.142959 - 0.0311072 i},
```

```
{0.79435 + 0. i, 0.303996 - 0.505163 i, 0.142959 + 0.0311072 i}}
```

```
{{{-3.61061, 2.3053 + 1.15441 i, 2.3053 - 1.15441 i},
```

```
{{0.519524, 0.0785895, -0.850834},
```

```
{0.79435 + 0. i, 0.303996 + 0.505163 i, 0.142959 - 0.0311072 i},
```

```
{0.79435 + 0. i, 0.303996 - 0.505163 i, 0.142959 + 0.0311072 i}}}
```



```
A // MatrixForm
Eigenvalues[A]
Eigenvectors[A]
Eigensystem[A]
```

对于非方阵，会出错。

```
MatrixForm=
```

```
( 1 2 3 )
(-1 3 0)
```

```
Eigenvalues::matsq :
```

```
Argument {{1, 2, 3}, {-1, 3, 0}} at position 1 is not a non-empty square matrix. >>
```

```
Eigenvalues[{{1, 2, 3}, {-1, 3, 0}}]
```

```
Eigenvectors::matsq :
```

```
Argument {{1, 2, 3}, {-1, 3, 0}} at position 1 is not a non-empty square matrix. >>
```

```
Eigenvectors[{{1, 2, 3}, {-1, 3, 0}}]
```

```
Eigensystem::matsq :
```

```
Argument {{1, 2, 3}, {-1, 3, 0}} at position 1 is not a non-empty square matrix. >>
```

```
Eigensystem[{{1, 2, 3}, {-1, 3, 0}}]
```

掌握矩阵的QR分解。

```
In[138]:= A // MatrixForm
RES = QRDecomposition[A]
Q = RES[[1]]
R = RES[[2]]
Q // MatrixForm
R // MatrixForm
```

```
Out[138]//MatrixForm=
```

$$\begin{pmatrix} 1 & 2 & 3 \\ -1 & 3 & 0 \end{pmatrix}$$

$$\text{Out[139]= } \left\{ \left\{ \left\{ \frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}} \right\}, \left\{ \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right\} \right\}, \left\{ \left\{ \sqrt{2}, -\frac{3}{\sqrt{2}} + \sqrt{2}, \frac{3}{\sqrt{2}} \right\}, \left\{ 0, \frac{5}{\sqrt{2}}, \frac{3}{\sqrt{2}} \right\} \right\} \right\}$$

$$\text{Out[140]= } \left\{ \left\{ \frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}} \right\}, \left\{ \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right\} \right\}$$

$$\text{Out[141]= } \left\{ \left\{ \sqrt{2}, -\frac{3}{\sqrt{2}} + \sqrt{2}, \frac{3}{\sqrt{2}} \right\}, \left\{ 0, \frac{5}{\sqrt{2}}, \frac{3}{\sqrt{2}} \right\} \right\}$$

```
Out[142]//MatrixForm=
```

$$\begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix}$$

```
Out[143]//MatrixForm=
```

$$\begin{pmatrix} \sqrt{2} & -\frac{3}{\sqrt{2}} + \sqrt{2} & \frac{3}{\sqrt{2}} \\ 0 & \frac{5}{\sqrt{2}} & \frac{3}{\sqrt{2}} \end{pmatrix}$$

```
In[189]:= F = {{5, 2, 0}, {2, 2, 3}, {0, 3, 9}}
F // MatrixForm
L = CholeskyDecomposition[F]
L // MatrixForm
ConjugateTranspose[L].L - F
```

```
Out[189]= {{5, 2, 0}, {2, 2, 3}, {0, 3, 9}}
```

```
Out[190]//MatrixForm=
```

$$\begin{pmatrix} 5 & 2 & 0 \\ 2 & 2 & 3 \\ 0 & 3 & 9 \end{pmatrix}$$

```
Out[191]= {{\sqrt{5}, \frac{2}{\sqrt{5}}, 0}, {0, \sqrt{\frac{6}{5}}, \sqrt{\frac{15}{2}}}, {0, 0, \sqrt{\frac{3}{2}}}}
```

```
Out[192]//MatrixForm=
```

$$\begin{pmatrix} \sqrt{5} & \frac{2}{\sqrt{5}} & 0 \\ 0 & \sqrt{\frac{6}{5}} & \sqrt{\frac{15}{2}} \\ 0 & 0 & \sqrt{\frac{3}{2}} \end{pmatrix}$$

```
Out[193]= {{0, 0, 0}, {0, 0, 0}, {0, 0, 0}}
```

Cholesky 【乔列斯基】

分解是把一个对称正定的矩阵表示成一个下三角矩阵L和其转置的乘积的分解。它要求矩阵的所有特征值必须大于零，故分解的下三角的对角元也是大于零的。Cholesky分解法又称平方根法，是当A为实对称正定矩阵时，LU三角分解法的变形。

$$A = LL^T$$

```
In[240]:= T = {{1, 2}, {3, 4}};
          {lu, p, c} = LUdecomposition[T]
          lu // MatrixForm
          l = lu * SparseArray[{{i_, j_} /; j < i -> 1, {2, 2}}] + IdentityMatrix[2]
          u = lu * SparseArray[{{i_, j_} /; j >= i -> 1, {2, 2}}]
          (l.u - T) // MatrixForm
```

```
Out[241]= {{{{1, 2}, {3, -2}}, {1, 2}}, 1}
```

```
Out[242]//MatrixForm=
```

$$\begin{pmatrix} 1 & 2 \\ 3 & -2 \end{pmatrix}$$

```
Out[243]= {{{1, 0}, {3, 1}}
```

```
Out[244]= {{{1, 2}, {0, -2}}}
```

```
Out[245]//MatrixForm=
```

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

LU分解稍微复杂一点点
命令在help中可以找到

SingularValueDecomposition

```
In[277]:= A = {{1, 2, 3}, {-1, 3, 0.0}};  
A // MatrixForm  
{u, w, v} = SingularValueDecomposition[A]  
u.w.Transpose[v] // MatrixForm
```

```
ut[278]//MatrixForm=
```

$$\begin{pmatrix} 1 & 2 & 3 \\ -1 & 3 & 0. \end{pmatrix}$$

```
Out[279]= {{{{-0.828067, 0.560629}, {-0.560629, -0.828067}},  
{{4.16955, 0., 0.}, {0., 2.57193, 0.}}, {{{-0.0641408, 0.539943, -0.839254},  
{-0.800571, -0.52993, -0.279751}}, {-0.595796, 0.653939, 0.466252}}}}
```

```
ut[280]//MatrixForm=
```

$$\begin{pmatrix} 1. & 2. & 3. \\ -1. & 3. & 0. \end{pmatrix}$$