

# MATLAB

basic mathematical applications

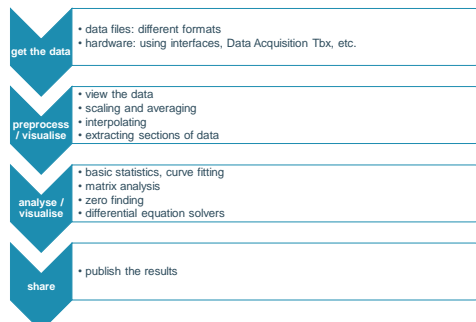
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## Topics

- basic data analysis: summary statistics
- interpolation
- curve fitting
- polynomials
- solving linear equations
- function optimisation

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## Workflow data analysis



Based on Steve Lantz: Data Analysis with MATLAB, <http://www.cac.comell.edu/education/training>

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## Basic data analysis functions

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## Summary Statistics functions

• `help datafun`

- Minimum of in a Data Set `>> min(v)`
- Maximum of in a Data Set `>> max(v)`
- Sum of a Data Set `>> sum(v)`
- Standard Deviation `>> std(v)`
- Mean of a Data Set `>> mean(v)`
- Sort a Data Set in ascending order `>> sort(v)`

remark:

MATLAB considers data sets stored in column-oriented arrays

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## Summary Statistics functions

- Cumulative functions
  - Cumulative sum `>> cumsum`
  - Cumulative product `>> cumprod`
- Integration
  - Trapezoidal numerical integration `>> trapz`
  - Cumulative trapezoidal numerical integration `>> cumtrapz`

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## mean

```
>> tempcity =
12 18
10 15
12 15
14 22
11 19
15 15
18 20
19 18
14 10
11 10
11 23
9 17
15 18
10 18
12 19
10 19
12 20
13 18
14 20
10 22
12 22
13 22
14 22
12 22
12 22
```

```
>> mean(tempcity, 1)
ans =
11.9577 8.2286 19.8710
>> mean(tempcity, 2)
ans =
12.6667
15.3333
12.0000
15.0000
13.3333
13.0000
13.0000
12.6667
14.6667
12.3333
14.3333
12.0000
13.0000
11.6667
13.6667
12.3333
11.3333
13.6667
12.0000
13.6667
13.3333
12.0000
14.3333
13.0000
```

- vectors: `mean(X)` is the mean value of the elements in X.
- matrices, `mean(X)` is a row vector containing the mean value of each column.
- `mean(X, DIM)` takes the mean along the dimension DIM of X.

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## median

- `median(x)` same as `mean(x)`, only returns the median value.

```
>> X = [0 1 2; 3 4 5]
X =
     0     1     2
     3     4     5
>> median(X, 1)
ans =
     1.5000     2.5000     3.5000
>> median(X, 2)
ans =
     1
     4
```

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## std

Standard deviation

- There are two common textbook definitions for the standard deviation  $s$  of a data vector  $X$ .

- Def\_1

$$s = \left( \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 \right)^{\frac{1}{2}}$$

- Def\_2

$$s = \left( \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2 \right)^{\frac{1}{2}}$$

- `s = std(X)`  
X is a vector, returns the standard deviation using (1) above.
- `s = std(X, flag)`  
flag = 0, is the same as `std(X)`.  
flag = 1, `std(X,1)` returns the standard deviation using (2)
- `s = std(X, flag, dim)` computes the standard deviations along the dimension of X specified by scalar dim.

```
>> dat =
    12   8   18
    15   9   22
    12   19
    14   8   23
    15   9   15
    6   10   20
    8   7   18
    19   7   18
    12   7   18
    11   8   17
    9   9   23
    8   8   19
    15   8   18
    10   8   20
    12   7   22
    9   8   19
    12   8   21
    10   9   17
    13   12   17
    9   10   20
    10   9   17
    14   21
    12   5   22
    13   9   18
    15   10   23
    13   11   24
```

```
>> atdid(dat)
ans =
    2.5098    1.7660
    2.7322
    > diddid(dat,1)
ans =
    2.4490    1.7860
    2.1959
>> atdid(dat,0,2)
ans =
    5.0332
    6.5064
    7.0000
    7.5498
    8.0829
    9.2915
    3.4441
    6.4281
    6.6583
    5.5076
    4.5092
    4.5926
    4.6583
    6.3509
    5.1116
    4.6583
    5.1116
    7.6376
    4.3589
    3.918
...

```

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max (min)

- `max` Largest component.
- `vectors: MAX(X)` is the largest element in `X`.
- `matrices: MAX(X)` is a row vector containing the maximum element from each column.
- `[Y, I] = max(X)` returns the indices of the maximum values in vector `I`.
- `max(X, Y)` returns an array the same size as `X` and `Y` with the largest elements taken from `X` or `Y`. Either one can be a scalar.

```
>> dat =
  12      8      18
  12      5      19
  12      5      22
  12      6      22
  15      9      15
  15      9      15
  18     10     20
  18     10     20
  14     10     18
  11      8     17
  9       9     23
  8       8     18
  10      9     20
  12      7     17
  12      7     22
  8       8     15
  12     10     18
  10     10     17
  13     12     18
  9     10     20
  10     10     22
  14      7     21
  12     10     22
  15     10     18
  13     11     24
  12     12     22

>> max(dat)
ans =
    19    12    24

>> max(max(dat))
ans =
    24

>> max(dat(:))
ans =
    24

>> [Y, I] = max(dat)
Y =
    19    12    24
I =
     9     23     39
```



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## diff

- `Y = diff(X)` calculates differences between adjacent elements of `X`.
- `vector`: returns a vector, 1 element shorter than `X`, of differences between adjacent elements:  $[X(2)-X(1) \ X(3)-X(2) \ \dots \ X(n)-X(n-1)]$
- `matrix`, then `diff(X)` returns a matrix of row differences:  $[X(2:m,:)-X(1:m-1,:)]$
- `Y = diff(X,n)` applies `diff` recursively `n` times, `diff(X,2)` is same as `diff(diff(X))`.
- `Y = diff(X,n,dim)` is the `n`th difference function calculated along the dimension specified by scalar `dim`.

[illegible]

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filter

filters a data sequence using a digital filter

- `y = filter(b,a,X)` filters the data in vector X with the filter described by numerator coefficient vector b and denominator coefficient vector a.  
If X is a matrix, filter operates on the columns of X.
- Algorithm:  

$$a(1)y(n) = b(1)*x(n) + b(2)*x(n-1) + \dots + b(nb+1)*x(n-nb) - a(2)*y(n-1) - \dots - a(na+1)*y(n-na)$$
- more detail in `help filter`  
[http://www.mathworks.nl/help/techdoc/data\\_analysis/bqmq3l7m-1.html](http://www.mathworks.nl/help/techdoc/data_analysis/bqmq3l7m-1.html)
- File: `demo_filter_movingaverage`



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## filter

**Table 1. Filtering**

Year	Filter	Time Series	Filtered Values
1		12	
2	25 x	17	14.00
3	50 x	10	14.75
4	25 x	22	17.25
5		15	15.75
6		11	13.75
7		18	18.50
8		27	21.50
9		14	

```
>> time_series = [12 17 10 22 15 11 18 27 14]
time_series =
    12    17    10    22    15    11    18    27    14
>> a = [.25 .50 .25]
a =
    0.2500    0.5000    0.2500
>> filter(a,1,time_series)
ans =
Columns 1 through 8
    3.0000    10.2500    14.0000    14.7500    17.2500    15.7500
   13.7500    18.5000
Column 9
   21.5000
```

## more functions

- `corrcoef`: correlation coefficients
- `cov`: covariance matrix
- `histc(x, edges)`: histogram count and bin locations using bins marked by edges
- `sort`: sorts in ascending or descending order
- `sortrows`: sort rows in ascending order (`demo_sort_sortrow.m`)

## Hands-on: Data Preprocessing

- Missing values:
  - remove NaNs from the data before performing computations.
  - `demo_check_NaN`
- Removing outliers:
  - remove outliers or misplaced data points from a data set
    - Calculate the mean and standard deviation from the data set.
    - Get the column of points that lies outside the 3\*std. (3σ-rule)
    - Remove these points
    - Check `isoutlier`
  - `demo_check_outlier.m`
- More examples:
  - `demo_basic_DataAnalysis`
  - `data_treatment_cambridge`

## interpolation

## Fitting/interpolation

Tips:

- Young/Mohlenkamp
  - <http://www.ohiouniversityfaculty.com/youngt/IntNumMeth/>
- Check Cleve Moler's website
  - <https://nl.mathworks.com/moler.html>
- Help
  - Search in helpdesk data analysis
  - Interpolation
  - fitting

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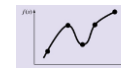
## Interpolation

- A way of estimating values of a function between those given by some set of data points
  - When you take data, how do you predict what other data points might be?
  - Two techniques are :

- Linear Interpolation



- Cubic Spline Interpolation



- Extrapolation: be careful, make sure extrapolation makes sense with your data

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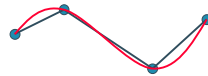
## Interpolation

`.YI = interp1(X,Y,XI)` interpolates to find YI, the values of the underlying function Y at the points in the vector or array XI

- Finding values between data points
  - linear interpolation (default)
  - cubic splines

```
yi = interp1 ( x, y, xi )
Yi = interp1 ( x, y, xi, 'spline' )
```

- `demo_interp1`
- `demo_interp2`
- `demo_interp3`
- `demo_interp4`



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## Fitting

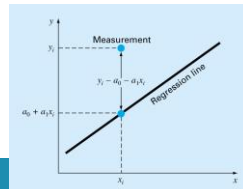
- Idea: modeling data with an equation
- We can estimate what equation represents the data by “eyeballing” a graph
  - there will be points that do not fall on the line we estimate
  - There is scatter in collected data
- Linear and cubic spline interpolations fit curves constrained to go through the data points
- A curve fitted using least squares may not pass through any data point but it will be “close” to all of them in a “least squares” sense
- Find a function, e.g. a polynomial of whatever order, that minimizes the mean square error

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## Linear regression

- Linear equation that is the best fit to a set of data points
- Minimize the sum of squared distances between the line and the data points
- Use `polyfit`
- `demo_linreg1.m`



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## Polynomial Fit

- A pair of vectors (x, y), representing the independent and dependent variables of a (possibly noisy) relationship.
- Get the curve that most closely fits the data.
- `polyfit` finds the coefficients of a polynomial representing the data
- generates a "best fit" polynomial (in the least squares sense) of a specified order for a given set of data.
- used to generate the n + 1 coefficients a<sub>j</sub> of the nth-degree polynomial used for fitting the data.

$$p_n(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$$

`as = polyfit(x, y, n)`

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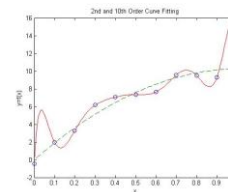
## Polynomial Fit

- `polyval`: use the coefficients to find new values of y, that correspond to the known values of x
- Create a new set of y points for the approximate curve with
- `yapprox = polyval(as, x)`

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## polyfit



`demo_polyfit`

```
>> x = [0 .1 .2 .3 .4 .5 .6 .7 .8 .9
1];
>> y_c = [-.447 1.978 3.28 5.16 7.08 7.34
7.66 9.56 9.48 9.30 11.2];
>> n = 2;
>> p = polyfit(x, y, n);
>> p =
-9.8108 20.1293 -0.0317
xi = linspace(0,1,100);
yi = polyval(p, xi);
pp = polyfit(x, y, 10);
y10 = polyval(pp, xi);
plot(x, y, 'o', xi, yi, '--', xi, y10) % plot
data
xlabel('x'), ylabel('y=f(x)')
title('2nd and 10th Order Curve
Fitting')
```

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## polynomials

- Polynomials are described by using a row vector of the coefficients of the polynomial beginning with the highest power of  $x$  and inserting zeros for "missing" terms:

```
f=[9 -5 3 7];
```

```
g=[6 -1 2];
```

- add and subtract polynomial functions in MATLAB. To do this we must "zero" pad the polynomials so that their row vector representations are the same length:

```
h=f+[0 g];
```

- multiply and divide polynomials by using the `conv` and `deconv` functions

```
y=conv(f,g)
```

```
[q r]=deconv(y,f)
```

- evaluate a polynomial at any value of  $x$ :

```
p=[1 3 -4];
```

```
x=[-5:.1:5];
```

```
px=polyval(p,x);
```

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## polynomials

- `roots` finds polynomial roots.

```
roots([1 6 11 6])
```

```
ans = -3.0000
```

```
-2.0000
```

```
-1.0000
```

- `polyder(P)` returns the derivative of the polynomial whose coefficients are the elements of vector  $P$ .

```
polyder([1 6 11 6])
```

```
ans = [3 12 11]
```

- `polyint(P)` returns the integral of the polynomial whose coefficients are the elements of vector  $P$ .

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## Linear equations

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## Linear equations

Tips:

- Check Cleve Moler's website
  - [www.mathworks.com/moler](http://www.mathworks.com/moler)
- Help
  - Systems of linear equations
  - Backslash operator

Basic Data  
Analysis\_3

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## Simultaneous Linear Equations

- Basic form:

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \dots + a_{1n}x_n &= b_1 \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \dots + a_{2n}x_n &= b_2 \\ \dots\dots\dots \\ a_{m1}x_1 + a_{m2}x_2 + a_{m3}x_3 + \dots + a_{mn}x_n &= b_m \end{aligned}$$

- Where:

- $a_{ij}$  are known coefficients
- $x_i$  are unknowns
- $b_i$  are known right hand side values

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

– or –

$$Ax = b$$

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## Simultaneous Linear Equations

- The coefficient matrix A need not be square.  
If A is  $m \times n$ , there are three cases.

- $m = n$ 
  - Square system.
  - Seek an exact solution.
- $m > n$ 
  - Overdetermined system.
  - Find a least squares solution.
- $m < n$ 
  - Underdetermined system.
  - Find a basic solution with at most  $m$  nonzero components.
- Check *rank* of a matrix (number of independent rows or columns). : rank

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## Linear Equations

- Set of linear equations  $Ax = b$  can be solved using the Inverse operation on A  
 $x = \text{inv}(A) * b$
- Advice:
  - Do not use this method
  - Inefficient

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## backslash \

- MATLAB has a bunch of methods available to solve a system of linear equations
- If you desire the solution of  $Ax = b$ , then the simplest method using Matlab to find  $x$  is to set  $x = A \backslash b$ 

```
A = [ 1 5 6; 7 9 6; 2 3 4]
b = [29; 43; 20]
x=A\b
```
- Use backslash operator \ (LU decomposition + pivoting)
- For non-square and singular systems, the operation  $A \backslash b$  gives the solution in the least squares sense. No error message
- The " $\backslash$ " backslash operator uses a combination of numerical methods including LU decomposition.
- doc mldivide

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## Steps in $A \setminus b$

Stop if successful

- If  $A$  is upper or lower triangular, solve by back/forward substitution
- If  $A$  is permutation of triangular matrix, solve by permuted back substitution (useful for  $[L,U]=\text{lu}(A)$  since  $L$  is permuted)
- If  $A$  is symmetric/hermitian
  - Check if all diagonal elements are positive
  - Try Cholesky, if successful solve by back substitutions
- If  $A$  is Hessenberg (upper triangular plus one subdiagonal), reduce to upper triangular then solve by back substitution
- If  $A$  is square, factorize  $PA = LU$  and solve by back substitutions
- If  $A$  is not square, run Householder QR, solve least squares problem
  - `demo_lsqeqN`
  - `check`
  - `scicomp.stackexchange.com/questions/1001/how-does-the-matlab-backslash-operator-solve-ax-b-for-square-matrices`
  - `http://www.mathworks.nl/support/solutions/en/data/1-172BD/index.html?product=ML&solution=1-172BD`

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More

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## Basic optimization

- Optimization deals with finding the maxima and minima of a function that depends on one or more variables.
- Basic functions available in Matlab + Optimization toolbox
- Minimization in 1 dimension
  - $X = \text{fminbnd}(\text{FUN}, x1, x2)$  attempts to find a local minimizer  $X$  of the function  $\text{FUN}$  in the interval  $x1 < X < x2$ .  $\text{FUN}$  accepts scalar input  $X$  and returns a scalar function value  $F$  evaluated at  $X$ .
    - `demo_fminbnd`
    - `demo_fminbnd_2`
    - `demo_fminbnd_3`
- Minimization in  $N$  dimensions
  - $X = \text{fminsearch}(\text{FUN}, X0)$  starts at  $X0$  and attempts to find a local minimizer  $X$  of the function  $\text{FUN}$ .  $\text{FUN}$  accepts input  $X$  and returns a scalar function value  $F$  evaluated at  $X$ .  $X0$  can be a scalar, vector or matrix.
  - Multidimensional unconstrained nonlinear minimization (Nelder-Mead).

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## Zero finding

- `fzero`: Find zero of a function of one variable  
find  $x$  satisfying  $f(x) = 0$
- An implementation of T. Dekker's algorithm
- Combination of
  - Secant method / quadratic fit extension of Secant method
  - Bisection
- While not universally effective, combining techniques can improve the reliability and speed of convergence
- `fzero` uses bisection to avoid large steps, other methods for fast convergence
- syntax:
  - $x = \text{fzero}(\text{fun}, x0)$ 
    - **fun** user supplied function specifying  $f(x)$
    - **x0** is an initial guess of an  $x$

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## Even More

- Solving ODE
- Eigenvalues
- Solving PDE
- Sparse systems
- ...