

# MATLAB

Matlab by Example

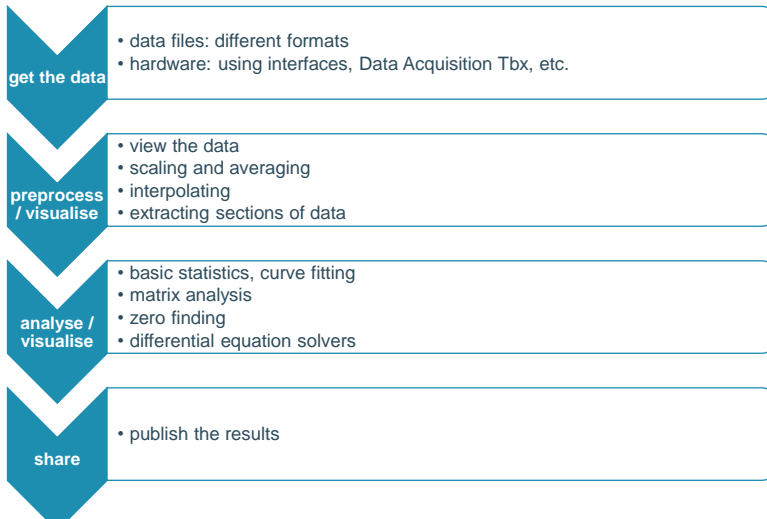
## Topics



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



# Typical Workflow



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

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## Basic data analysis functions

# Data Analysis

- See 5 Types of Data and How to Analyze them with MATLAB  
<https://nl.mathworks.com/company/newsletters/articles/5-types-of-data-and-how-to-analyze-them-with-matlab.html>



## 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



# Summary Statistics functions

- Cumulative functions

- Cumulative sum

```
>> cumsum
```

- Cumulative product

```
>> cumprod
```

- Integration

- Trapezoidal numerical integration

```
>> trapz
```

- Cumulative trapezoidal numerical integration

```
>> cumtrapz
```



## mean

```
>> temp3city =  
12 8 18  
15 9 22  
12 5 19  
14 8 23  
12 6 22  
11 9 19  
15 9 15  
8 10 20  
19 7 18  
12 7 18  
14 10 19  
11 8 17  
9 7 23  
8 8 19  
15 8 18  
8 9 20  
10 7 17  
12 7 22  
9 8 19  
12 8 21  
12 8 20  
10 9 17  
13 12 18  
9 10 20  
10 6 22  
14 7 21  
12 5 22  
13 7 18  
15 10 23  
13 11 24  
12 12 22
```

```
>> mean(temp3city, 1)  
ans =  
11.9677 8.2258 19.8710  
>> mean(temp3city, 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`.



# 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
```



# 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     5    19
    14     8    23
    12     6    22
    11     9    19
    15     9    15
     8    10    20
    19     7    18
    12     7    18
    14    10    19
    11     8    17
     9     7    23
     8     8    19
    15     8    18
     8     9    20
    10     7    17
    12     7    22
     9     8    19
    12     8    21
    12     8    20
    10     9    17
    13    12    18
     9    10    20
    10     6    22
    14     7    21
    12     5    22
    13     7    18
    15    10    23
    13    11    24
    12    12    22

>> std(dat)
ans =
     2.5098     1.7646
     2.2322
>> std(dat,1)
ans =
     2.4690     1.7360
     2.1959
>> std(dat,0,2)
ans =
     5.0332
     6.5064
     7.0000
     7.5498
     8.0829
     5.2915
     3.4641
     6.4291
     6.6583
     5.5076
     4.5092
     4.5826
     8.7178
     6.3509
     5.1316
     6.6583
     5.1316
     7.6376
     4.3589
     ...
```



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

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

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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`

File: `demo_filter_movingaverage`



## 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`)



## Example: Data Preprocessing

- Missing values:
  - remove NaNs from the data before performing computations.
  - File: `demo_check_NaN.m`
- 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  $3\sigma$ -rule
    - Remove these points
    - Check `isoutlier`
  - File: `demo_check_outlier.m`
- More examples:
  - File: `demo_basic_DataAnalysis`
  - File: `data_treatment_cambridge.mlx`



# Interpolation / Curve fitting

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## 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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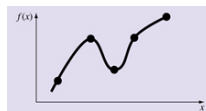
# Interpolation

- A way of estimating values of a function between those given by some set of known 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



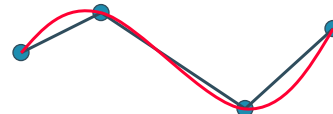
# Interpolation

`.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
- The default is linear interpolation, but there are other types available (`doc interp1`)

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

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





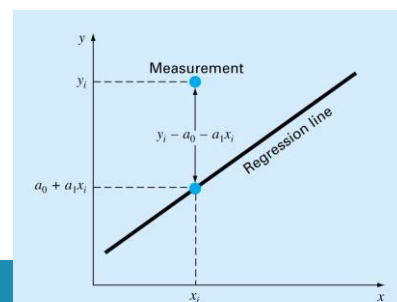
# Fitting

- 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



# Linear regression

- Linear equation that is the best fit to a set of data points
- $y(x) = a_1x + a_0$
- Minimize the sum of squared distances between the line and the data points
- Use `polyfit`
- File: `demo_linreg1.m`





# 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  $a_j$  of the  $n$ th-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)
```



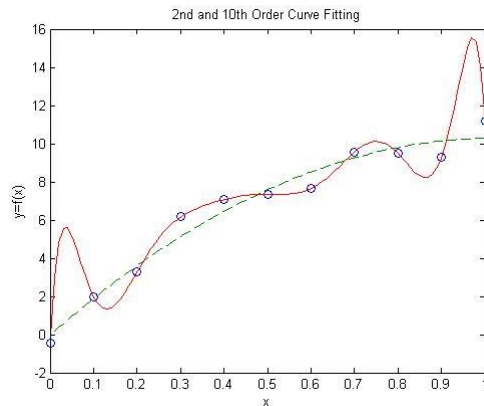
# Polynomial Fit

- `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)`



# polyfit

File: demo\_polyfit



```
>> x = [0 .1 .2 .3 .4 .5 .6 .7 .8 .9 1];
>> y = [-.447 1.978 3.28 6.16 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`.

## Linear equations



# Linear equations

Tips:

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



# Simultaneous Linear Equations

- Basic form:

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \dots + a_{2n}x_n = b_2$$

.....

$$a_{m1}x_1 + a_{m2}x_2 + a_{m3}x_3 + \dots + a_{mn}x_n = b_m$$

- Where:

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

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

– or –

$$Ax = b$$



# 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



# 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





## 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  $\backslash$  (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`



## Steps in $A \backslash b$

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  $[L,U]=lu(A)$  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_lineqN`
  - `check`
  - [scicomp.stackexchange.com/questions/1001/how-does-the-matlab-backslash-operator-solve-ax-b-for-square-matrices](http://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>

## Example: solving a mass balance

- File: mass\_balance\_plant.mlx
- Source: Solving Mass Balances using Matrix Algebra  
<https://sagmilling.com/articles/4/view/MassBalance-MatrixMethod.pdf?s=1>

## Basic optimization



# 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 = fminbnd(FUN,x1,x2)` attempts to find a local minimizer `X` of the function `FUN` in the interval  $x1 < X < x2$ . `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 = fminsearch(FUN,X0)` starts at `X0` and attempts to find a local minimizer `X` of the function `FUN`. `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).



# 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 = fzero(fun,x0)`
  - **`fun`** user supplied function specifying  $f(x)$
  - **`x0`** is an initial guess of an `x`

# ODE

## ode

- MATLAB has a collection of m-files, called the ODE suite to solve initial value problems of the form

$$y' = f(t, y)$$

$$y(t_0) = y_0$$

- solve for  $y(t)$  for  $t > t_0$

- Steps to use

- Express the differential equation as a set of first-order ODEs

- Write an m-file to compute the state derivative

function dydt = myprob(t, y)

- Use one of the ODE solvers to solve the equations

[t, y] = ode\_solver('myprob', tspan, y0);

MATLAB has a wide variety of ODE solvers which can vary in how they solve and what ODE's they are certified to solve.

ode45, ode23, ode113, ode15s, ode23s, ode23t, ode23tb

## General format

`[T,Y] = solver(odefun,tspan,y0) :`

1. “Solver” is usually “ODE45” which is utilizing RK4. Or “ODE15s”
2. odefun is a function that evaluates the right-hand side of the differential equations  $y'=f(x,y)$ .
3. tspan is a vector specifying the interval of integration,  $[t_0, t_f]$ . To obtain solutions at specific times, use `tspan = [t0,t1,...,tf]`.
4.  $Y_0$  is a vector of initial conditions

`driver_fo_ode.m`

`ode_ex1.m`

## Example: predator – prey model

- <https://nl.mathworks.com/help/matlab/math/numerical-integration-of-differential-equations.html>
- <https://www.mathworks.com/content/dam/mathworks/mathworks-dot-com/moler/exm/chapters/predprey.pdf> (Moler text)



## Even More

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