



Misc stuff

**MATLAB**

# Saving and reading your workspace

```
>> eig_mov
```

```
>> whos
```

Name	Size	Bytes	Class	Attributes
D	2x2	32	double	
V	2x2	32	double	
. . .				
cnt	1x1	8	double	
x	2x361	5776	double	

```
>> save eig_mov_ex.mat
```

## Saves workspace in file stuff.mat

```
>> clear
```

```
>> whos
```

```
>> load eig_mov_ex.mat
```

```
>> whos
```

Name	Size	Bytes	Class	Attributes
D	2x2	32	double	
V	2x2	32	double	
. . .				
cnt	1x1	8	double	
x	2x361	5776	double	

```
>>
```



# Saving what you type

```
>>diary everythingItype.txt
```

Saves everything you type

```
>>dairy
```


To turn it off





## Garbage collection

Any system such as MATLAB that maintains an environment with variables continually being created and destroyed must have a form of "garbage collection" to remove dead (or unused) space.








Unfortunately, MATLAB has no automatic  
garbage collection mechanism.



The function `clear` allows the user to manage his workspace and do his own house cleaning.

Even that is not enough, since other temporary arrays might be created and destroyed whenever M-files are run.



In place of garbage collection, there is a MATLAB function called

pack

which saves all the variables in the entire workspace, clears the workspace, and then loads the saved variables.



This is time-consuming, but it is the best way to get some room to work if memory limits start to hinder your progress.



Supressing "Current plot held", etc.  
messages

be more specific - hold ('on'), etc.





# Importance of thinking through how to program something

<http://www.joelonsoftware.com/articles/fog0000000319.html>

(so the world is not stuck with your mistake forever).





# Random numbers

How random are they?

Compare - run multiple times

Restart Matlab - run multiple times

Mac vs PC



(seed)



Linear Algebra (a la Matlab) Review

**MATLAB**



## Acknowledgement

This lecture borrows heavily from online lectures/ppt files posted by

David Jacobs at Univ. of Maryland

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Joseph Bradley at Carnegie Mellon

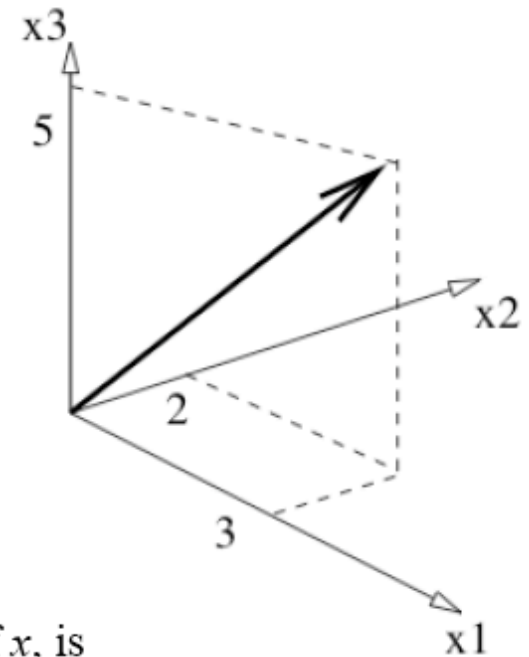




# Vectors

```
>> a=[2 3 5];  
>> norm(a)  
6.1644  
>> norm(a)^2  
38.000
```

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \quad \text{e.g.} \quad \mathbf{x} = \begin{pmatrix} 3 \\ 2 \\ 5 \end{pmatrix}$$



The **length** of  $\mathbf{x}$ , a.k.a. the **norm** or **2-norm** of  $\mathbf{x}$ , is

$$\|\mathbf{x}\| = \sqrt{x_1^2 + x_2^2 + \cdots + x_n^2}$$

e.g.,

$$\|\mathbf{x}\| = \sqrt{3^2 + 2^2 + 5^2} = \sqrt{38}$$

# Matrices

$$U = [u_{mn}] = \begin{bmatrix} u_{11} & u_{12} & \dots & u_{1n} \\ u_{21} & u_{22} & \dots & u_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ u_{m1} & u_{m2} & \dots & u_{mn} \end{bmatrix}$$

The general matrix consists of  $m$  rows and  $n$  columns. It is also known as an  $m \times n$  (read  $m$  by  $n$ ) array.

Each individual number,  $u_{ij}$ , of the array is called the *element*

Elements  $u_{ij}$  where  $m=n$  is called the principal diagonal

# Transpose of a Matrix

```
>> a=[6 1;2 5]
```

```
a =
```

```
     6     1
```

```
     2     5
```

```
>> a'
```

```
ans =
```

```
     6     2
```

```
     1     5
```

Transpose:

$$C_{m \times n} = A^T_{n \times m}$$

$$c_{ij} = a_{ji}$$

$$(A + B)^T = A^T + B^T$$

$$(AB)^T = B^T A^T$$

Examples:

$$\begin{bmatrix} 6 & 2 \\ 1 & 5 \end{bmatrix}^T = \begin{bmatrix} 6 & 1 \\ 2 & 5 \end{bmatrix}$$

$$\begin{bmatrix} 6 & 2 \\ 1 & 5 \\ 3 & 8 \end{bmatrix}^T = \begin{bmatrix} 6 & 1 & 3 \\ 2 & 5 & 8 \end{bmatrix}$$

If  $A^T = A$ , we say  $A$  is **symmetric**.

Notice how Matlab looks like math.



# Matrix & Vector Addition

```
>> a=[1; 2]
```

```
a =
```

```
1
```

```
2
```

```
>> b=[3; 4]
```

```
b =
```

```
3
```

```
4
```

```
>> c=a+b
```

```
c =
```

```
4
```

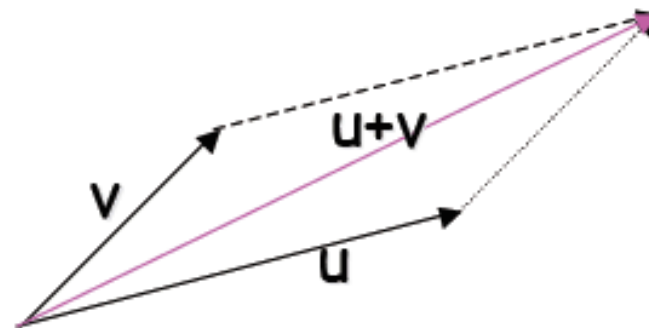
```
6
```

```
>>
```

NO LOOPS  
Looks like  
Math - just  
add them.

Vector/Matrix addition is  
associative and commutative  
 $(A+B)+C=A+(B+C)$ ;  $A+B=B+A$

$$u + v = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} + \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} u_1 + v_1 \\ u_2 + v_2 \end{bmatrix}$$



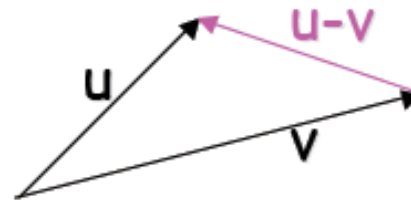
# Matrix and Vector Subtraction

Same as addition

Vector/Matrix subtraction is also  
associative and commutative

$$(A-B)-C = A-(B-C); \quad A-B=B-A$$

$$u - v = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} - \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} u_1 - v_1 \\ u_2 - v_2 \end{bmatrix}$$



# Matrix and Vector Scaling

$$z = \alpha x$$

for a scalar  $\alpha$  then

$$z = \alpha \begin{pmatrix} 3 \\ 2 \\ 5 \end{pmatrix} = \begin{pmatrix} 3\alpha \\ 5\alpha \\ 2\alpha \end{pmatrix}$$

```
>> x=[1 2 3]
```

```
x =
```

```
    1    2    3
```

```
>> y=3*x
```

```
y =
```

```
    3    6    9
```

```
>>
```



For addition and subtraction, the size of the matrices must be the same

$$A_{nm} + B_{nm} = C_{nm}$$

For scalar multiplication,  
the size of  $A_{nm}$  does not matter

All three of these operations do not differ  
from their ordinary number counterparts

The operators work element-by-element  
through the array,  $a_{mn} + b_{mn} = c_{mn}$

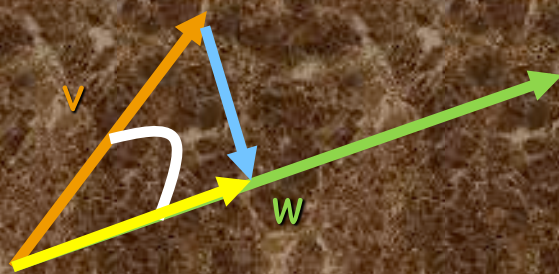


# Vector Multiplication: inner or dot product

The inner product of vector multiplication  
is a **SCALAR**

$$v \cdot w = (x_1, x_2) \cdot (y_1, y_2) = \|v\| \cdot \|w\| \cos \alpha$$

$$v \cdot w = (x_1, x_2) \cdot (y_1, y_2) = x_1 y_1 + x_2 y_2$$



Projection of one vector (orange) onto another  
(green, result - projection - yellow).

Dot product is zero for perpendicular  
vectors.

The inner/dot product can be represented as row matrix multiplied by a column matrix. A row matrix can be multiplied by a column matrix, in that order, only if they each have the same number of elements!

```
>> x=[1 2]
x =
     1     2
>> y=[2 1]
y =
     2     1
>> x*y'
ans =
     4
>> y=[-2 1]
y =
    -2     1

>> x*y'
ans =
     0
>> y=[2 -1]
y =
     2    -1
>> x*y'
ans =
     0
```

$$a = \begin{bmatrix} 6 \\ 2 \\ -3 \end{bmatrix} \quad b = \begin{bmatrix} 4 \\ 1 \\ 5 \end{bmatrix}$$

$$a \cdot b = a^T b$$

$$= [6 \quad 2 \quad -3] \begin{bmatrix} 4 \\ 1 \\ 5 \end{bmatrix}$$

$$= 6 \cdot 4 + 2 \cdot 1 + (-3) \cdot 5$$

$$= 11$$

# Several ways to properly calculate the dot product of two vectors

```
>>sum(a.*b)
```

element by element multiplication (.\*), then sum the results - based on definition.

Or making it look like matrix multiplication

```
>>a'*b
```

```
>>a*b'
```

Or using matlab function

```
>>c=dot(a,b)
```



# The outer product

A column vector multiplied by a row vector.  
The outer product of vector multiplication  
is a **MATRIX**.

```
>> a=[ 6  2 -3]
```

```
a =
```

```
     6     2    -3
```

```
>> b=[ 4;1;5]
```

```
b =
```

```
     4
```

```
     1
```

```
     5
```

```
>> c=b*a
```

```
c =
```

```
    24     8    -12
```

```
     6     2     -3
```

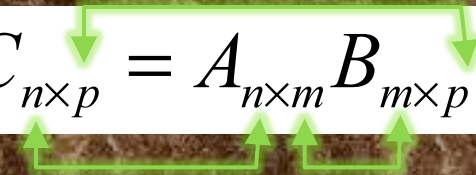
```
    30    10    -15
```

```
>>
```



# Matrix Multiplication

Two matrices can be multiplied together  
if and only if  
the number of columns in the first equals  
the number of rows in the second.

$$C_{n \times p} = A_{n \times m} B_{m \times p}$$


$$c_{ij} = \sum_{k=1}^m a_{ik} b_{kj}$$

In MATLAB, the \* symbol represents matrix multiplication :

```
>> a=[1 2 3;3 2 1]
```

```
a =
```

```
1 2 3
3 2 1
```

```
>> b=[4 5;10 2;2 10]
```

```
b =
```

```
4 5
10 2
2 10
```

```
>> c=a*b
```

```
c =
```

```
30 39
34 29
```

```
>>
```

```
>> a(1,:)⊙
```

```
ans =
```

```
1 2 3
```

```
>> b(:,1)
```

```
ans =
```

```
4
```

```
10
```

```
2
```

```
>> a(1,:)*b(:,1)
```

```
ans =
```

```
30
```

```
>> a(1,:)*b(:,2)
```

```
ans =
```

```
39
```

```
>> a(2,:)*b(:,2)
```

```
ans =
```

```
29
```

```
>> a(2,:)*b(:,1)
```

```
ans =
```

```
34
```

- Matrix multiplication is not commutative!

$$A_{n \times n} B_{n \times n} \neq B_{n \times n} A_{n \times n}$$

```
>> c=a*b
```

```
c =
```

```
30 39
```

```
34 29
```

```
>> c=b*a
```

```
c =
```

```
19 18 17
```

```
16 24 32
```

```
32 24 16
```

- Matrix multiplication is distributive and associative

$$A(B+C) = AB + AC$$

$$(AB)C = A(BC)$$



Matrices can represent sets of equations

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

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

...

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n$$

What's the matrix representation?

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \quad b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$Ax = b$$



# Determinant of a Matrix

```
>> a=magic(3)
```

```
a =
```

```
      8      1      6
      3      5      7
      4      9      2
```

```
>> det(a)
```

```
Ans
```

```
   -360
```

```
>>
```

**Determinant:**     A must be square

$$\det \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{21}a_{12}$$

**Example:**      $\det \begin{bmatrix} 2 & 5 \\ 3 & 1 \end{bmatrix} = 2 - 15 = -13$

$$\det \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

# Inverse of a Matrix

```
>> a=[1 2 3;4 5 6; 7 8 9]
a =
     1     2     3
     4     5     6
     7     8     9

>> ainv=inv(a)
ainv =
    0.1472   -0.1444    0.0639
   -0.0611    0.0222    0.1056
   -0.0194    0.1889   -0.1028

>> ainv*a
ans =
    1.0000         0   -0.0000
         0    1.0000         0
         0    0.0000    1.0000

>> a*ainv
ans =
    1.0000         0   -0.0000
   -0.0000    1.0000         0
    0.0000         0    1.0000

>>
```

If  $A$  is a square matrix, the **inverse** of  $A$ , called  $A^{-1}$ , satisfies

$$AA^{-1} = I \quad \text{and} \quad A^{-1}A = I,$$

Where  $I$ , the **identity matrix**, is a diagonal matrix with all 1's on the diagonal.

$$I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

If square matrix  
invertible, has same  
right and left inverse.

For a 2-D matrix

if

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

the principal diagonal elements switch positions

$$A^{-1} = \frac{\begin{bmatrix} d & -b \\ -c & a \end{bmatrix}}{|A|}$$

the off diagonal elements  
change sign

the determinant

Example:  $\begin{bmatrix} 6 & 2 \\ 1 & 5 \end{bmatrix}^{-1} = \frac{1}{28} \begin{bmatrix} 5 & -2 \\ -1 & 6 \end{bmatrix}$

$$\begin{bmatrix} 6 & 2 \\ 1 & 5 \end{bmatrix}^{-1} \cdot \begin{bmatrix} 6 & 2 \\ 1 & 5 \end{bmatrix} = \frac{1}{28} \begin{bmatrix} 5 & -2 \\ -1 & 6 \end{bmatrix} \cdot \begin{bmatrix} 6 & 2 \\ 1 & 5 \end{bmatrix} = \frac{1}{28} \begin{bmatrix} 28 & 0 \\ 0 & 28 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Square matrices with inverses are said to be nonsingular

Not all square matrices have an inverse.  
These are said to be singular.

Square matrices with determinants = 0 are singular. (determinant in denominator)

Rectangular matrices are always singular.



## Right- and Left- Inverse

If a matrix  $G$  exists such that  $GA = I$ , then  
 $G$  is a left-inverse of  $A$

If a matrix  $H$  exists such that  $AH = I$ , then  
 $H$  is a right-inverse of  $A$

Rectangular matrices may have right- or  
left- inverses, but they are still singular.

For

$A = m \times n, m > n$ : we have a left inverse  $(A^T A)^{-1} A^T A = I_n$ ,

$$A_{left}^{-1} = (A^T A)^{-1} A^T$$

$A = m \times n, n > m$ : we have a left inverse  $AA^T (AA^T)^{-1} = I_m$ ,

$$A_{right}^{-1} = A^T (AA^T)^{-1}$$

```
>> a=[1 2 3;4 5 6]
```

```
a =
```

```
1    2    3
```

```
4    5    6
```

```
>> ainv=a'*inv(a*a')
```

```
ainv =
```

```
-0.9444    0.4444
```

```
-0.1111    0.1111
```

```
0.7222   -0.2222
```

```
>> a*ainv
```

```
ans =
```

```
1.0000   -0.0000
```

```
-0.0000    1.0000
```

```
>> >> det(a'*a)
```

```
ans =
```

```
0
```

Right inverse exists,  
but left doesn't.

## Matrix Division in Matlab

In ordinary math, division ( $a/b$ ) can be thought of as  $a*(1/b)$  or  $a*b^{-1}$ .

There really is no such thing as matrix division in any simple sense.

A unique inverse matrix of  $b$ ,  $b^{-1}$ , only potentially exists if  $b$  is square.

Also, matrix multiplication is not commutative, unlike ordinary multiplication.

## Matrix Division in Matlab

$/$  :  $B/A$  (right division) is roughly the same as  $B*inv(A)$ .

A and B must have the same number of columns for right division.

More precisely,  $B/A = (A' \setminus B')'$ .



# Matrix Division in Matlab

`\` : If  $A$  is a square matrix,  $A \setminus B$  (left division) is roughly the same as  $\text{inv}(A)*B$ , except it is computed in a different way.

$A$  and  $B$  must have the same number of rows for left division.

```
a =  
    1    2  
    3    4
```

```
>> b  
b =  
    1  
    2
```

```
>> c=a\b
```

```
c =  
    0  
 0.5000
```

```
>> ainv=inv(a)
```

```
ainv =  
 -2.0000    1.0000  
  1.5000   -0.5000
```

```
>> ainv*b
```

```
ans =  
    0  
 0.5000
```

```
>>>> a*c
```

```
ans =  
    1  
    2
```

# Matrix Division in Matlab

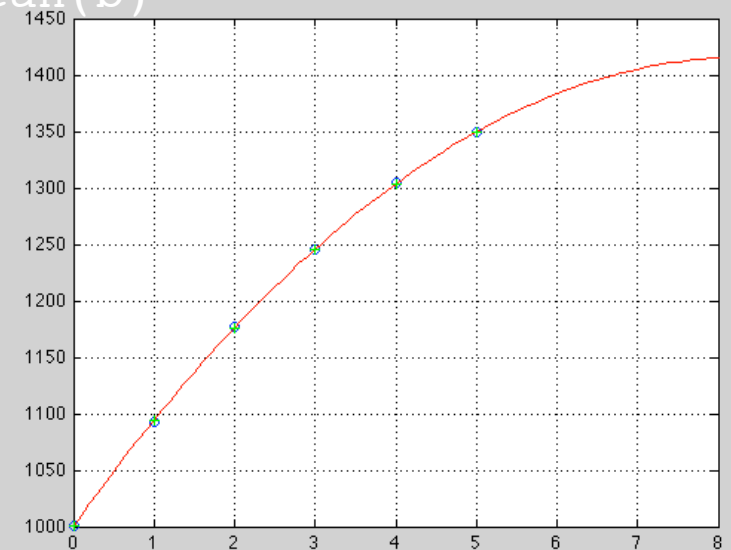
`\` : If  $A$  is an  $m$ -by- $n$  matrix (not square) and  $b$  is a matrix of  $m$  rows,  $Ax=b$  is solved by least squares using  $A \backslash b$  (left division).

```
>> A
A =
1.0000    0    0
1.0000  1.0000  0.5000
1.0000  2.0000  2.0000
1.0000  3.0000  4.5000
1.0000  4.0000  8.0000
1.0000  5.0000 12.5000

>> x_ = A \ b
x_ =
1.0e+03 *
1.0004
0.0998
-0.0120

>> (A*x_-b)/mean(b)
ans =
-0.0005
0.0011
-0.0008
0.0008
-0.0011
0.0005

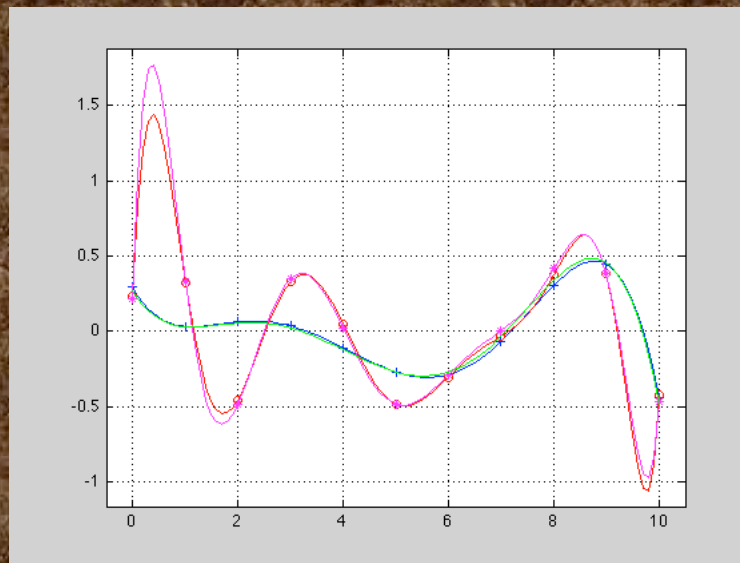
>> b
b =
1001
1093
1177
1245
1305
1349
```



Matlab also has routines to do polynomial fits (positive powers only).

Data - 11 points (red circles)

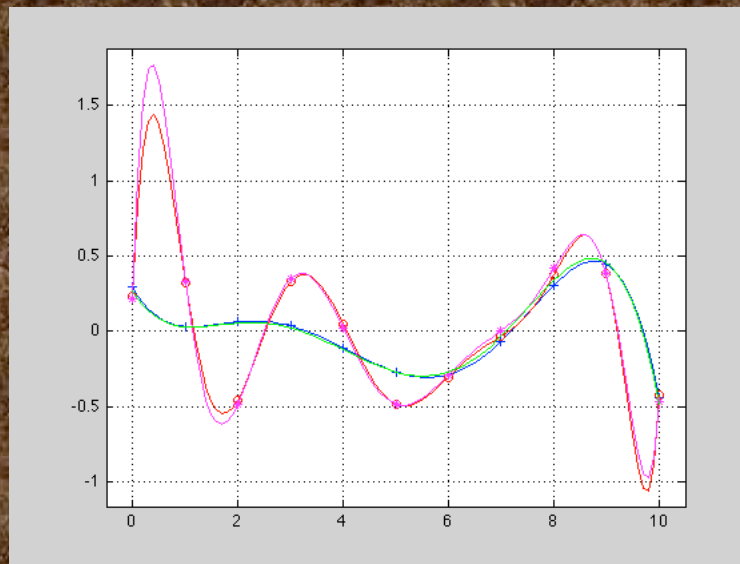
5<sup>th</sup> order polynomial fit to 11 points - fewer parameters than data - get LS fit (blue line, blue '+') - does not go through data points (but misfit "minimized").





Data - 11 points (red circles)

10<sup>th</sup> order polynomial fit to 11 points - same number parameters as data - get exact solution (red line). Goes through each point exactly.



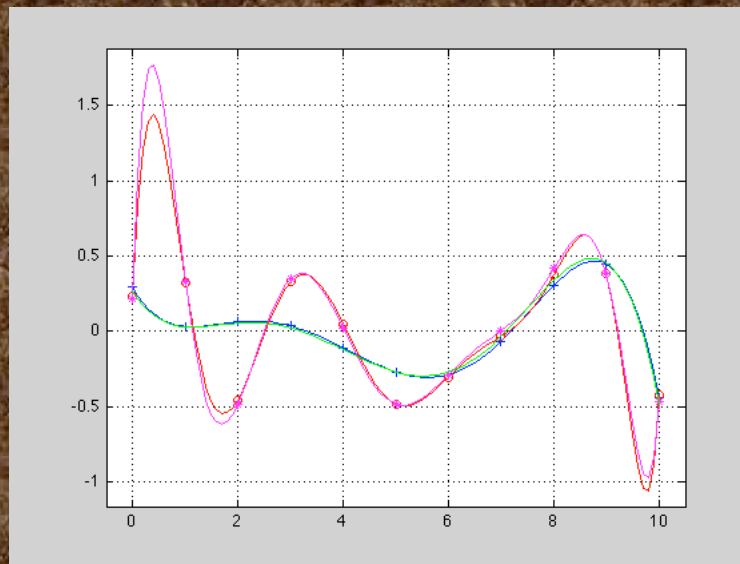


Now add some noise.

Data - 11 points (red circles)

Magenta and green - 5<sup>th</sup> and 10<sup>th</sup> order polynomials fit to data with 10% noise.

The fits to the noisy and perfect data look pretty much the same when plotted.



```
>> poly_demo
```

```
p5 =
```

```
pn5 =
```

```
p10 =
```

```
Columns 1 through 11
```

0.0000	-0.0003	0.0051	-0.0474	0.2251	-0.3195	-1.7232	8.6334	-14.0467	7.3647	0.2333
--------	---------	--------	---------	--------	---------	---------	--------	----------	--------	--------

```
pn10 = ?
```

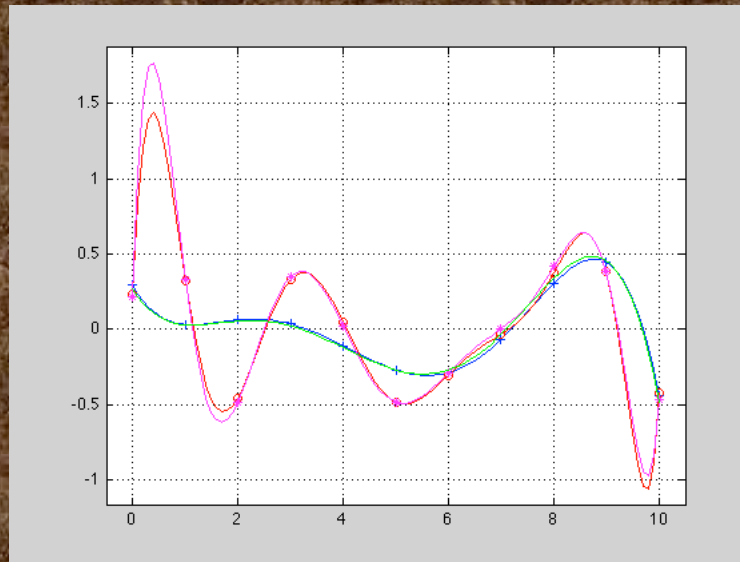
```
Columns 1 through 11
```

0.0000	-0.0002	0.0024	-0.0145	-0.0298	0.9338	-5.6002	15.8159	-21.1548	10.1628	0.2137
--------	---------	--------	---------	---------	--------	---------	---------	----------	---------	--------

-0.0010	0.0215	-0.1629	0.4980	-0.6166	0.2926
---------	--------	---------	--------	---------	--------

-0.0009	0.0209	-0.1559	0.4696	-0.5767	0.2733
---------	--------	---------	--------	---------	--------

The models (the values for polynomial coefficients), however, are quite different.  
Compare "stability" of the solutions.



# Array Operators (review)

+ Addition

- Subtraction

.\* Element-by-element multiplication

./ Element-by-element division.

(A./B: divides A by B by element)

.\ Element-by-element left division

(A.\B divides B by A by element)

.^ Element-by-element power

.' Unconjugated array transpose

(does not take complex conjugate, unlike a regular [no dot] matrix transpose)



## Some Special Matrices

Square matrix:  $m$  (# rows) =  $n$  (# columns)

Symmetric matrix: subset of square matrices where  $A^T = A$

Diagonal matrix: subset of square matrices where elements off the principal diagonal are zero,  $a_{ij} = 0$  if  $i \neq j$

Identity or unit matrix: special diagonal matrix where all principal diagonal elements are 1



```
>> a=[1 2 3;4 5 6;7 8 9]
```

```
a =
```

```
1 2 3
4 5 6
7 8 9
```

```
>> c=trace(a)
```

```
c =
15
```

```
>> a=[.96 -.28; .28 .96]
```

```
a =
```

```
0.9600 -0.2800
0.2800 0.9600
```

```
>> inv(a)
```

```
ans =
```

```
0.9600 0.2800
-0.2800 0.9600
```

```
>> a'*a
```

```
ans =
```

```
1 0
0 1
```

```
>>
```

**trace** of a Matrix is  $\text{Tr}(A) =$

$$\sum_{i=1}^N a_{ii}$$

(the sum of the diagonal entries)

In matlab use `trace(A)`

a matrix **A** is **orthonormal** if

$$A^T = A^{-1}$$

and in this case

$$AA^T = I$$