#### Lecture 3: Solving Equations, Curve Fitting, and Numerical Techniques

## **Outline**

- (1) Linear Algebra
- (2) Polynomials
- (3) Optimization
- (4) Differentiation/Integration
- (5) Differential Equations

# **Systems of Linear Equations**

- Given a system of linear equations
  - x+2y-3z=5-3x-y+z=-8x-y+z=0

MATLAB makes linear algebra fun!

- Construct matrices so the system is described by Ax=b
  - » A=[1 2 -3;-3 -1 1;1 -1 1];
    » b=[5;-8;0];
- And solve with a single line of code!
  - $x=A\b;$ 
    - $\rightarrow$  x is a 3x1 vector containing the values of x, y, and z
- The \ will work with square or rectangular systems.
- Gives least squares solution for rectangular systems. Solution depends on whether the system is over or underdetermined.

## Worked Example: Linear Algebra

Solve the following systems of equations:

> System 1:  

$$x + 4y = 34$$
  
 $-3x + y = 2$   
> System 2:  
 $2x - 2y = 4$ 

```
System 2: 
2x-2y=4
-x+y=3
3x+4y=2

*** A=[2 -2; **]

*** b=[4;3;2]

*** rank (A)

> rectang

*** x=A\b;
```

```
b=[34;2];
» rank(A)
» x=inv(A)*b;
x=A\b;
b = [4;3;2];
  > rectangular matrix
  gives least squares solution
» error=abs(A*x1-b)
```

## **More Linear Algebra**

- Given a matrix
  - » mat=[1 2 -3;-3 -1 1;1 -1 1];
- Calculate the rank of a matrix
  - » r=rank(mat);
    - > the number of linearly independent rows or columns
- Calculate the determinant
  - » d=det(mat);
    - > mat must be square; matrix invertible if det nonzero
- Get the matrix inverse
  - » E=inv(mat);
    - if an equation is of the form A\*x=b with A a square matrix, x=A\b is (mostly) the same as x=inv(A)\*b
- Get the condition number
  - » c=cond(mat); (or its reciprocal: c = rcond(mat);)
    - $\rightarrow$  if condition number is large, when solving A\*x=b, small errors in b can lead to large errors in x (optimal c==1)

## **Matrix Decompositions**

- MATLAB has many built-in matrix decomposition methods
- The most common ones are
  - » [V,D]=eig(X)
    - Eigenvalue decomposition
  - > [U,S,V] = svd(X)
    - Singular value decomposition
  - $\gg [Q,R]=qr(X)$ 
    - > QR decomposition
  - $\gg [L,U]=lu(X)$ 
    - > LU decomposition
  - » R=chol(X)
    - Cholesky decomposition (R must be positive definite)

# **Exercise: Fitting Polynomials**

• Find the best second-order polynomial that fits the points: (-1,0), (0,-1), (2,3).

$$a(-1)^{2} + b(-1) + c = 0$$

$$a(0)^{2} + b(0) + c = -1$$

$$a(2)^{2} + b(2) + c = 3$$

### **Outline**

- (1) Linear Algebra
- (2) Polynomials
- (3) Optimization
- (4) Differentiation/Integration
- (5) Differential Equations

# **Polynomials**

- Many functions can be well described by a high-order polynomial
- MATLAB represents a polynomials by a vector of coefficients
  - > if vector P describes a polynomial

$$ax^3+bx^2+cx+d$$

- $P=[1\ 0\ -2]$  represents the polynomial  $x^2-2$
- $P=[2\ 0\ 0\ 0]$  represents the polynomial  $2x^3$

## **Polynomial Operations**

- P is a vector of length N+1 describing an N-th order polynomial
- To get the roots of a polynomial
  - » r=roots(P)
    - r is a vector of length N
- Can also get the polynomial from the roots
  - » P=poly(r)
    - r is a vector length N
- To evaluate a polynomial at a point
  - » y0=polyval(P,x0)
    - > x0 is a single value; y0 is a single value
- To evaluate a polynomial at many points
  - » y=polyval(P,x)
    - > x is a vector; y is a vector of the same size

## **Polynomial Fitting**

MATLAB makes it very easy to fit polynomials to data

```
Given data vectors X=[-1 0 2] and Y=[0 -1 3]
» p2=polyfit(X,Y,2);
» finds the best (least-squares sense) second-order polynomial that fits the points (-1,0),(0,-1), and (2,3)
» see help polyfit for more information
» plot(X,Y,'o', 'MarkerSize', 10);
» hold on;
» x = -3:.01:3;
» plot(x,polyval(p2,x), 'r--');
```

# **Exercise: Polynomial Fitting**

• Evaluate  $y = x^2$  for x=-4:0.1:4.

 Add random noise to these samples. Use randn. Plot the noisy signal with \_ markers

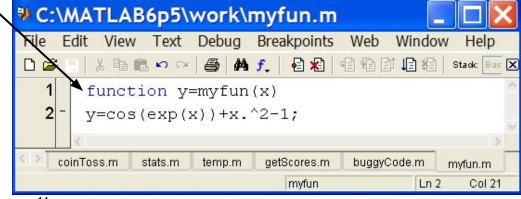
- Fit a 2<sup>nd</sup> degree polynomial to the noisy data
- Plot the fitted polynomial on the same plot, using the same x values and a red line

## **Outline**

- (1) Linear Algebra
- (2) Polynomials
- (3) Optimization
- (4) Differentiation/Integration
- (5) Differential Equations

# **Nonlinear Root Finding**

- Many real-world problems require us to solve f(x)=0
- Can use fzero to calculate roots for any arbitrary function
- fzero needs a function passed to it.
- We will see this more and more as we delve into solving equations.
- Make a separate function file
  - » x=fzero('myfun',1)
  - » x=fzero(@myfun,1)
    - ➤ 1 specifies a point close to where you think the root is



## Minimizing a Function

- fminbnd: minimizing a function over a bounded interval
  - » x=fminbnd('myfun',-1,2);
    - > myfun takes a scalar input and returns a scalar output
    - $\rightarrow$  myfun(x) will be the minimum of myfun for  $-1 \le x \le 2$
- fminsearch: unconstrained interval
  - » x=fminsearch('myfun',.5)
    - $\rightarrow$  finds the local minimum of myfun starting at x=0.5
- Maximize g(x) by minimizing f(x)=-g(x)
- Solutions may be local!

## **Anonymous Functions**

- You do not have to make a separate function file
  - » x=fzero(@myfun,1)
    - ➤ What if myfun is really simple?
- Instead, you can make an anonymous function

```
» x=fzero(@(x)(cos(exp(x))+x.^2-1), 1);
input function to evaluate
```

```
x=fminbnd(@(x) (cos(exp(x))+x.^2-1),-1,2);
```

- Can also store the function handle
  - » func=@(x) (cos(exp(x))+x.^2-1);
    » func(1:10);

## **Optimization Toolbox**

- If you are familiar with optimization methods, use the optimization toolbox
- Useful for larger, more structured optimization problems
- Sample functions (see help for more info)
  - » linprog
    - > linear programming using interior point methods
  - » quadprog
    - quadratic programming solver
  - » fmincon
    - > constrained nonlinear optimization

## **Exercise: Min-Finding**

- Find the minimum of the function  $f(x) = \cos(4x)\sin(10x)e^{-|x|}$  over the range  $-\pi$  to  $\pi$ . Use **fminbnd**.
- Plot the function on this range to check that this is the minimum.

## **Digression: Numerical Issues**

- Many techniques in this lecture use floating point numbers
- This is an approximation!
- Examples:

#### A Word of Caution

- MATLAB knows no fear!
- Give it a function, it optimizes / differentiates / integrates
  - That's great! It's so powerful!
- Numerical techniques are powerful **but** not magic
- Beware of overtrusting the solution!
  - > You will get an answer, but it may not be what you want
- Analytical forms may give more intuition
  - Symbolic Math Toolbox

## **Outline**

- (1) Linear Algebra
- (2) Polynomials
- (3) Optimization
- (4) Differentiation/Integration
- (5) Differential Equations

### **Numerical Differentiation**

- 2D gradient
  - » [dx,dy]=gradient(mat);
- Higher derivatives / complicated problems: Fit spline (see help)

## **Numerical Integration**

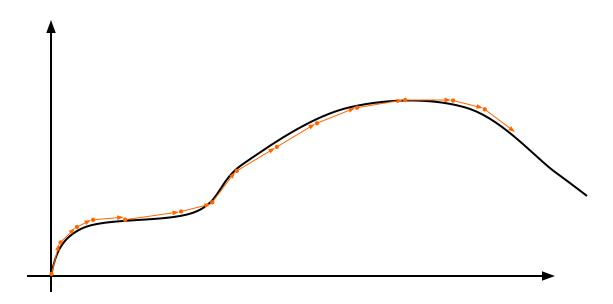
- MATLAB contains common integration methods
- Adaptive Simpson's quadrature (input is a function)
  - » q=quad('myFun',0,10)
    - > q is the integral of the function myFun from 0 to 10
  - » q2=quad(@(x) sin(x).\*x,0,pi)
    - > q2 is the integral of sin(x).\*x from 0 to pi
- Trapezoidal rule (input is a vector)
  - » x=0:0.01:pi;
  - » z=trapz(x,sin(x))
    - > z is the integral of sin(x) from 0 to pi
  - » z2=trapz(x,sqrt(exp(x))./x)
    - > z2 is the integral of  $\sqrt{e^x}/x$  from 0 to pi

#### **Outline**

- (1) Linear Algebra
- (2) Polynomials
- (3) Optimization
- (4) Differentiation/Integration
- (5) Differential Equations

#### **ODE Solvers: Method**

 Given a differential equation, the solution can be found by integration:



- > Evaluate the derivative at a point and approximate by straight line
- > Errors accumulate!
- > Variable timestep can decrease the number of iterations

### **ODE Solvers: MATLAB**

- MATLAB contains implementations of common ODE solvers
- Using the correct ODE solver can save you lots of time and give more accurate results

#### » ode23

➤ Low-order solver. Use when integrating over small intervals or when accuracy is less important than speed

#### » ode45

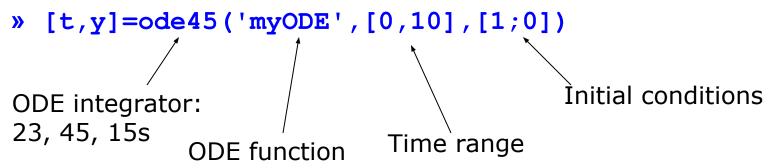
➤ High order (Runge-Kutta) solver. High accuracy and reasonable speed. Most commonly used.

#### » ode15s

> Stiff ODE solver (Gear's algorithm), use when the diff eq's have time constants that vary by orders of magnitude

## **ODE Solvers: Standard Syntax**

To use standard options and variable time step



#### Inputs:

- > ODE function name (or anonymous function). This function should take inputs (t,y), and returns dy/dt
- > Time interval: 2-element vector with initial and final time
- ➤ Initial conditions: column vector with an initial condition for each ODE. This is the first input to the ODE function
- > Make sure all inputs are in the same (variable) order

#### Outputs:

- > t contains the time points
- > y contains the corresponding values of the variables

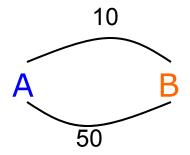
### **ODE Function**

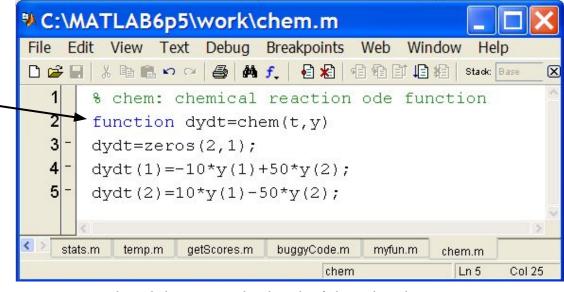
- The ODE function must return the value of the derivative at a given time and function value
- Example: chemical reaction
  - > Two equations

$$\frac{dA}{dt} = -10A + 50B$$

$$\frac{dB}{dt} = 10A - 50B$$

- ➤ ODE file:
  - y has [A;B]
  - dydt has
    [dA/dt;dB/dt]



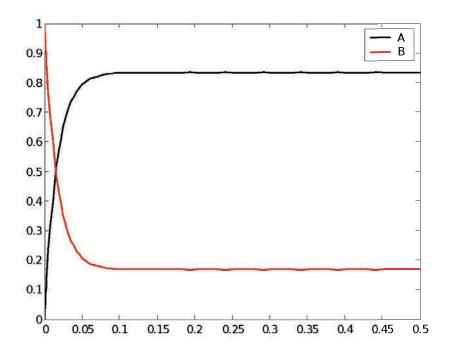


## **ODE Function: viewing results**

To solve and plot the ODEs on the previous slide:

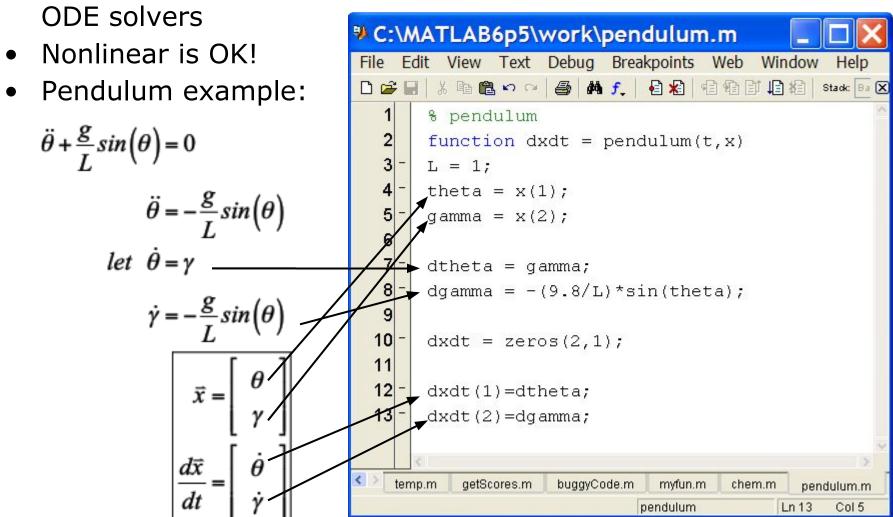
## **ODE Function: viewing results**

The code on the previous slide produces this figure



# **Higher Order Equations**

Must make into a system of first-order equations to use



## **Plotting the Output**

We can solve for the position and velocity of the pendulum:

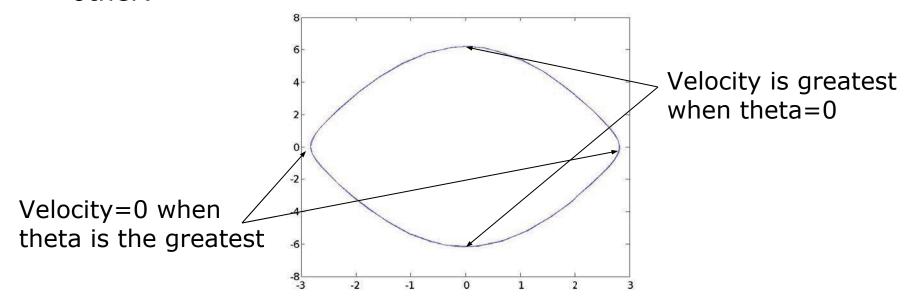
```
» [t,x]=ode45('pendulum',[0 10],[0.9*pi 0]);
        > assume pendulum is almost horizontal
     » plot(t,x(:,1));
     » hold on;
     » plot(t,x(:,2),'r');
     » legend('Position','Velocity');
                                           Position
                                           Velocity
                                                       Velocity (m/s)
Position in terms of
angle (rad)
                     -2
                     -4
```

## **Plotting the Output**

Or we can plot in the phase plane:

```
» plot(x(:,1),x(:,2));
» xlabel('Position');
» yLabel('Velocity');
```

 The phase plane is just a plot of one variable versus the other:



## **ODE Solvers: Custom Options**

- MATLAB's ODE solvers use a variable timestep
- Sometimes a fixed timestep is desirable

```
» [t,y]=ode45('chem',[0:0.001:0.5],[0 1]);
```

- > Specify timestep by giving a vector of (increasing) times
- > The function value will be returned at the specified points
- You can customize the error tolerances using odeset

```
» options=odeset('RelTol',1e-6,'AbsTol',1e-10);
```

- » [t,y]=ode45('chem',[0 0.5],[0 1],options);
  - > This guarantees that the error at each step is less than RelTol times the value at that step, and less than AbsTol
  - > Decreasing error tolerance can considerably slow the solver
  - > See doc odeset for a list of options you can customize

### **Exercise: ODE**

- Use ode45 to solve for y(t) on the range t=[0 10], with initial condition y(0)=10 and dy/dt=-t y/10
- Plot the result.

#### **Exercise: ODE**

- Use ode45 to solve for y(t) on the range t=[0 10], with initial condition y(0) = 10 and dy/dt = -t y/10
- Plot the result.
- Make the following function

```
» function dydt=odefun(t,y)
» dydt=-t*y/10;
```

Integrate the ODE function and plot the result

```
» [t,y]=ode45('odefun',[0 10],10);
```

Alternatively, use an anonymous function

```
[t,y]=ode45(@(t,y) -t*y/10,[0 10],10);
```

Plot the result

```
» plot(t,y);xlabel('Time');ylabel('y(t)');
```

## **Exercise: ODE**

• The integrated function looks like this:

