

Robot Learning

2. Numerical Method

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Numerical Method

1. Equation Solver
2. Multiple Nonlinear Equation
3. Linear Regression
4. Nonlinear Regression
5. Stochastic Regression (RANSAC)
6. Optimization

1

Deterministic Vs. Stochastic Method

First Start with Numerical Methods

Deterministic Vs. Stochastic World

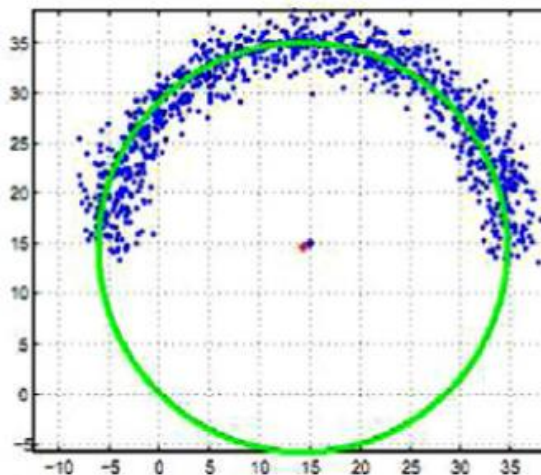
- Deterministic World
 - Everything must be determined.
 - Everything is Understood by Modeling
 - Manipulator-based Robotics, PID or even Robust Control
 - Pseudo code : $a = 3$
- Stochastic World
 - Everything is PROBABILISTICALLY determined
 - Every phenomenon occurs by Probabilistic Results
 - Autonomous locomotion(SLAM), Learning, and so on
 - Pseudo code:

$$a \square N(3, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} \exp \frac{-1}{2} \left(\frac{x-3}{\sigma} \right)^2$$

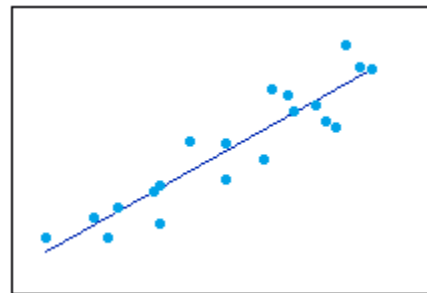
Equality "=" is
NOT allowed.

Why we learn first Deterministic Method?

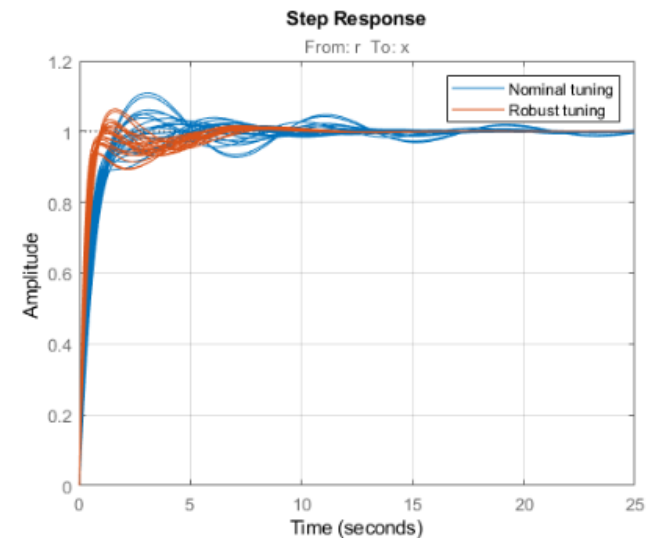
- Some methods look like Non-Deterministic Problem.
- Fitting, Regression, Control



Regression



Curve fitting



Control

- Ex) Control tries to be in the desired goal in spite of unnecessary system dynamics, Is it probabilistic?
- Absolutely, Not.

Many Deterministic Methods are based on Mathematical Model

- Controller is well designed to OVERCOME marginal error → Deterministic method
- Fitting or Regression is to Minimize Errors → Deterministic Method
- Then, What is Stochastic Process?
 - Probabilistically, next state is determined.
- 1.Many Deterministic Methods are applied to Learning Methods
- 2. We learn the differences of

Deterministic and Stochastic Methods.

Numerical Method:

The goal is to find the solution

- Analytical Solution

$$2a = 3$$

$$\therefore a = 1.5$$

- Numerical Solution(or Computational Solution)
- Solve $2a=3$ equation by using computer program

Equation : $2a = 3$

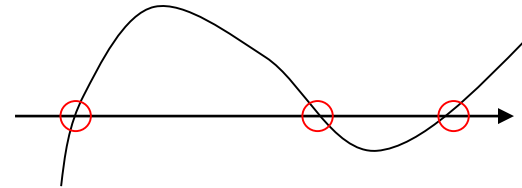
Function : $f(a) = 2a - 3$

How to find $f(a) = 2a - 3 = 0$?

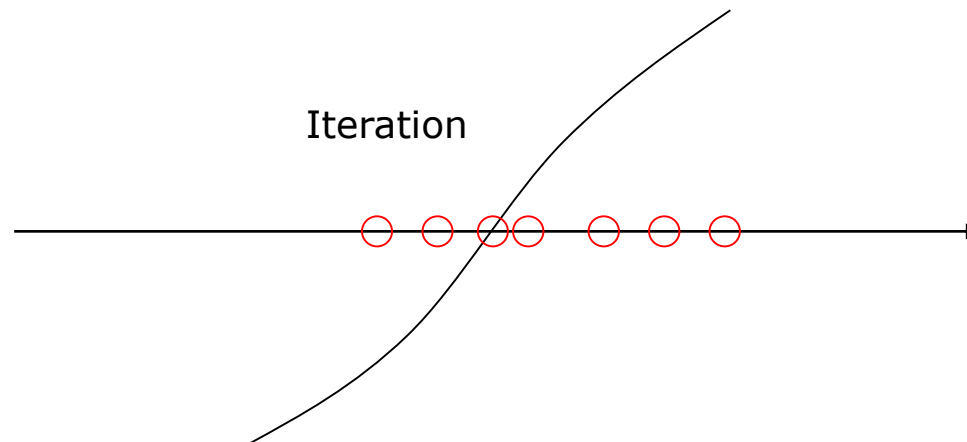
Numerical Method

Solve Equation(or Finding Roots)

- Solve Equation
 - Equation: $f(x,y,s,t) = 0$



- How to solve it by Numerical Method?
- Iteratively, find a solution by a Computer



Why Numerical Methods are Required?

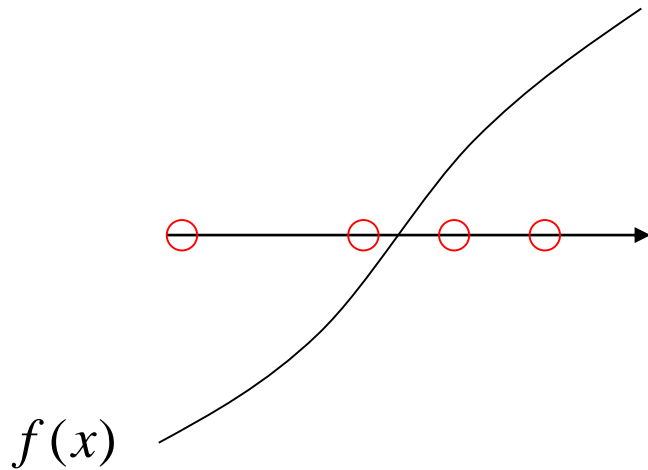
- 1. Equations are Complex
 - Remind Robot Kinematics or Dynamics are very complex
 - Generally, we CANNOT solve it by analytical methods

$$\begin{pmatrix} m_1 l_1^2 + m_2 l_1^2 + m_2 l_2^2 + 2m_2 l_1 l_2 c_2 & m_2 l_2^2 + m_2 l_1 l_2 c_2 \\ m_2 l_2^2 + m_2 l_1 l_2 c_2 & m_2 l_2^2 \end{pmatrix} \begin{pmatrix} \ddot{\theta}_1 \\ \ddot{\theta}_2 \end{pmatrix} + \begin{pmatrix} -m_2 l_1 l_2 (2\dot{\theta}_1 + \dot{\theta}_2) s_2 \dot{\theta}_2 \\ m_2 l_1 l_2 \dot{\theta}_1^2 s_2 \end{pmatrix} + g \begin{pmatrix} m_1 l_1 c_1 + m_2 (l_2 c_{12} + l_1 c_1) \\ m_2 l_2 c_{12} \end{pmatrix} = 0$$

- 2. We learn Convergence by Iterative Methods
 - Iteration: In Each turn, a method moves to the solution
 - Convergence or Divergence Problem
 - **Dynamic Programming:** Learning, Control, Numerical Methods, etc.

Numerical Method: 1. Bisection Method

- See test1.m



Algorithm

1. Start with two points

Assume

$$x_L < x_s < x_R$$

$$f(x_L)f(x_R) < 0$$

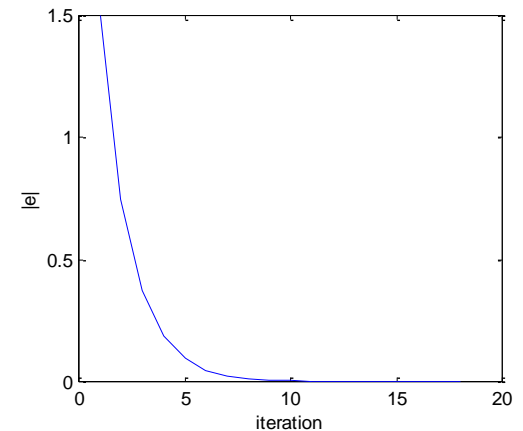
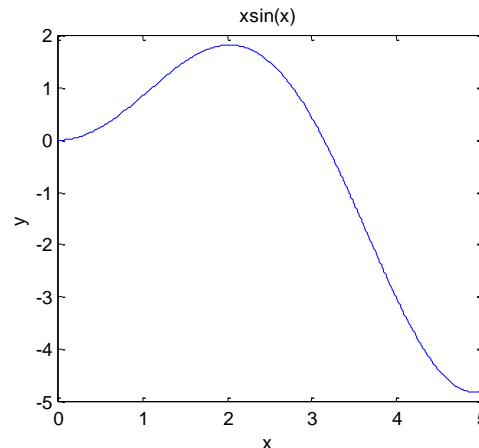
2. Find mid point

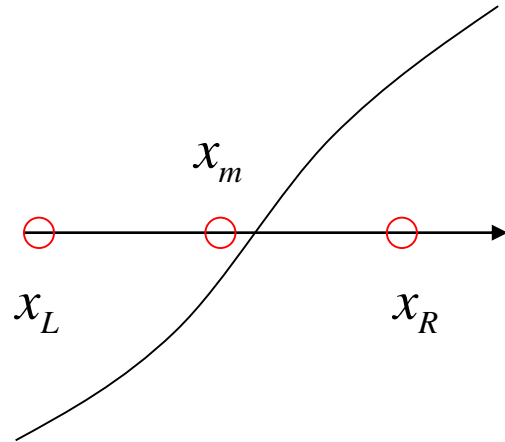
$$x_m = \frac{x_L + x_R}{2}$$

$$f(x_m)f(x_L) < 0 : x_m \rightarrow x_R$$

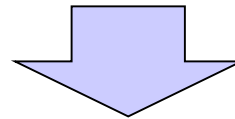
$$f(x_m)f(x_R) < 0 : x_m \rightarrow x_L$$

$$x \sin x = 0$$



1st turn

$$x_m = \frac{x_L + x_R}{2}$$

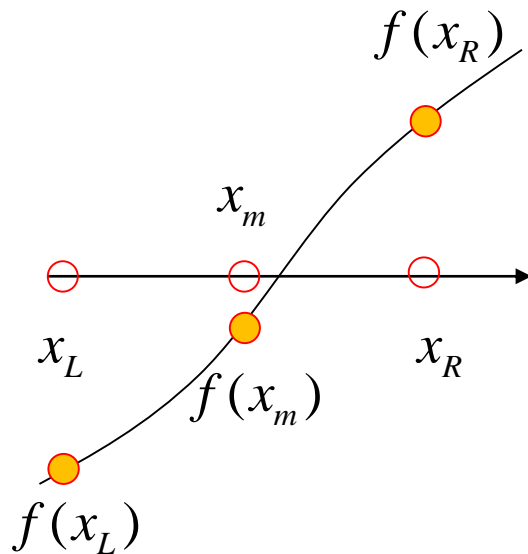
Question: x_m is left or Right?Solution, x is always between
 $x_L < x < x_R$

$$\therefore f(x_L)f(x_R) < 0$$

Thus, x_m will be left or right

if $f(x_L)f(x_m) < 0$, new $x_R = x_m$

if $f(x_R)f(x_m) < 0$, new $x_L = x_m$

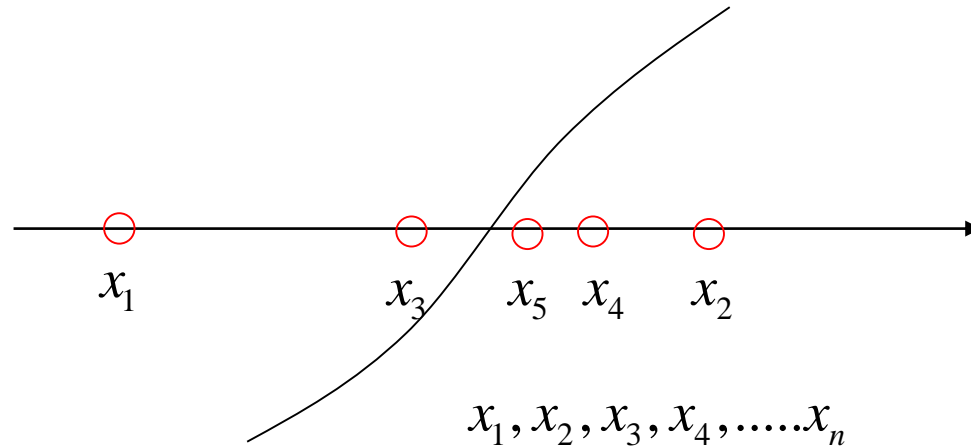


$\rightarrow x_L \leftarrow x_m$

Convergence

- Modeling it as you have Learned in other classes

$$x_m = \frac{x_L + x_R}{2}$$



$$\rightarrow |e| \propto |x_L - x_R|$$

$$|e_1| = |x_2 - x_1|$$

$$|e_2| = |x_3 - x_2|$$

...

$$\therefore |x_3 - x_2| \leq \frac{|x_2 - x_1|}{2}$$

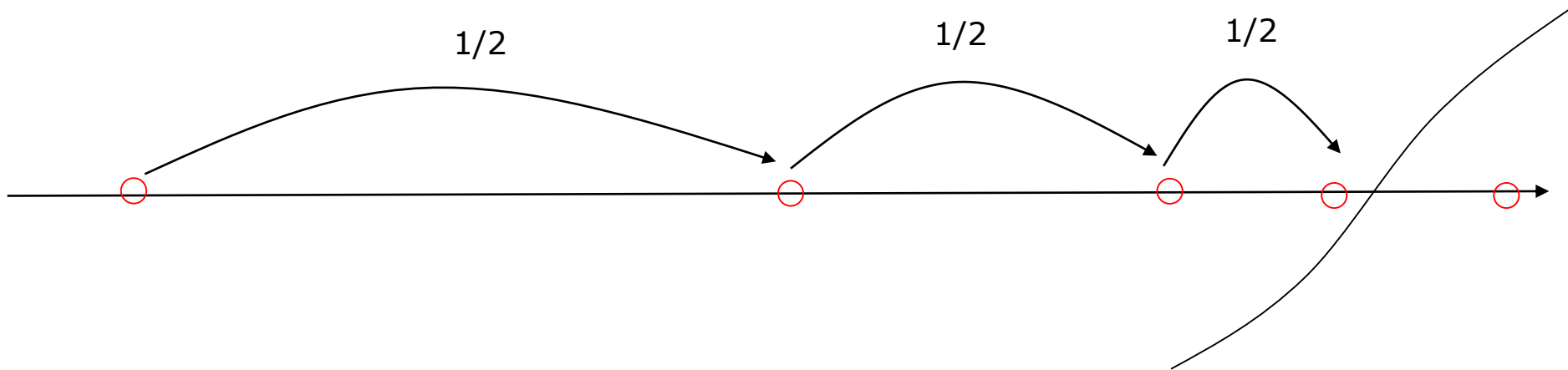
$$|e_n| = |x_{n+1} - x_n| \leq \frac{|x_n - x_{n-1}|}{2} \leq \frac{|x_{n-1} - x_{n-2}|}{4} \dots \leq \frac{|x_2 - x_1|}{2^{n-1}}$$

$$\lim_{n \rightarrow \infty} |e_n| \leq \lim_{n \rightarrow \infty} \frac{|x_2 - x_1|}{2^{n-1}} = |x_2 - x_1| \lim_{n \rightarrow \infty} \frac{1}{2^{n-1}} = 0$$

$$\therefore \lim_{n \rightarrow \infty} |e_n| \leq 0$$

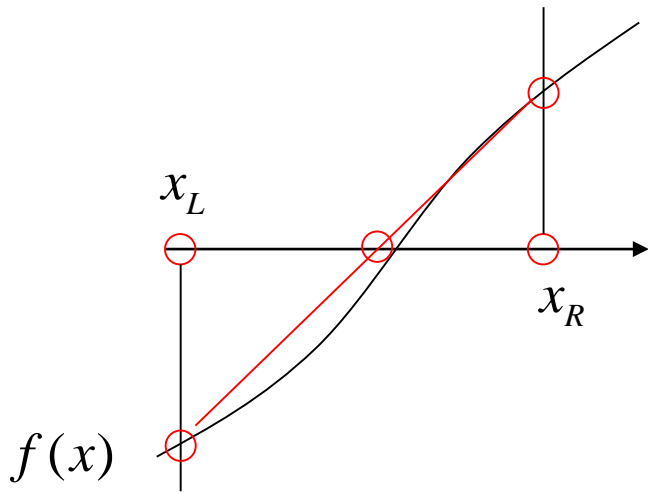


Bisection Method is Too Slow



- If the initial X_L or X_R is too far from a solution, $1/2$ is NOT so Fast!
- How can we speed up?
 - Ratio of a Function is better than $1/2$.

Numerical Method: 2. Secant Method



Algorithm

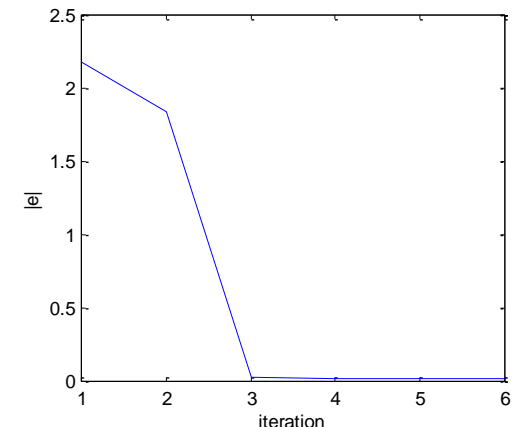
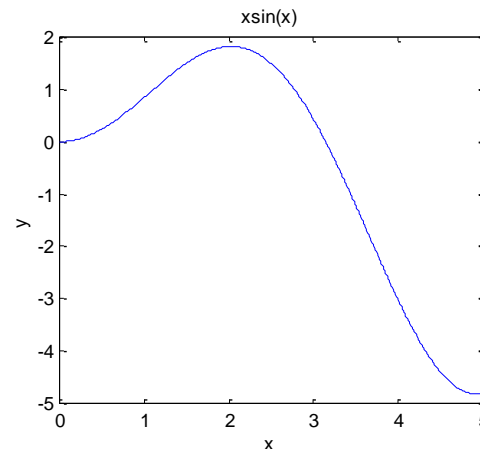
1. Start with two points

$$y = \frac{f(x_R) - f(x_L)}{x_R - x_L} (x - x_L) + f(x_L) = 0$$

2. Find mid point

$$\frac{f(x_R) - f(x_L)}{x_R - x_L} (x_m - x_L) + f(x_L) = 0$$

- See test2.m

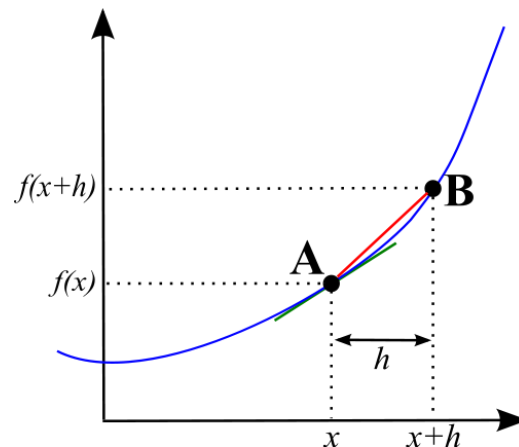


Why Secant Method is Faster than Bisection Method?

- Focus on a new Mid point.

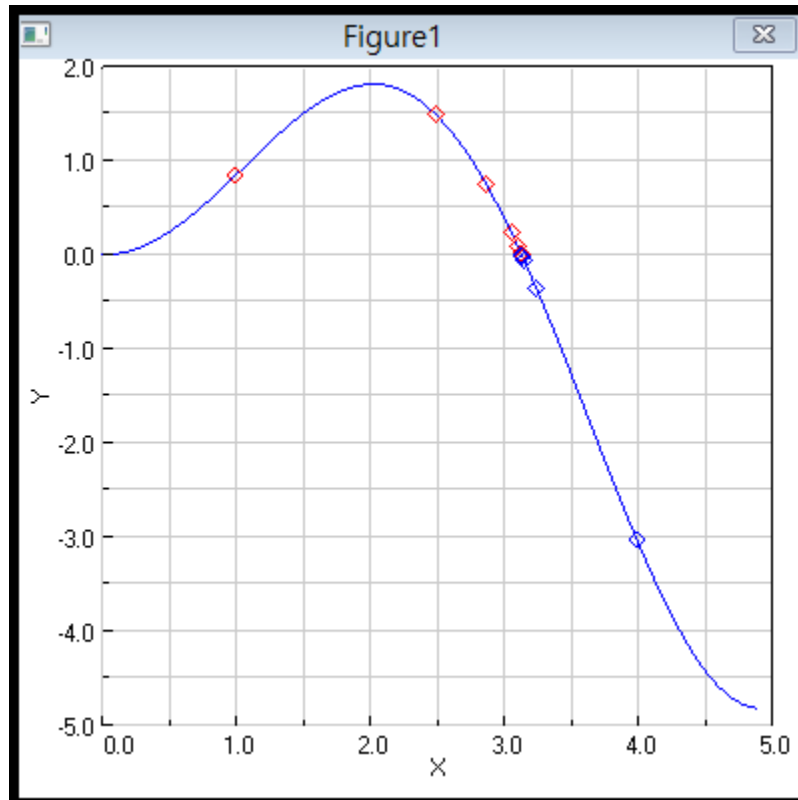
$$y = \frac{f(x_R) - f(x_L)}{x_R - x_L} (x - x_L) + f(x_L) = 0$$

- Function Ratio is good for faster convergence.
- What does the Function Ratio remind us of?

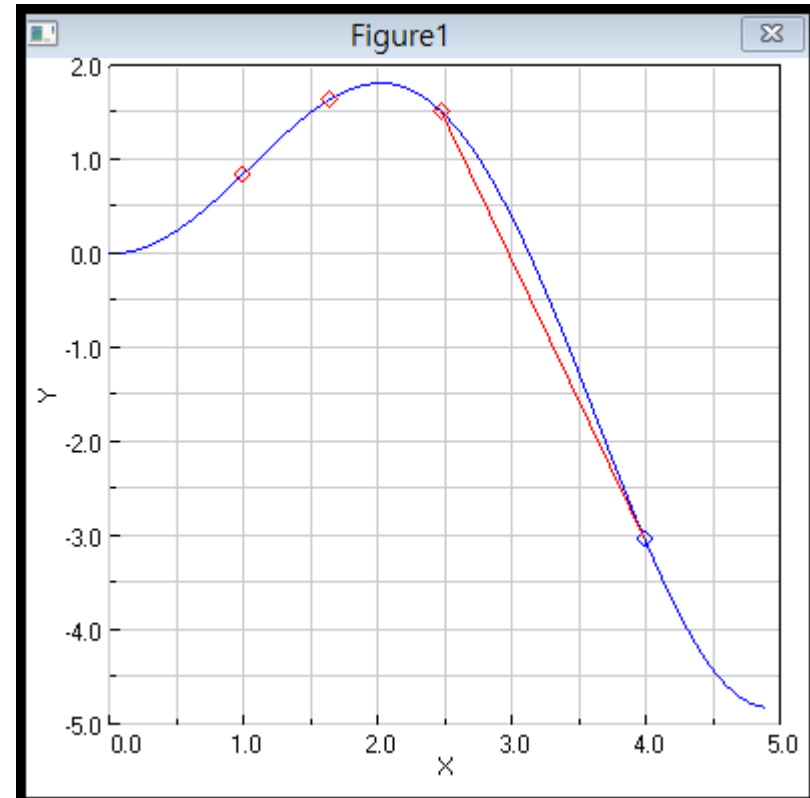


Differentiation!

Example) I2bisect and I2secant ex/ml/I2bisect and ex/ml/I2secant



```
import I2bisect
I2bisect.test(1,4)
```



```
import I2secant
I2secant.test(1,4)
```

ex/ml/l2bisect and ex/ml/l2secant

```
def test(xl,xr):
    eold = 0;
    drawfunc()

    for i in range(0,100):
        fl = func(xl)
        fr = func(xr)

        if (fl*fr>=0):
            print("Wrong initial guess.");
            return

        # New mid
        xm = (xl+xr)/2;
        fm = func(xm)

        # print data
        loop.io.print(i,":",xl,xr,"xm=",xm)

        # draw xl and xr
        graph(2)
        plot(xl,fl,"rd")
        graph(3)
        plot(xr,fr,"bd")

        # find New xl or xr
        if (fl*fm<0): # xl and xm as new guesses
            xr = xm
        elif (fr*fm<0): # xm and xr as new guesses
            xl = xm

        error = abs(xl-xr)

        # stop or not
        if (abs(error-eold)<1e-5):
            break;

        eold = error
        loop.pause()

    print("XL=",xl,"XR=",xr, "Error",error)
```

```
def test(xl,xr):
    clear()
    eold = 0;
    drawfunc()

    for i in range(0,100):
        loop.pause()
        fl = func(xl)
        fr = func(xr)

        if (fl*fr>=0):
            print("Wrong initial guess.");
            return

        # New mid with Secant Method
        r = (fr-fl)/(xr-xl)
        xm = xl-fl/r
        fm = func(xm)

        # print data
        loop.io.print(i,":",xl,xr,"xm=",xm)

        # draw xl,xr and line
        graph(2)
        plot(xl,fl,"rd")
        graph(3)
        plot(xr,fr,"bd")
        drawline(xl,fl, xr,fr)

        # find New xl or xr
        if (fl*fm<0): # xl and xm as new guesses
            xr = xm
        elif (fr*fm<0): # xm and xr as new guesses
            xl = xm

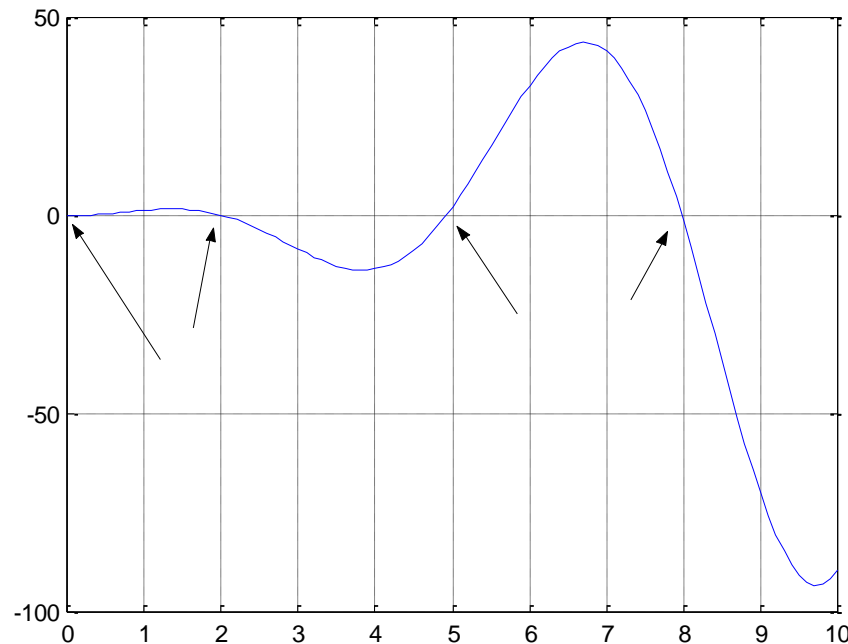
        error = abs(xl-xr)

        # stop or not
        if (abs(error-eold)<1e-5):
            break;
```



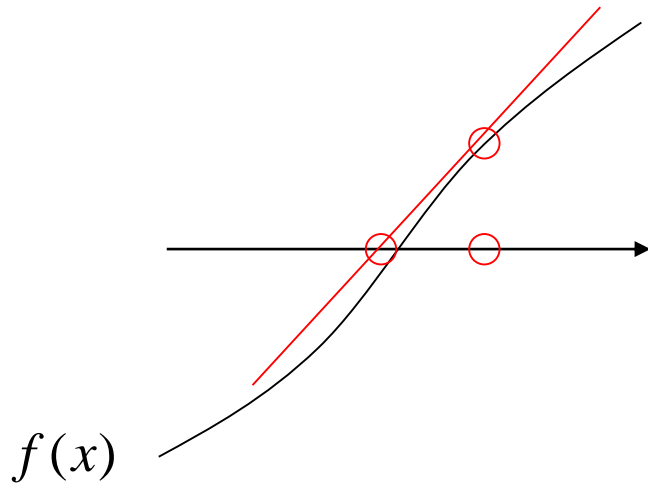
HW1) Find the Solution X with Bisection and Secant Methods

- Given Equation $x \cdot \sin(x) + x^2 \cdot \cos(x) = 0$
- Condition $0 < x < 10$
- Find all solution, x within the range $[0, 10]$



Numerical Method:

3. Newton-Raphson (NR) Method



Algorithm

1. Start with one point(a GUESS value)

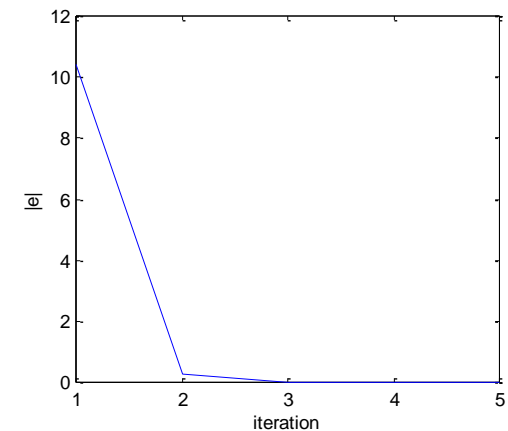
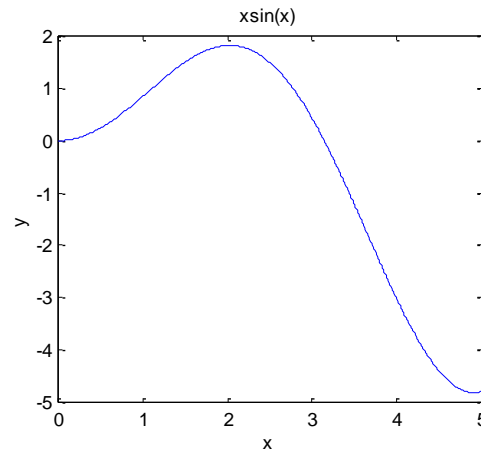
$$y = f'(x_n)(x - x_n) + f(x_n) = 0$$

2. Find a new point

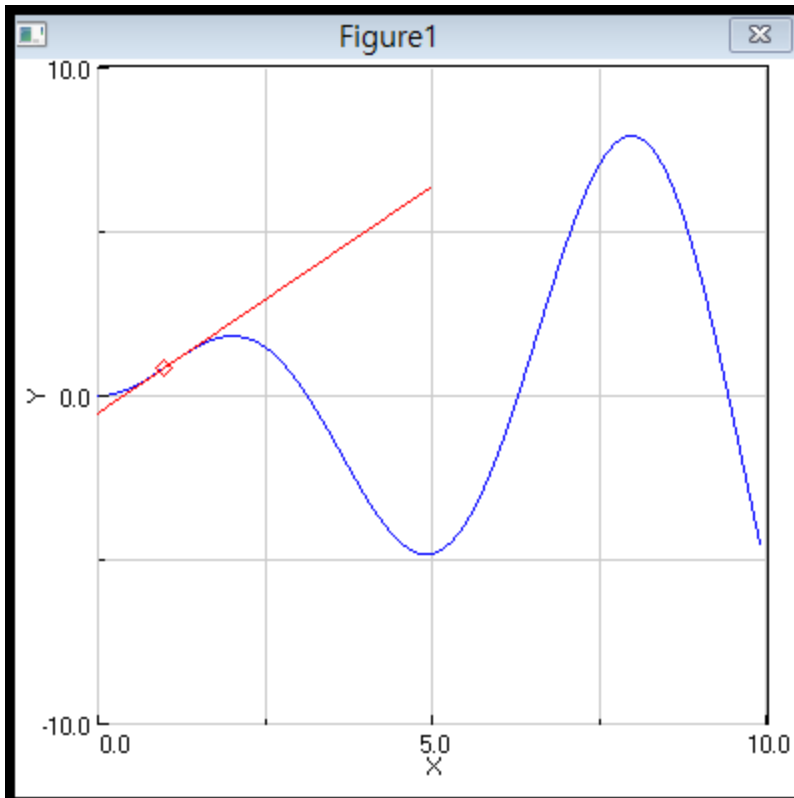
$$f'(x_n)(x_{n+1} - x_n) + f(x_n) = 0$$

$$x_{n+1} = -\frac{f(x_n)}{f'(x_n)} + x_n$$

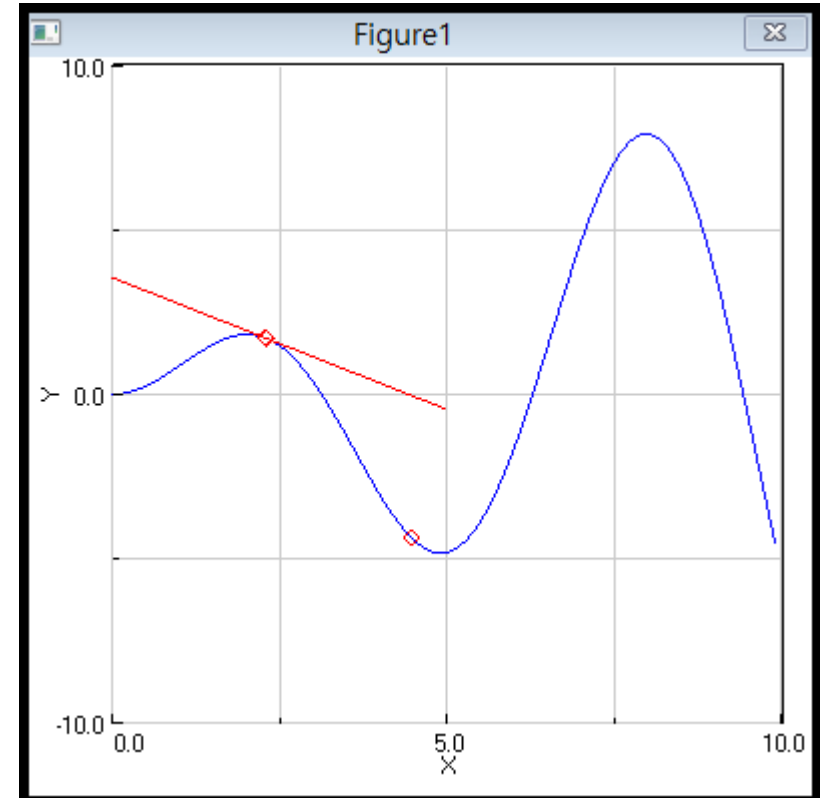
- See test3.m



Example) Newton-Raphson ex/ml/l2nr



```
import l2nr
l2nr.test(0.1)
```

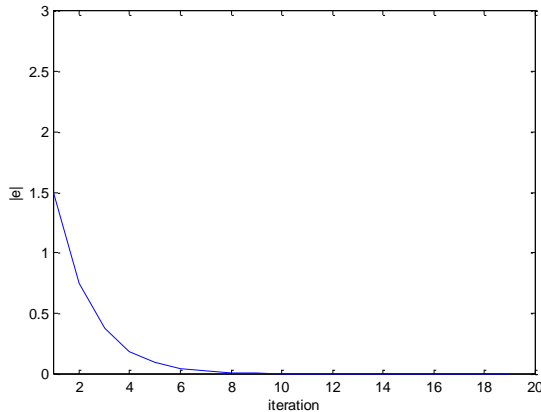


```
import l2nr
l2nr.test(2.3)
```

Comparison Three Cases

Bisection

$$x_m = \frac{x_L + x_R}{2}$$



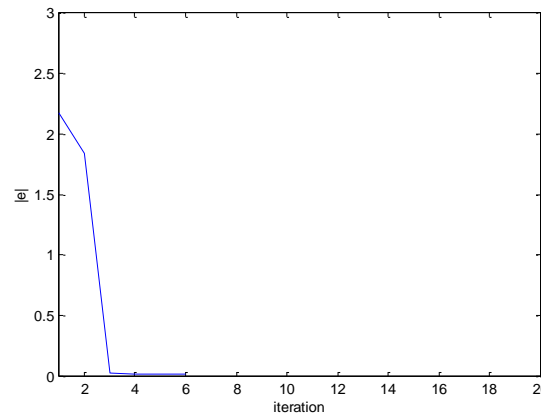
Exponentially Converged

Good for convergence:

- Stable
- Continuous and smooth convergence but Slow

Secant

$$x_m = -\frac{f(x_L)}{\left(\frac{f(x_R) - f(x_L)}{x_R - x_L}\right)} + x_L$$

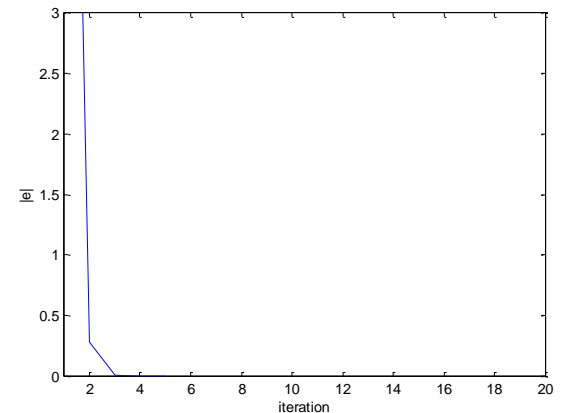


Using Ratio = Linearization

Many Nonlinear problems are approximated for linearity.

Newton-Raphson

$$x_{n+1} = -\frac{f(x_n)}{f'(x_n)} + x_n$$



Using Ratio = Linearized method with Differentiation

- Fast
- But, initial guess is important for stability



Solution Analysis

- Bisection Method

- Is it Bad?

- Exponentially Converged \rightarrow Stable $|e_n| \propto 2^{-k}$

- Secant Method(Faster)

$$x_m = -\frac{f(x_L)}{\left(\frac{f(x_R) - f(x_L)}{x_R - x_L}\right)} + x_L \quad \text{if } \frac{f(x_R) - f(x_L)}{x_R - x_L} = 0, \text{ it fails.}$$

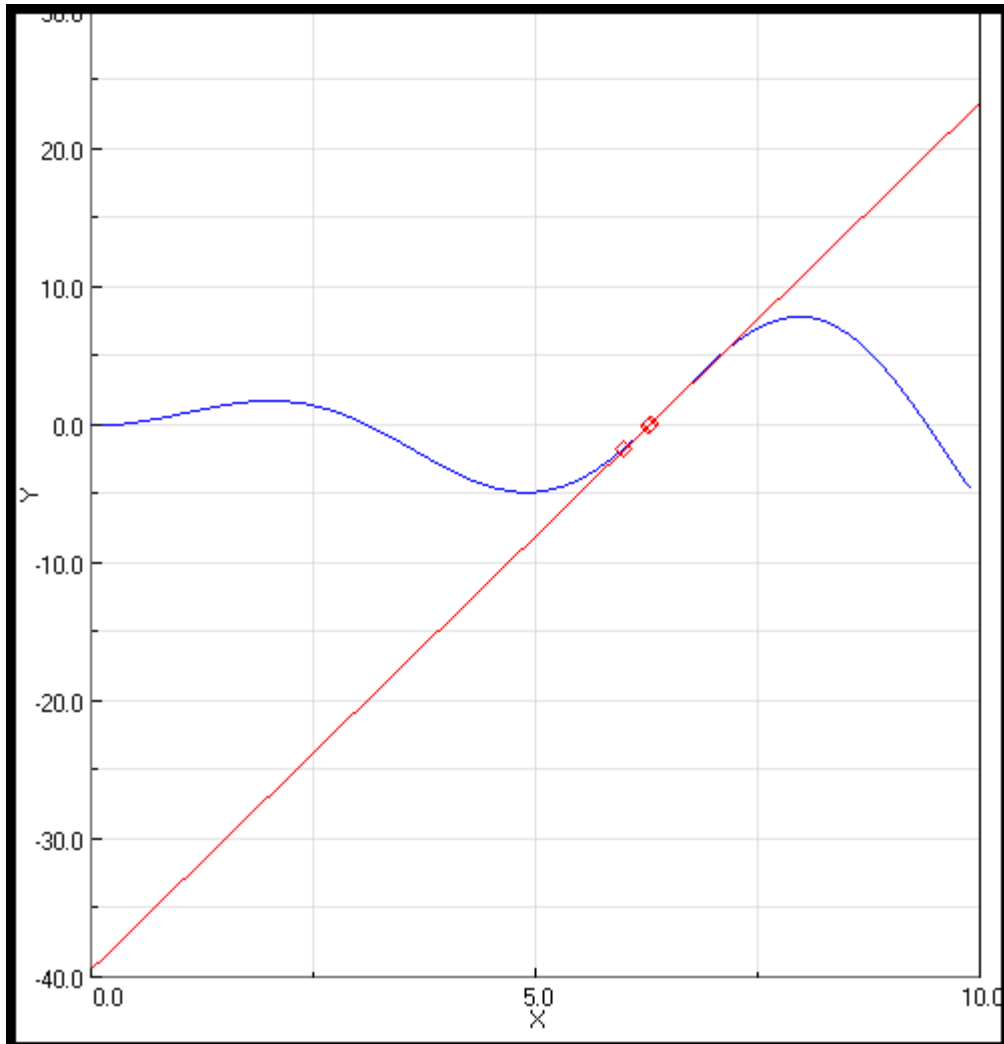
- NR Method(much Faster)

$$x_{n+1} = -\frac{f(x_n)}{f'(x_n)} + x_n \quad \text{if } f'(x_n) = 0, \text{ it fails.}$$

- Sometimes, it becomes very unstable.

HW2)N-R method

How we get solution $x= 6.28318$?



- Hint: Change your initial guess

Numerical Methods for Multiple Equations

3

$f(x,y)=0$ and $g(x,y)=0$

$$\begin{pmatrix} m_1 l_1^2 + m_2 l_1^2 + m_2 l_2^2 + 2m_2 l_1 l_2 c_2 & m_2 l_2^2 + m_2 l_1 l_2 c_2 \\ m_2 l_2^2 + m_2 l_1 l_2 c_2 & m_2 l_2^2 \end{pmatrix} \begin{pmatrix} \ddot{\theta}_1 \\ \ddot{\theta}_2 \end{pmatrix} + \begin{pmatrix} -m_2 l_1 l_2 (2\dot{\theta}_1 + \dot{\theta}_2) s_2 \dot{\theta}_2 \\ m_2 l_1 l_2 \dot{\theta}_1^2 s_2 \end{pmatrix} + g \begin{pmatrix} m_1 l_1 c_1 + m_2 (l_2 c_{12} + l_1 c_1) \\ m_2 l_2 c_{12} \end{pmatrix} = 0$$

$$f(\ddot{\theta}_1, \ddot{\theta}_2, \dot{\theta}_1, \dot{\theta}_2, \theta_1, \theta_2) = 0$$

$$g(\ddot{\theta}_1, \ddot{\theta}_2, \dot{\theta}_1, \dot{\theta}_2, \theta_1, \theta_2) = 0$$

Multi Dimension Equation

- 1 Dim. Problem
 - Solve x with $f(x)=0$

$$x_{n+1} = -\frac{f(x_n)}{f'(x_n)} + x_n, \text{ Iteration} \quad \therefore x_n \rightarrow x_s$$

- 2 Dim. Problem
 - Solve vector x

$$f(x, y) = 0$$

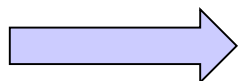
$$g(x, y) = 0$$

$$\text{Ex) } f(x, y) = x^2 + y^2 - 3 = 0$$

$$g(x, y) = x + y - 1 = 0$$

$$\hat{X}_{n+1} = \begin{pmatrix} x_{n+1} \\ y_{n+1} \end{pmatrix}, \quad F = \begin{pmatrix} f(x, y) \\ g(x, y) \end{pmatrix}$$

Matrix and Vector



$$\hat{X}_{n+1} = -\frac{F(x_n)}{F'(x_n)} + \hat{X}_n, \text{ Iteration} \quad \hat{X}_n \rightarrow X_s$$

Multi Dim. NR uses Matrix

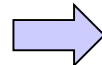
$$\hat{X}_{n+1} = \begin{pmatrix} x_{n+1} \\ y_{n+1} \end{pmatrix}, \quad F = \begin{pmatrix} f(x, y) \\ g(x, y) \end{pmatrix}$$

$$\hat{X}_{n+1} = -\frac{F(x_n)}{F'(x_n)} + \hat{X}_n$$

Matrix has NO DIVISION.

$$y = f'(x_n)(x - x_n) + f(x_n) = 0$$

$$f'(x_n)(x_{n+1} - x_n) + f(x_n) = 0$$



$$\therefore x_{n+1} = -\frac{f(x_n)}{f'(x_n)} + x_n$$

$$\hat{F}'(\hat{X}_{n+1} - \hat{X}_n) + \hat{F} = 0$$

$$\hat{F}'(\hat{X}_{n+1} - \hat{X}_n) = -\hat{F}$$

$$\hat{X}_{n+1} - \hat{X}_n = -(\hat{F}')^{-1} \hat{F}$$

$$\begin{aligned} \therefore \hat{X}_{n+1} &= \hat{X}_n - (\hat{F}')^{-1} \hat{F} \\ &= \hat{X}_n - J^{-1} \hat{F} \end{aligned}$$

Remind that Multi Dim. NR requires,

- Matrix calculation
- Differentiation \rightarrow Jacobian Matrix
- Division \rightarrow Inverse matrix

$J = \text{Jacobian Matrix}$

Another Derivation: NR Method for Non-Linear Equations

- Non linear Equations, $f(x,y)=0$, $g(x,y)=0$
 - Solve x,y
- Define $F(x,y)$

$$F(x, y) = \begin{pmatrix} f(x, y) \\ g(x, y) \end{pmatrix}$$

- Taylor series(or Differentiation)

$$F_i(\hat{x} + \hat{h}) \approx F_i(\hat{x}) + \sum_j \frac{\partial F_i}{\partial x_j} h_j = F_i(\hat{x}) + J_i h$$

$$\therefore F(\hat{x} + h) = F(\hat{x}) + J\hat{h}$$

- Remind $F(x,y)=0$

$$F(\hat{x} + h) = 0 = F(\hat{x}) + J\hat{h}$$

$$\therefore \hat{h} = -J^{-1}F \quad \rightarrow \hat{x}_{k+1} = \hat{x}_k + \hat{h} = \hat{x}_k - J^{-1}F$$

Example) nlnr.m

```

%NR example
x=3;
y=1;

J=[];

for i=1:1000
    f = x^2+y^2-10;
    g =x*y-5;
    J(1,1) = 2*x;
    J(1,2) = 2*y;
    J(2,1) = y;
    J(2,2) =x;

    Ji = inv(J);
    F = [f;g];
    h = [-Ji*F];

    X=[x;y];
    X=X+h;

    x=X(1);
    y=X(2);

    [x y]
    [f g];
    if (f^2+g^2<1e-7)
        break;
    end
    pause
end

```

- $X_0=3, y_0=1$

$$f = x^2 + y^2 - 10$$

$$g = xy - 5$$

$$J = \frac{\partial(f, g)}{\partial(x, y)} = \begin{bmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \\ \frac{\partial g}{\partial x} & \frac{\partial g}{\partial y} \end{bmatrix} = \begin{bmatrix} 2x & 2y \\ y & x \end{bmatrix}$$

$$\hat{h} = -J^{-1}\hat{F} = -J^{-1} \begin{bmatrix} f \\ g \end{bmatrix}$$

$$\hat{X} = \begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix} + \hat{h} = \begin{bmatrix} x \\ y \end{bmatrix} - J^{-1} \begin{bmatrix} f \\ g \end{bmatrix}$$

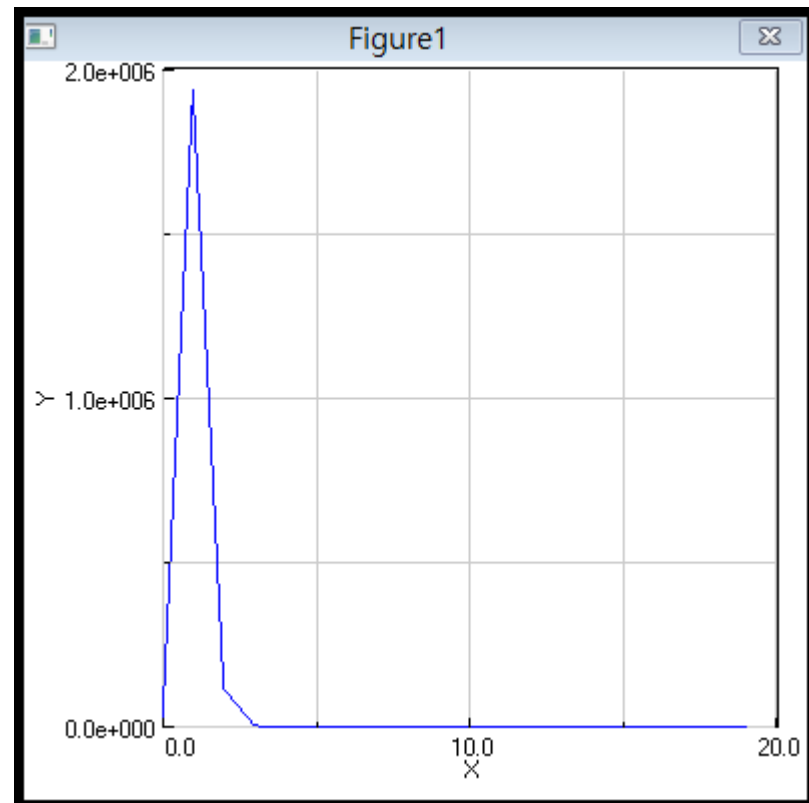


Non-Linear Newton-Raphson Method

- See l2nlr

```

l2nlr.test(0.1,0.1)
124.50048857879665
1937546.4477837086
119174.55138009507
6988.417253141524
339.8059228162356
8.228910930054738
0.02144254383316363
2.2397752679586273e-07
2.4751775419627393e-17
3.944304526105059e-30
3.944304526105059e-30
3.944304526105059e-30
3.944304526105059e-30
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3.944304526105059e-30
3.944304526105059e-30
3.944304526105059e-30
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3.944304526105059e-30
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3.944304526105059e-30
3.944304526105059e-30
3.944304526105059e-30
3.944304526105059e-30
result 2.23606797749979
2.23606797749979
  
```



Error becomes 3.9 e-30

HW 3

2DOF SCARA Robot Inverse Kinematics Solver with Nonlinear NR Method ex/ml/l2nrscara

- Forward Kinematics(See Robotics Lecture4, pp.20)

$$x = l_1 \cos(\theta_1) + l_2 \cos(\theta_1 + \theta_2) \quad L1 = L2 = 5$$

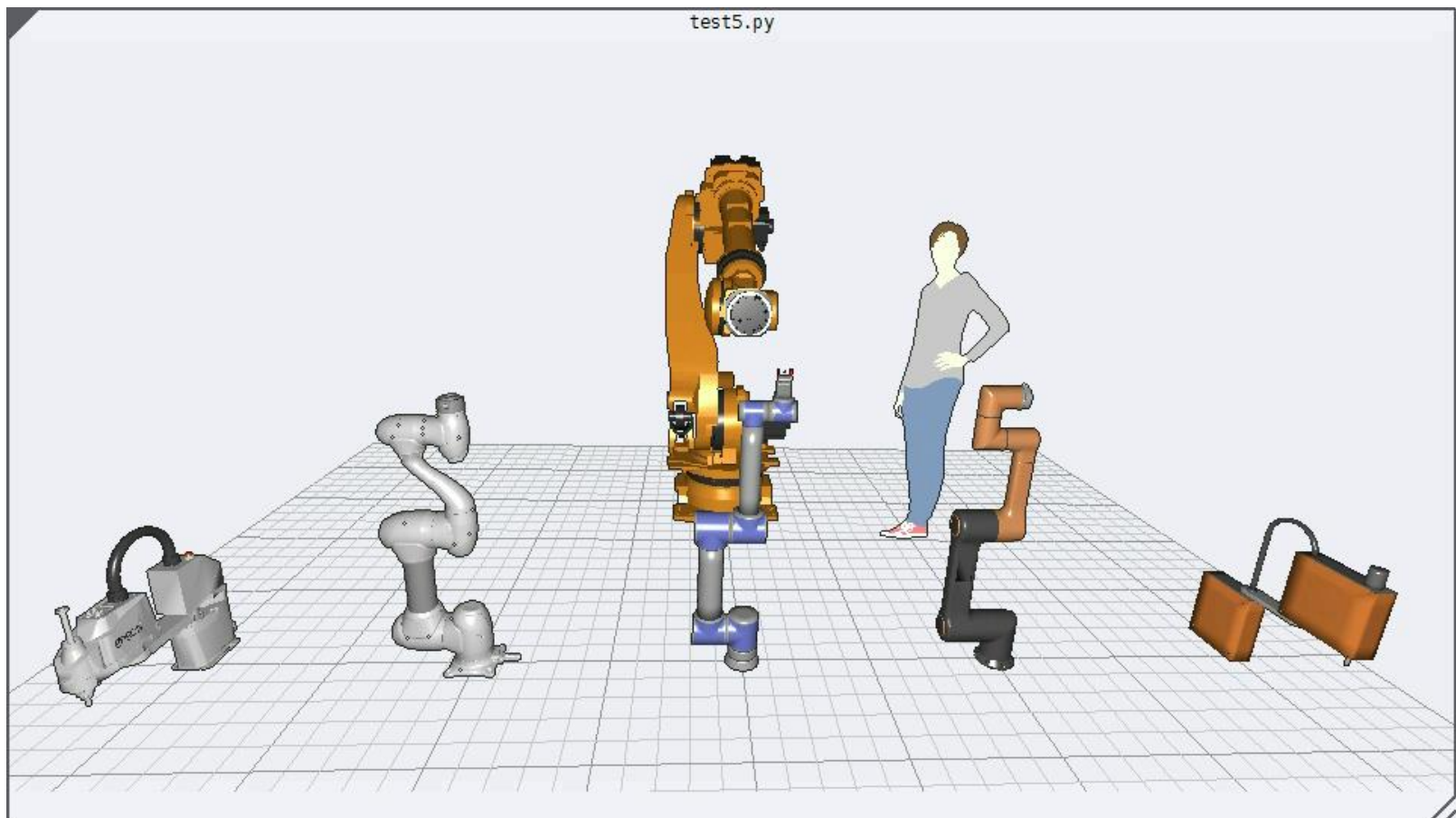
$$y = l_1 \sin(\theta_1) + l_2 \sin(\theta_1 + \theta_2)$$

- When x,y are given,
 - $X = 8.1603$
 - $Y = 5.7139$
- Question: What are q_1, q_2 ?
- Solve it with Nonlinear Newton Raphson Method

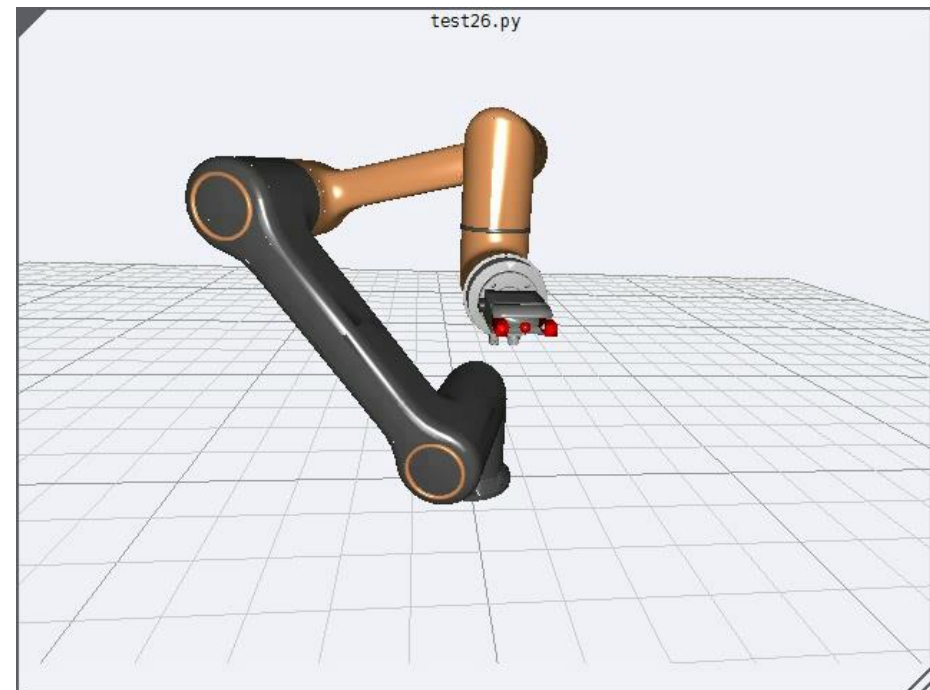
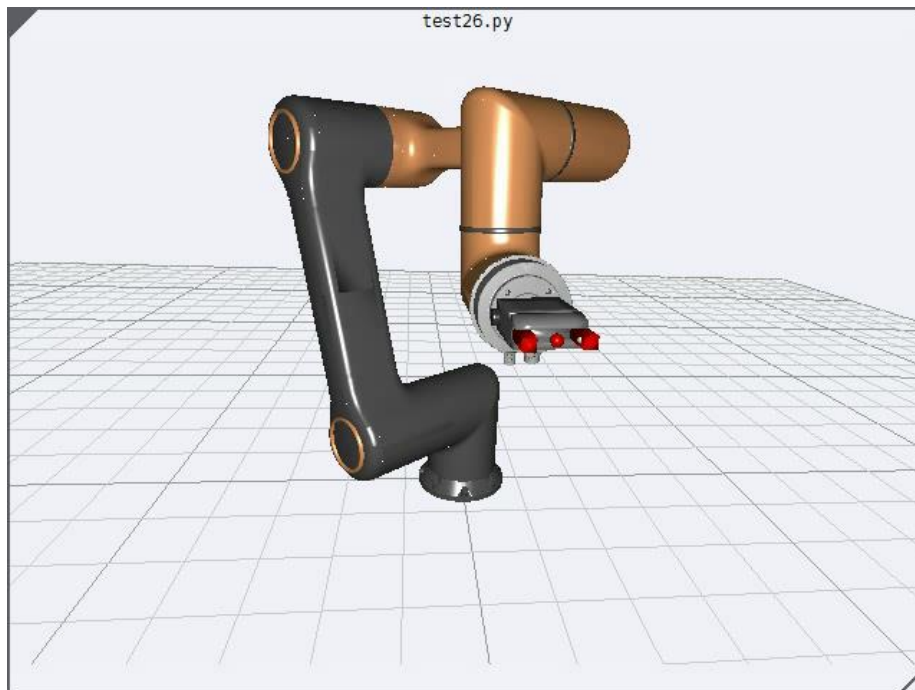


Inverse Kinematics in Robotics

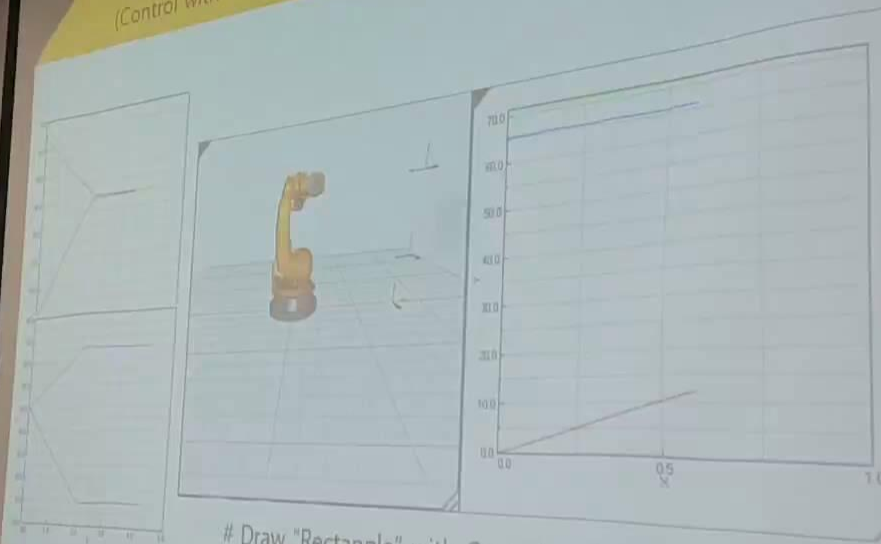
- In recent years, Most robots solve IK with Nonlinear Newtown-Raphson Method



Many Cooperative Robots do NOT have
Analytical Inverse Kinematics
Nonlinear Newton-Raphson Method
(or Inverse Jacobian Method)



Mission 4. Cartesian Space Control (Control with Jacobian Matrix)



Draw "Rectangle" with Cartesian Space Control