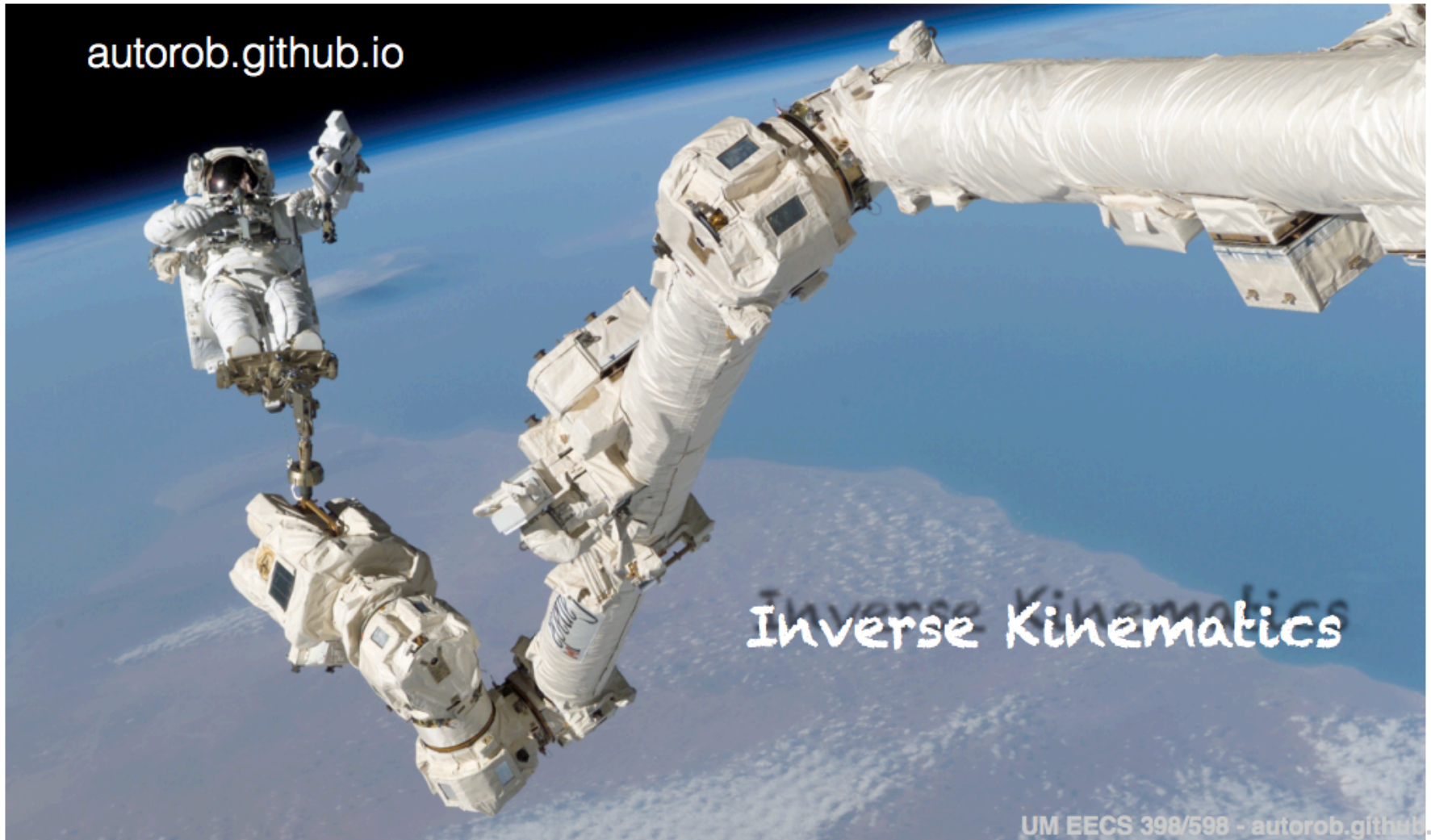
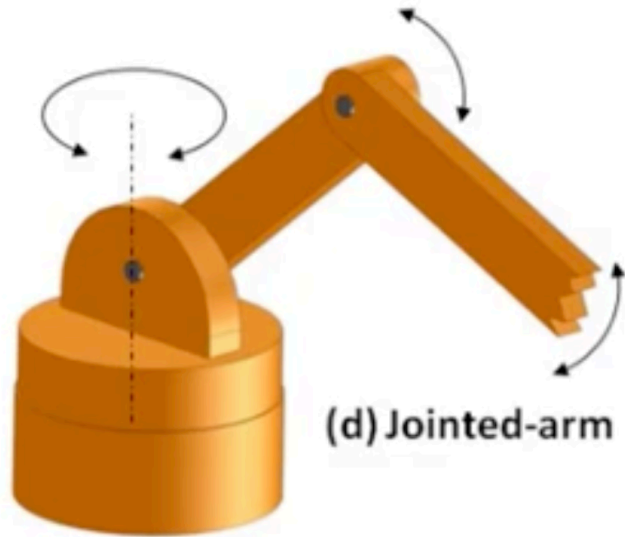


# Nonlinear Equations

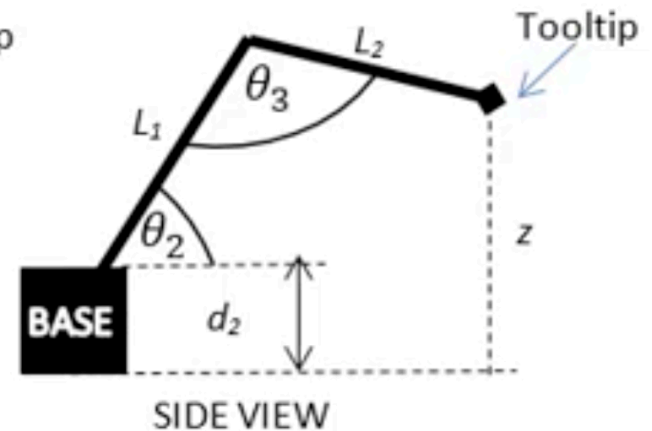
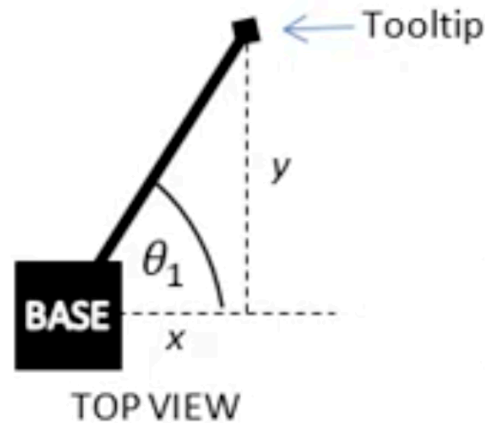
# Nonlinear system of equations



# Robotic arms



(d) Jointed-arm

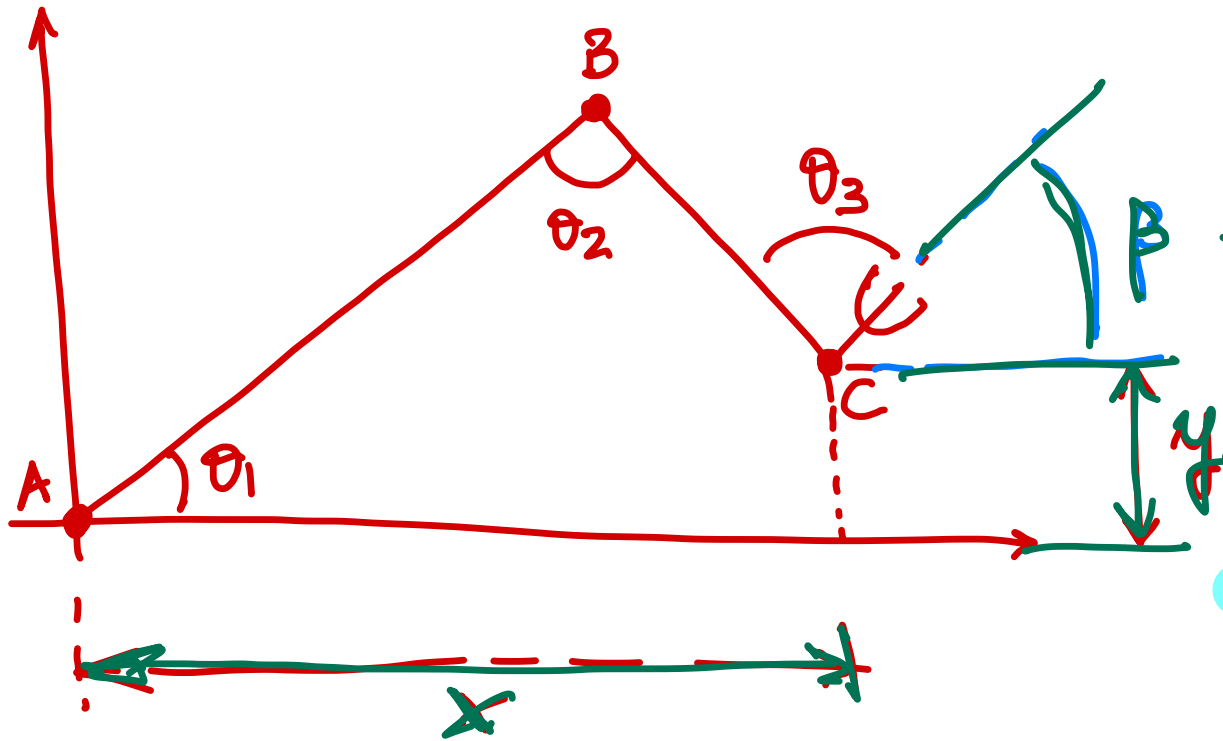


<https://www.youtube.com/watch?v=NRgNDIVtmz0> (Robotic arm 1)

<https://www.youtube.com/watch?v=9DqRkLQ5Sv8> (Robotic arm 2)

[https://www.youtube.com/watch?v=DZ\\_oemY8xEI](https://www.youtube.com/watch?v=DZ_oemY8xEI) (Blender)

# Inverse Kinematics



Given :

$x, y, \beta$

Find :

$\theta_1, \theta_2, \theta_3$

$$f(\theta) = 0$$

$$f_1(\theta_1, \theta_2, \theta_3) = 0$$

$$f_2(\theta_1, \theta_2, \theta_3) = 0$$

$$f_3(\theta_1, \theta_2, \theta_3) = 0$$

3 eq.

+

3 unknowns

solve

iterative

# Nonlinear system of equations

**Goal:** Solve  $\underline{f}(\underline{x}) = \underline{0}$  for  $\underline{f}: \mathcal{R}^n \rightarrow \mathcal{R}^n$

$$\underline{f}(\underline{x}) = \underline{0}$$

$$\underline{f}(\underline{x}) = \begin{bmatrix} f_1(x_1, x_2, x_3, \dots, x_n) \\ f_2(x_1, x_2, x_3, \dots, x_n) \\ \vdots \\ f_n(x_1, x_2, x_3, \dots, x_n) \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

Suppose

$$\begin{cases} x_1^2 + 2x_1x_2 + x_2^2 = 4 \\ 2x_1 + 3x_2 = 5 \end{cases}$$

$$\begin{aligned} f_1 &= x_1^2 + 2x_1x_2 + x_2^2 - 4 = 0 \\ \implies f_2 &= -2x_1 - 3x_2 + 5 = 0 \end{aligned}$$

# Newton's method (ND)

Approximate the nonlinear function  $f(x)$  by a linear function using Taylor expansion:

$$\underbrace{f(\tilde{x} + \tilde{s})}_{\text{vector}} \approx \underbrace{f(\tilde{x})}_{\text{vector}} + \underbrace{J(\tilde{x})}_{\text{matrix}} \underbrace{\tilde{s}}_{\text{vector}}$$

nonlinear

linear approximation

$$\tilde{f} = \begin{bmatrix} f_1(x_1, \dots, x_n) \\ f_2 \\ \vdots \\ f_n(x_1, \dots, x_n) \end{bmatrix}$$

Jacobian

$$= \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}_{n \times n}$$

$$J_{ij} = \frac{\partial f_i}{\partial x_j}$$

$$\underline{f}(\underline{x} + \underline{s}) \approx \underline{f}(\underline{x}) + \underline{J}(\underline{x}) \underline{s}$$

$\parallel \qquad \qquad \parallel$   
 $0 \qquad \qquad 0$

$$\underline{f}(\underline{x}) + \underline{J}(\underline{x}) \underline{s} = 0$$

$$\underline{J}(\underline{x}) \underline{s} = -\underline{f}(\underline{x})$$

→ solve for  $\underline{s}$   
Linear system of equation

$\underline{x}_0 =$  initial vector

$$\underline{x}_{k+1} = \underline{x}_k + \underline{s}$$

↗

# Newton's method

$$f(\underline{x}) = 0$$

## Algorithm:

$\underline{x}_0$ : initial guess

for  $i = 1, 2, \dots$

evaluate  $J(\underline{x}_k) = J$

evaluate  $f(\underline{x}_k) = f$

solve  $J\underline{s} = -f \rightarrow$  Find  $\underline{s}$

update  $\underline{x}_{k+1} = \underline{x}_k + \underline{s}$

$O(n^3)$

iterative  
process  
sequence of  
linear system  
of equation

## Convergence:

- Typically has quadratic convergence
- Drawback: Still only locally convergent

## Cost:

- Main cost associated with computing the Jacobian matrix and solving the Newton step.

$n^3$

$n^2$



# Example

Consider solving the nonlinear system of equations

$$\begin{cases} 2 = 2y + x \\ 4 = x^2 + 4y^2 \end{cases}$$

$$\underline{f} = \begin{bmatrix} 2y + x - 2 \\ x^2 + 4y^2 - 4 \end{bmatrix}$$

What is the result of applying one iteration of Newton's method with the following initial guess?

$$\underline{x}_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\underline{J} = \begin{bmatrix} 1 & 2 \\ 2x & 8y \end{bmatrix}$$

$$\underline{x}_1 = \underline{x}_0 + \underline{s} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \begin{bmatrix} s_1 \\ s_2 \end{bmatrix}$$

$$\underline{J}(\underline{x}_0), \underline{f}(\underline{x}_0)$$

$$\underline{f}(\underline{x}_0) = \begin{bmatrix} -1 \\ -3 \end{bmatrix}$$

$$\underline{J}(\underline{x}_0) = \begin{bmatrix} 1 & 2 \\ 2 & 0 \end{bmatrix}$$

$$\underline{x}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \begin{bmatrix} 1.5 \\ -0.25 \end{bmatrix} = \begin{bmatrix} 2.5 \\ -0.25 \end{bmatrix}$$

$$\underline{J}\underline{s} = -\underline{f}$$

$$\begin{bmatrix} 1 & 2 \\ 2 & 0 \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$s_1 + 2s_2 = 1$$

$$2s_1 = 3 \rightarrow s_1 = 1.5$$

$$2s_2 = 1 - 1.5 = -0.5$$

$$s_2 = -0.25$$

# Newton's method

$\mathbf{x}_0 = \text{initial guess}$

For  $k = 1, 2, \dots$

Evaluate  $\mathbf{J} = J(\mathbf{x}_k)$

Evaluate  $\mathbf{f}(\mathbf{x}_k)$

to solve { Factorization of Jacobian (for example  $\mathbf{LU} = \mathbf{J}$ )

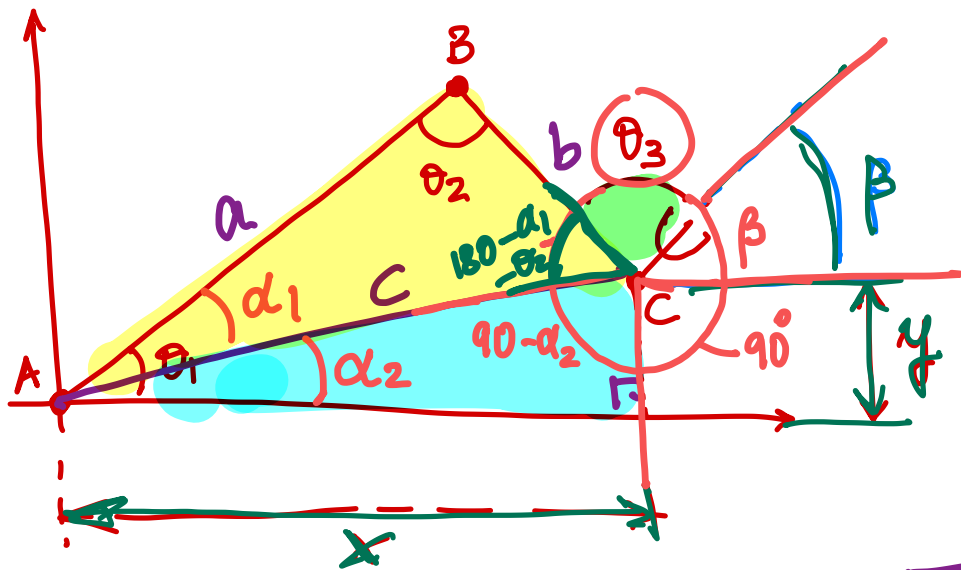
{ Solve using factorized J (for example  $\mathbf{LU} \mathbf{s}_k = -\mathbf{f}(\mathbf{x}_k)$ )

Update  $\mathbf{x}_{k+1} = \mathbf{x}_k + \mathbf{s}_k$

# Newton's method - summary

- ❑ Typically quadratic convergence (local convergence)
- ❑ Computing the Jacobian matrix requires the equivalent of  $n^2$  function evaluations for a dense problem (where every function of  $\mathbf{f}(\mathbf{x})$  depends on every component of  $\mathbf{x}$ ).
- ❑ Computation of the Jacobian may be cheaper if the matrix is sparse.
- ❑ The cost of calculating the step  $\mathbf{s}$  is  $O(n^3)$  for a dense Jacobian matrix (Factorization + Solve)
- ❑ If the same Jacobian matrix  $\mathbf{J}(\mathbf{x}_k)$  is reused for several consecutive iterations, the convergence rate will suffer accordingly (trade-off between cost per iteration and number of iterations needed for convergence)

# Inverse Kinematics



$$x, y, \beta \rightarrow \theta_1, \theta_2, \theta_3$$

$$c = \sqrt{x^2 + y^2}$$

$a, b$  given

$$\Rightarrow d_1 = \theta_1 - d_2$$

$$\theta_1 = d_1 + d_2$$

$$d_2 = \tan^{-1}(y/x)$$

$$c^2 = a^2 + b^2 - 2ab \cos \theta_2$$

$$f_1 = c^2 - a^2 - b^2 + 2ab \cos \theta_2 = 0$$

$$b^2 = a^2 + c^2 - 2ac \cos \alpha_1$$

$$f_2 = b^2 - a^2 - c^2 + 2ac \cos(\theta_1 - d_2) = 0$$

$$(\cancel{180 - d_1 - \theta_2}) + \theta_3 + \beta + \cancel{90} + (\cancel{90 - d_2}) = \cancel{360}$$

$$-d_1 - \theta_2 + \theta_3 + \beta - d_2 = 0$$

$$-(\cancel{\theta_1 - d_2}) - \theta_2 + \theta_3 + \beta - \cancel{d_2} = 0$$

$$f_3 = -\theta_1 - \theta_2 + \theta_3 + \beta = 0$$