



Centre for  
Robotics



# EGB339 Part 2: Robotic Arms

## Lecture 5: Path and Trajectory Planning

Chris Lehnert (Lecturer)

# Outline

- Subjects **covered** in this series of lectures
  - Rigid Body Motions (week 8)
  - Forward Kinematics (week 9)
  - Inverse Kinematics (week 10)
  - Velocity Kinematics (week 11)
  - **Path and Trajectory Planning (week 12)**
  - Revision (week 13)
- Subjects **not covered** in this series of lectures
  - Dynamics
  - Control
  - Hardware
  - (Artificial) Intelligence
  - ...

# Watch these online videos

- QUT Robot Academy (by Prof Peter Corke)
  - Paths and Trajectories
    - <https://robotacademy.net.au/masterclass/paths-and-trajectories/>

# Review of Week 11

- The **Jacobian** is a **matrix** that is a function of joint position, that linearly relates **joint velocity** to **toolpoint velocity**.
- The **Forward Velocity Kinematics** maps the velocity of the joints to the velocity of the tool.
- The **Inverse Velocity Kinematics** maps the velocity of the tool to the velocity of the joints.
- From the viewpoint of velocity kinematics, **singularity** means a **configuration** of the robot in which the Jacobian matrix becomes **rank-deficient**.
- The **static force/torque** analysis can be done with the **Jacobian** matrix.

# Common Questions

- Rank-deficient?
  - Rank of a matrix is the **dimension** of the **largest** submatrix that is invertible.

$$\text{Rank}\left(\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}\right)=3 \quad \text{Rank}\left(\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}\right)=2$$

$$\text{Rank}\left(\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}\right)=1 \quad \text{Rank}\left(\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}\right)=0$$



# Common Questions

- What if Jacobian is non-square?
  - Forward velocity kinematics is not affected
  - For inverse velocity kinematics, two cases:
    - $J_{m \times n}$ ,  $m < n$ , could be **infinite solutions**
      - May use pseudo-inverse to calculate a solution
        - Pseudo-inverse:  $J^{\dagger} = J^T (JJ^T)^{-1}$
        - $\dot{q} = J^{\dagger} \dot{p}$
    - $J_{m \times n}$ ,  $m > n$ , could be **no exact solutions**
      - May use pseudo-inverse to calculate an approximate solution
        - Pseudo-inverse:  $J^{\dagger} = (J^T J)^{-1} J^T$
        - $\dot{q} = J^{\dagger} \dot{p}$

# Common Questions

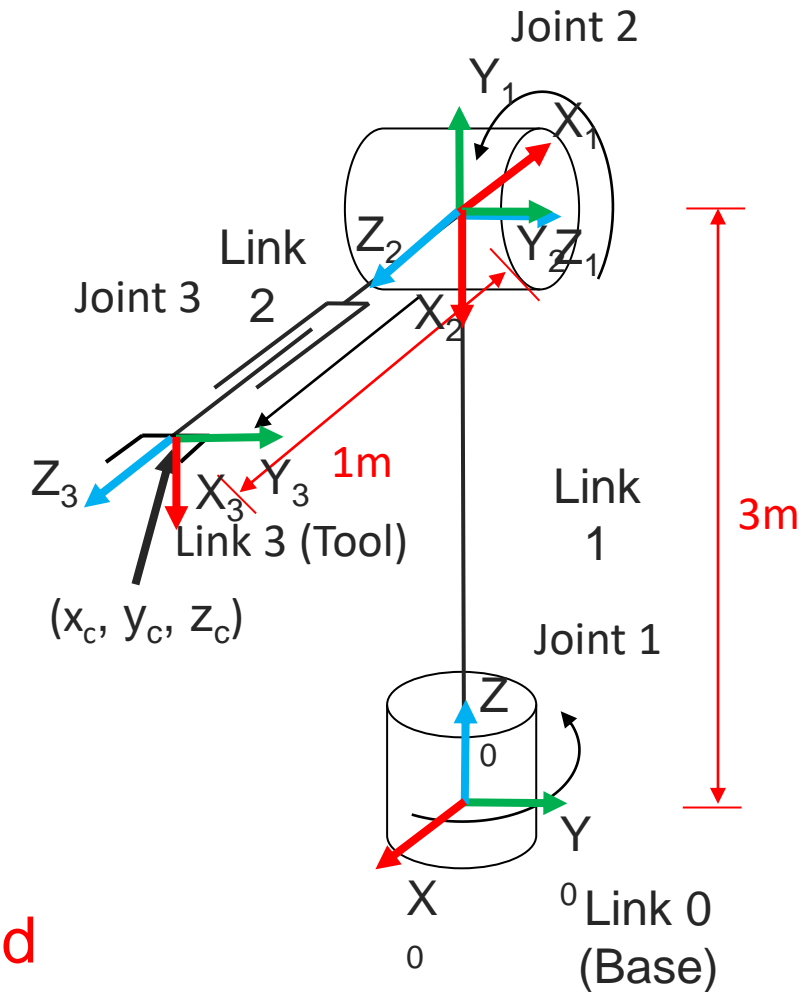
- Find the force/torque **required** at the joints to support a 3 kg (30 N) load at the tool. ( $q_1=0, q_2=0, q_3=0$ ) i.e., ( $\vartheta_1=0, \vartheta_2=-90, d_3=1$ )

$$J_v = \begin{bmatrix} d_3 s_1 s_2 & -d_3 c_1 c_2 & -c_1 s_2 \\ -d_3 c_1 s_2 & -d_3 s_1 c_2 & -s_1 s_2 \\ 0 & d_3 s_2 & -c_2 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & -1 & 0 \end{bmatrix}$$

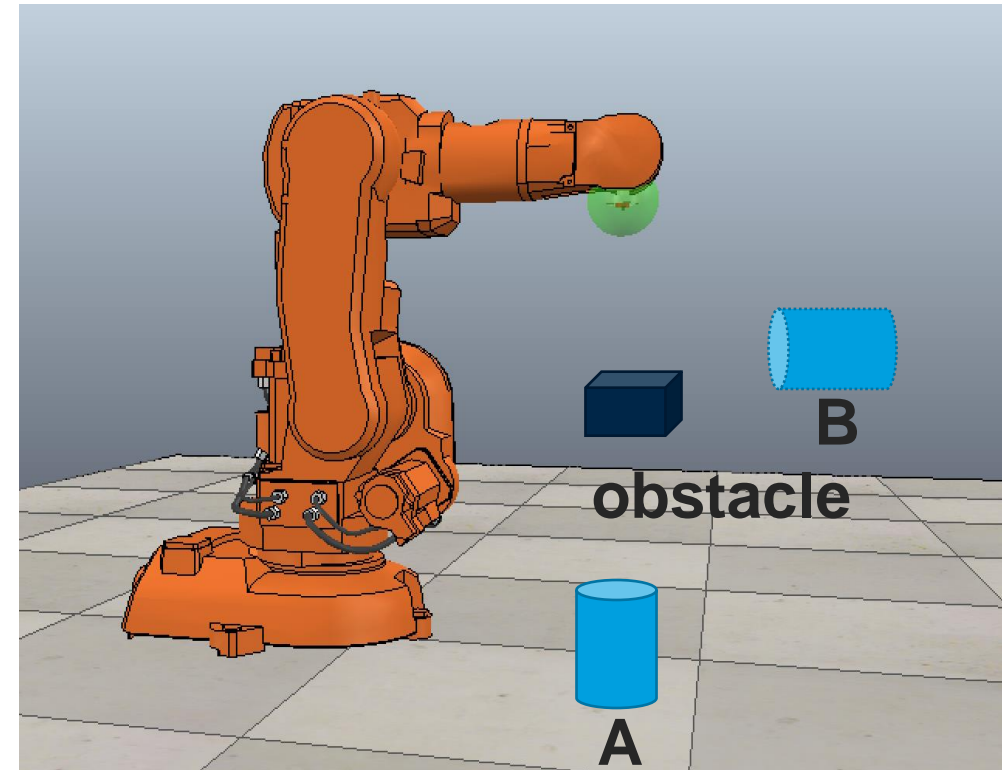
$$\tau = J_v^T F = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & -1 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ -30 \end{bmatrix} = \begin{bmatrix} 0 \\ 30 \\ 0 \end{bmatrix}$$

Joint 2 experiences 30 Nm due to the load, and requires -30 Nm to support the load.



# Motivating Problem

- Imagine one of your arms is replaced by a robotic arm. You are supposed to move an object from A to B.
- Now you know where the object is in front of you (homogeneous transformation).
- You know where your “hand” is with respect to your “body” (forward kinematics).
- You know how to move your “hand” to reach the object (inverse kinematics) at a certain speed (velocity kinematics).
- Can you find a path to move the object while avoiding the obstacle?





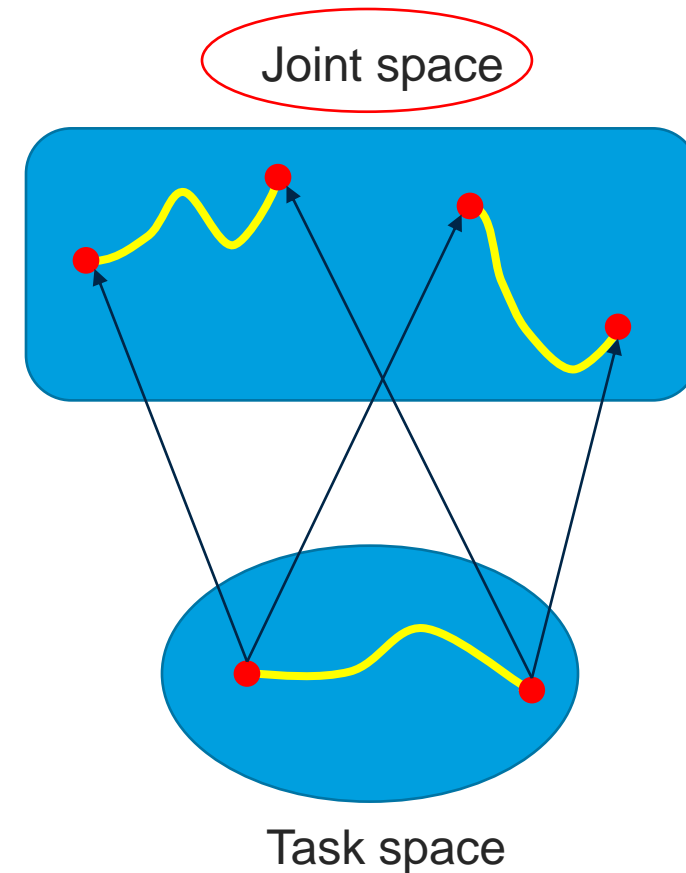
# Paths and trajectories



<https://robotacademy.net.au/masterclass/paths-and-trajectories/?lesson=109>

# Path Planning

- Path is a curve connecting a **starting** configuration  $q_0$  to a **goal** configuration  $q_1$ .
- **One** path in the **joint space** corresponds to **one** path in the **task space**.
- **One** path in the **task space** could correspond to **multiple** paths in the **joint space**.
  - Because of multiple inverse kinematics solutions
- Path planning is to find a valid **collision free** path in the **joint space** to connect the starting configuration to the goal configuration.

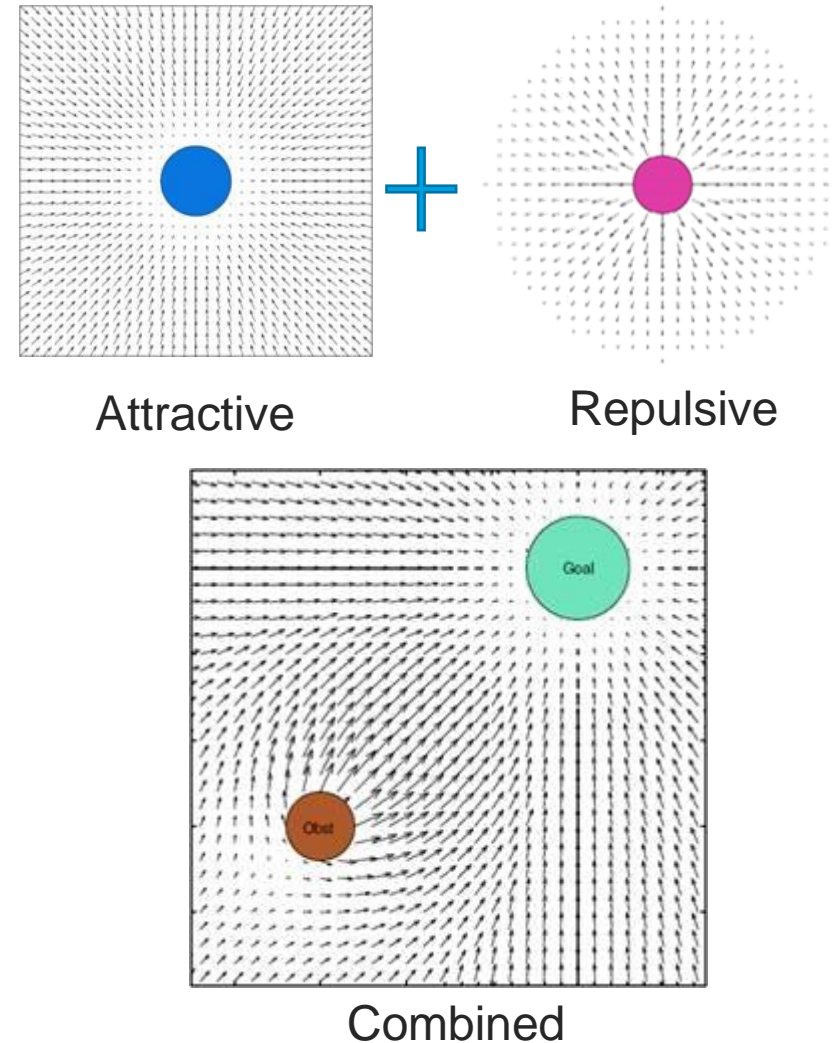


# Methods for Path Planning

- Artificial Potential Field
- Sampling Based Planning
- Grid Based Planning
- Reward Based Planning

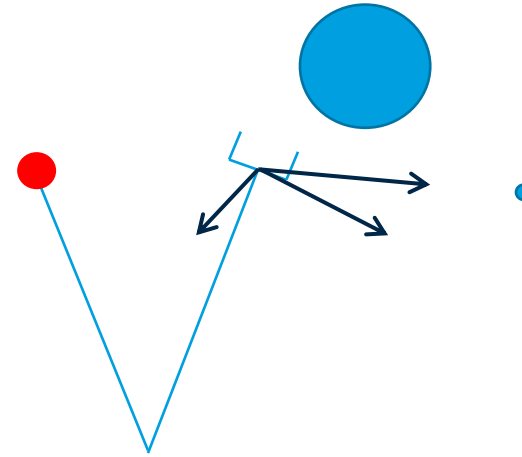
# Artificial Potential Field

- Treat the goal as an attractive potential field.
- Treat obstacles as repulsive potential field.
- Sum potential fields
- Follow the force (direction and magnitude) of the combined potential field
  - Should lead to the goal while avoiding obstacles



# Artificial Potential Field

- Treat the goal as an attractive virtual force.
- Treat obstacles as repulsive virtual forces.
- Sum virtual forces on toolpoint.
- Use  $J^T$  to change virtual force on toolpoint to virtual torques at joints. Remember  $\tau = J^T F$ .
- Follow virtual torques in joint space to goal while avoiding obstacles.
  - Can use gradient descent method to perform this iteratively.
- **YOU DO NOT NEED TO DO ANY OF THIS TO MAKE A PERFECT PRAC ROBOT. But it may help you think about it.**



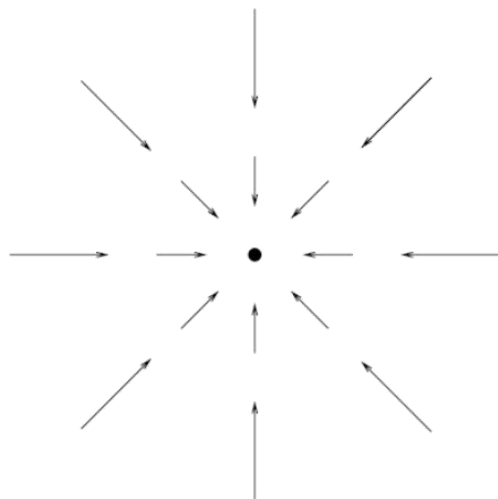


# Attractive Potential Functions

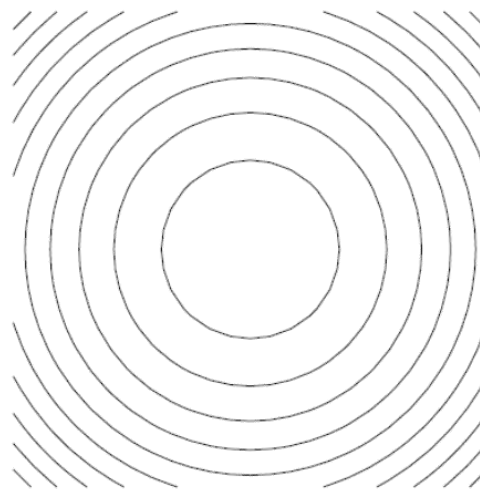
- Conical
- Quadratic

$$U(q) = \zeta d(q, q_{goal})$$

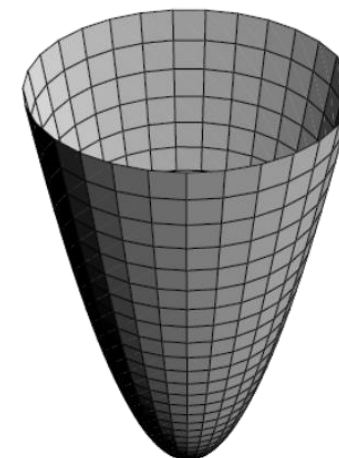
$$U(q) = \frac{1}{2} \zeta d^2(q, q_{goal})$$



(a)



(b)

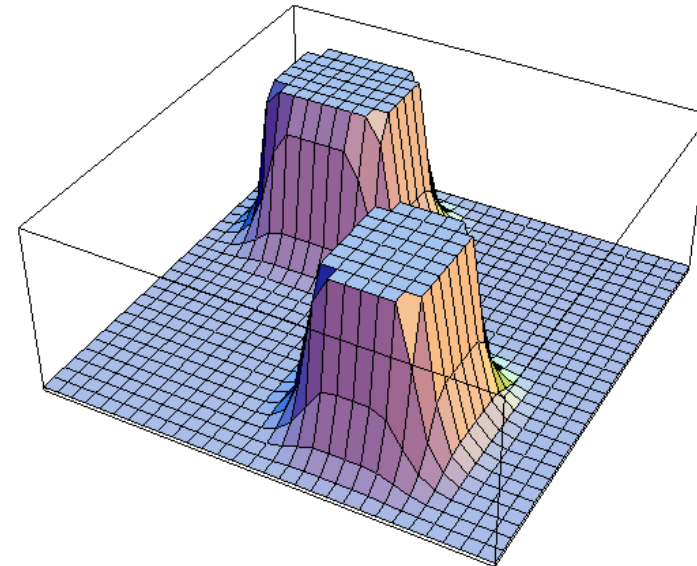
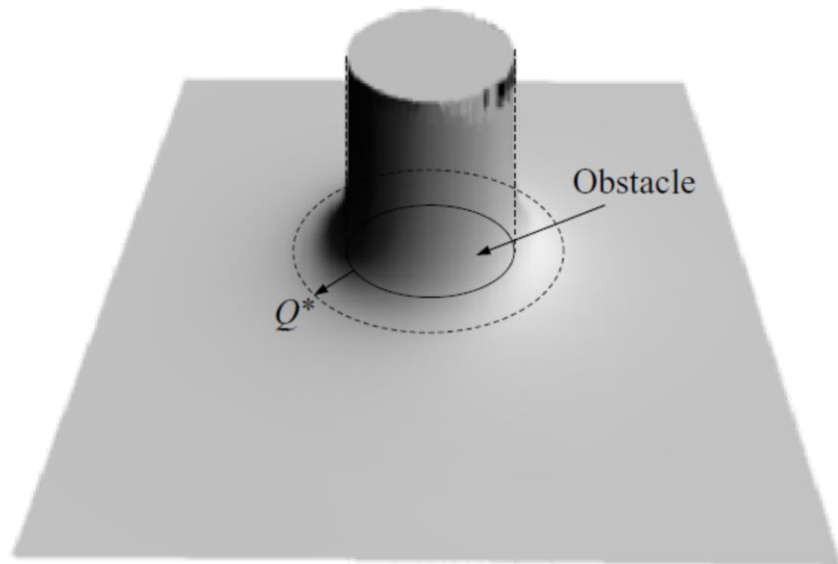


(c)

$\zeta = \text{scalar}$ ,  $d(*) = \text{distance function}$ ,  $q = \text{robot position}$

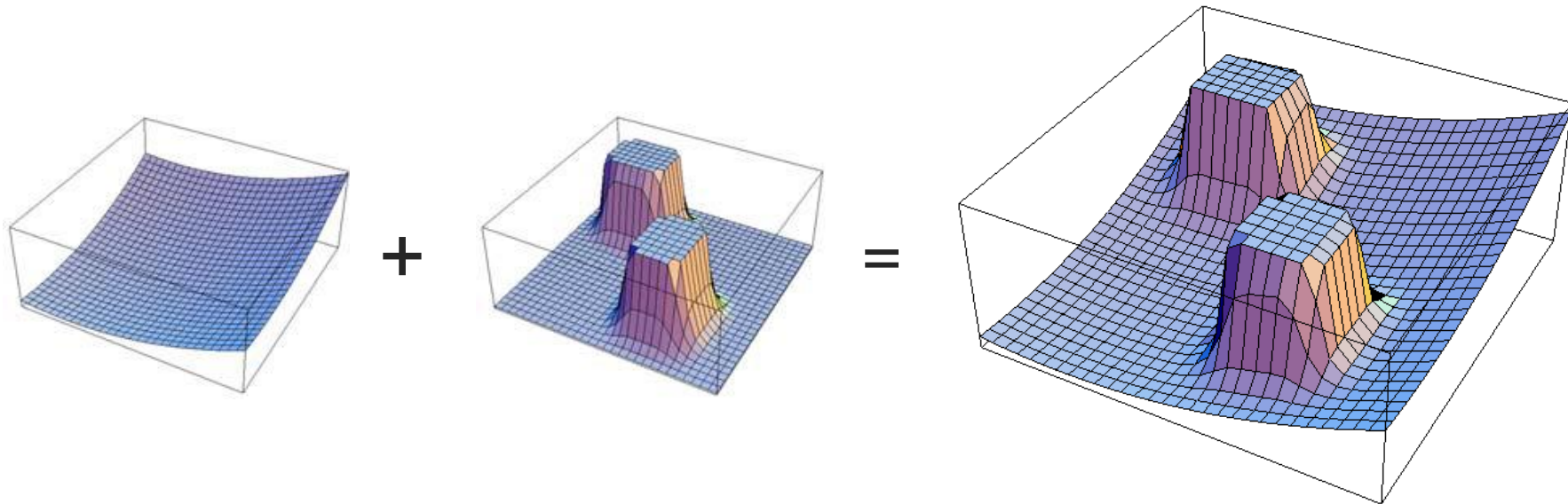
# Repulsive Potential Functions

$$U(q) = \begin{cases} \frac{1}{2} \eta \left( \frac{1}{d(q)} - \frac{1}{Q^*} \right)^2, & d(q) \leq Q^* \\ 0, & d(q) > Q^* \end{cases}$$



$\eta = \text{scalar}, Q^* = \text{scalar obstacle buffer}, d(*) = \text{distance function}$

# Total Potential Field



# Computing Virtual Forces – Gradient of Field

- Attractive field for goal
  - Force directed towards goal, magnitude proportional to distance.
  - For a goal at  $p$ :  $F_{att}(q) = -\zeta(o_i(q) - p)$
- Repulsive field for obstacles
  - Force directed away from obstacle, magnitude proportional to inverse square of distance
  - For an obstacle at  $b$ :

$$F_{rep}(q) = \eta \frac{(o_i(q) - b)}{\|o_i(q) - b\|^3}$$

Location of toolpoint (which is a function of the joint values  $q$ )

Location of goal

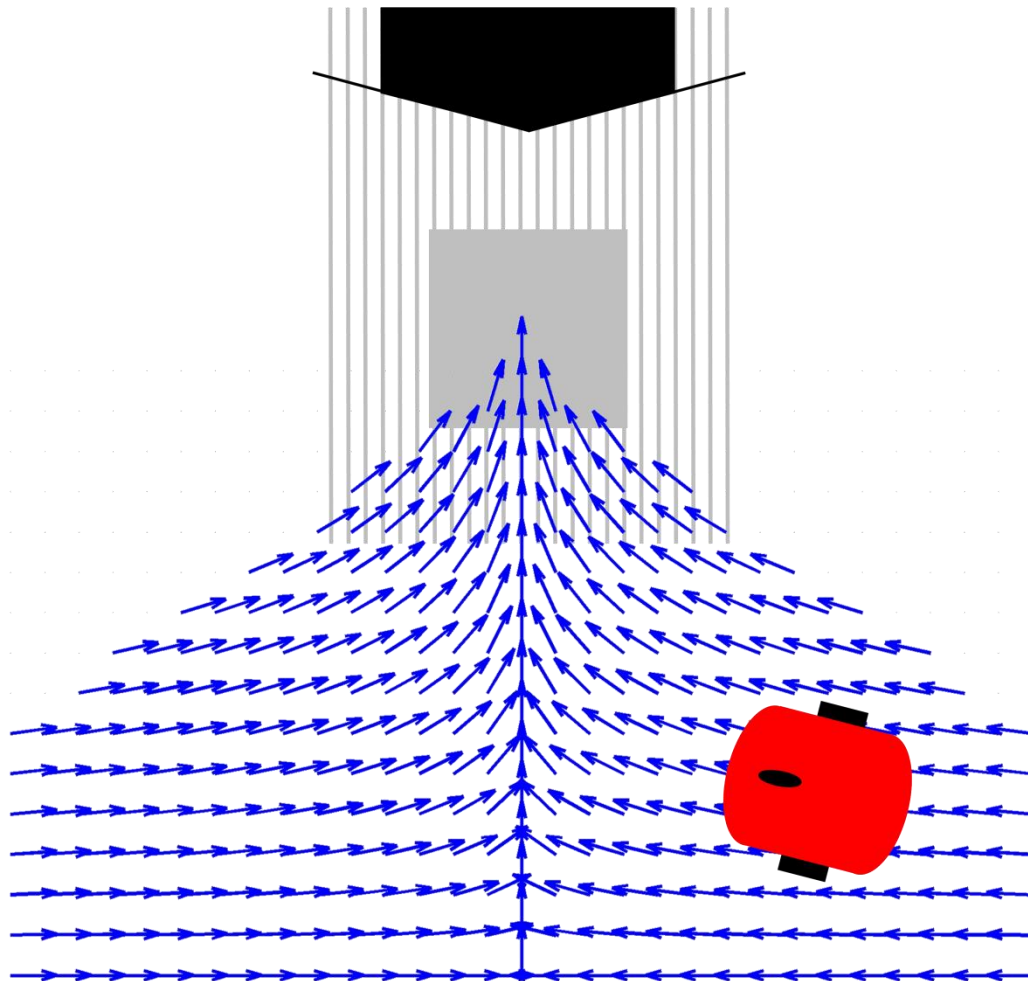
Location of obstacle

Double vertical lines are the “norm” – take the square root of the sum of the squares of each vector element

$$\|a\| = \sqrt{a_1^2 + a_2^2 + \dots + a_n^2}$$



# Artificial Potential Field in Practice





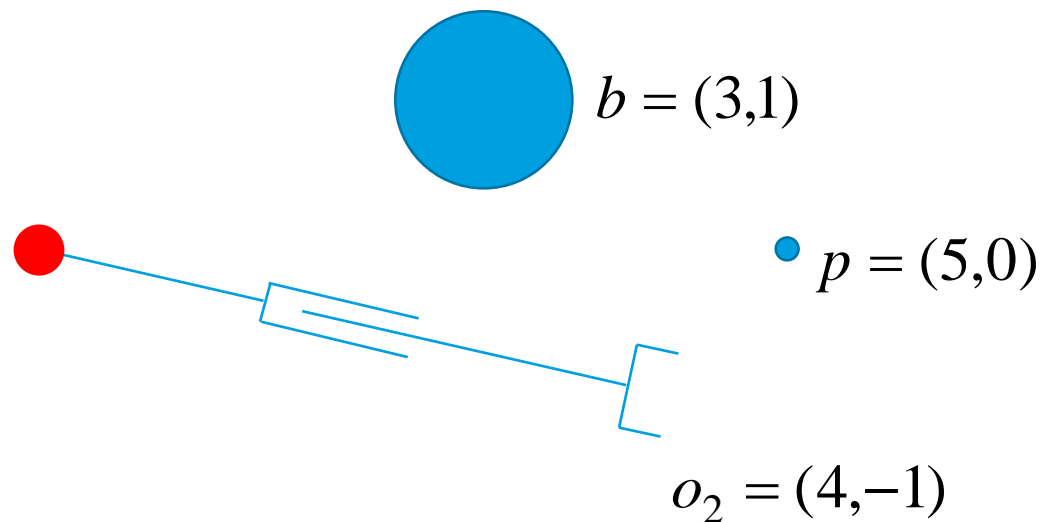


# Artificial Potential Field Parameters

- Need to set **goal and obstacle gains**  $\zeta$  and  $\eta$ .
  - **Ratio** of gains is more important than absolute values.
- Set up **gradient descent step size**  $\alpha$  and **time step**  $\Delta t$  in motion generation algorithm.
  - Choose a value that moves toolpoint at a safe speed.
  - Large steps can hit obstacles, but reduce compute time.
  - Use cubic polynomial between points – will practice in the tutorial.

# Artificial Potential Field - Example

- Calculate the net potential field force when the robot is at (4, -1).  
Use  $\zeta = 1$  and  $\eta = 3$ .

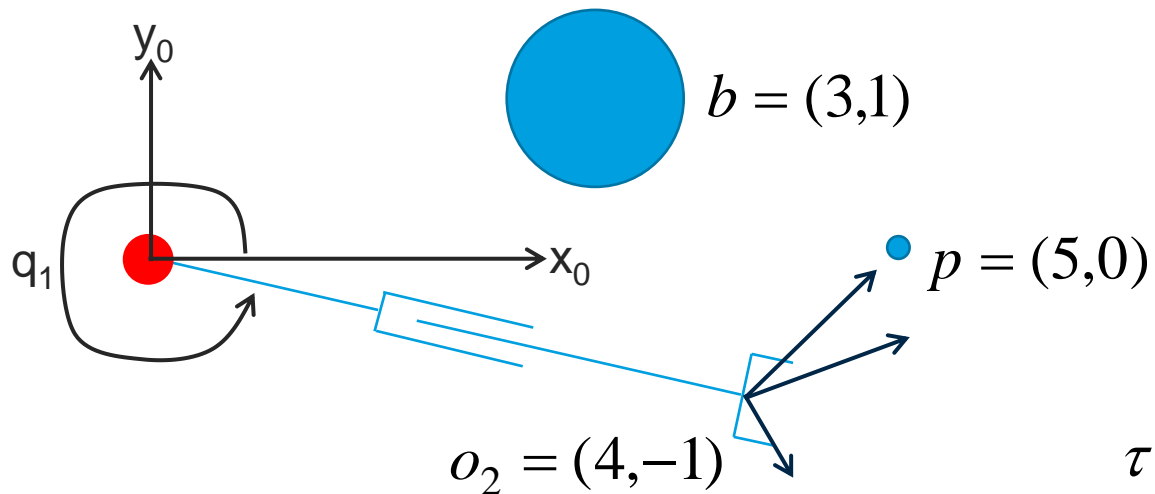


$$F_{att}(q) = -\zeta(o_i(q) - p)$$

$$F_{rep}(q) = \eta \frac{(o_i(q) - b)}{\|o_i(q) - b\|^3}$$

# Artificial Potential Field - Example

- Now convert the force to joint torques, given:



$$F_{att}(q) = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

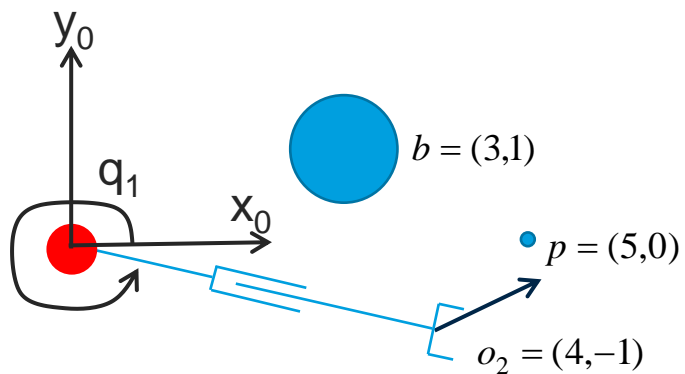
$$F_{rep}(q) = \begin{bmatrix} 0.26 \\ -0.52 \end{bmatrix}$$

$$F(q) = \begin{bmatrix} 1.26 \\ 0.48 \end{bmatrix}$$

$$\tau(q) = J^T F(q)$$

$$J = \begin{bmatrix} -q_2 s_1 & c_1 \\ q_2 c_1 & s_1 \end{bmatrix}$$

# Artificial Potential Field - Example



$$\tau(q) = J^T F(q)$$

$$J = \begin{bmatrix} -q_2 s_1 & c_1 \\ q_2 c_1 & s_1 \end{bmatrix}$$

$$F(q) = \begin{bmatrix} 1.26 \\ 0.48 \end{bmatrix}$$

$$q_1 = \text{atan2}(4, -1) = -0.245 \text{ rad}$$

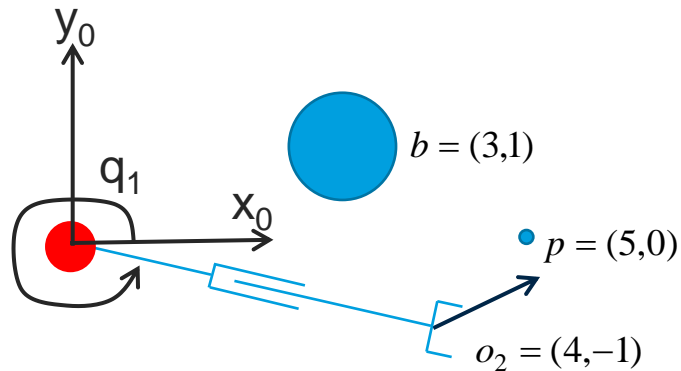
$$q_2 = \sqrt{16 + 1} = 4.12 \text{ m}$$

$$J = \begin{bmatrix} 1 & 0.970 \\ 4 & -0.243 \end{bmatrix}$$

$$\tau = \begin{bmatrix} 1 & 4 \\ 0.970 & -0.243 \end{bmatrix} \begin{bmatrix} 1.26 \\ 0.48 \end{bmatrix} = \begin{bmatrix} 3.18 \\ 1.11 \end{bmatrix}$$



# Artificial Potential Field - Example



$$q_1 = \text{atan2}(4, -1) = -0.245$$

$$q_2 = \sqrt{16+1} = 4.12$$

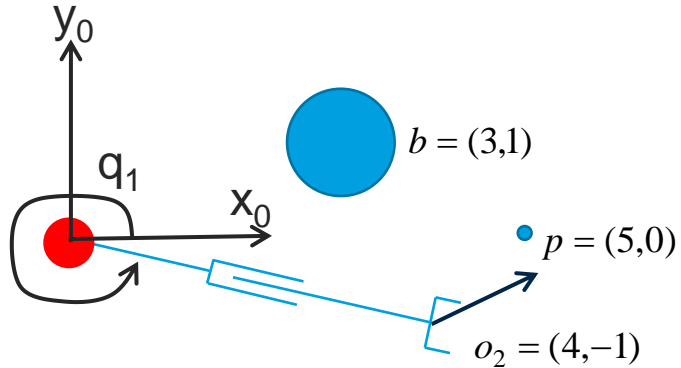
$$\tau = \begin{bmatrix} 1 & 4 \\ 0.970 & -0.243 \end{bmatrix} \begin{bmatrix} 1.26 \\ 0.48 \end{bmatrix} = \begin{bmatrix} 3.18 \\ 1.11 \end{bmatrix}$$

- Using a **gradient descent step size of  $\alpha = 0.1$**  and a **time step of  $\Delta t = 0.1s$** , find the new joint positions and velocities.

$$q(t + \Delta t) = q(t) + \alpha \frac{\tau(q)}{\|\tau(q)\|}$$

$$\dot{q}(t + \Delta t) = \frac{q(t + \Delta t) - q(t)}{\Delta t} = \frac{\alpha}{\Delta t} \frac{\tau(q)}{\|\tau(q)\|}$$

# Artificial Potential Field - Example



$$q_1 = \text{atan2}(4, -1) = -0.245$$

$$q_2 = \sqrt{16+1} = 4.12$$

$$\tau = \begin{bmatrix} 3.18 \\ 1.11 \end{bmatrix}$$

$$\alpha = 0.1$$

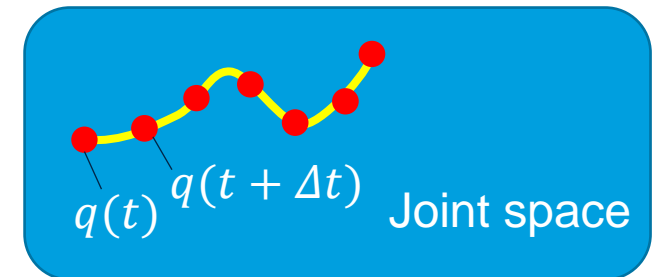
$$\Delta t = 0.1s$$

$$q(t + \Delta t) = q(t) + \alpha \frac{\tau(q)}{\|\tau(q)\|}$$

$$\begin{bmatrix} q_1 \\ q_2 \end{bmatrix} = \begin{bmatrix} -0.245 \\ 4.12 \end{bmatrix} + \frac{0.1}{3.36} \begin{bmatrix} 3.18 \\ 1.11 \end{bmatrix} = \begin{bmatrix} -0.15 \\ 4.15 \end{bmatrix}$$

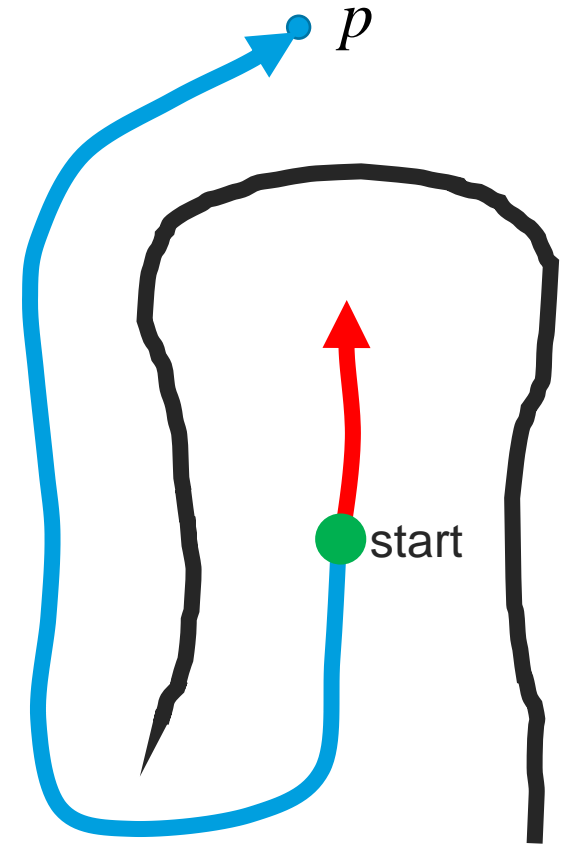
$$\dot{q}(t + \Delta t) = \frac{\alpha}{\Delta t} \frac{\tau(q)}{\|\tau(q)\|} = \begin{bmatrix} 0.946 \\ 0.330 \end{bmatrix}$$

- Using the current and next positions and velocities, you could go on to construct a **cubic polynomial trajectory**.



# Artificial Potential Field Features

- Useful for planning paths around obstacles.
  - Works with dynamic obstacles too.
- Won't land precisely at goal.
  - Use a simpler trajectory planner once close to goal.
- Can get stuck in **local minima**.
  - Places where attractive and repulsive forces cancel to zero.
  - Basins where obstacles “herd” toolpoint to centre.
  - Classic rookie mistake



# Trajectory

- A **path** and a **schedule** for getting from A to B
  - ➔ there is a notion of time or speed



# Desirable Properties of a Trajectory

- Spatial Accuracy
  - Arrives at the desired location (welding)
- Temporal Accuracy
  - Arrives at the right time (picking up from conveyor belt)
- Temporal Efficiency
  - Doesn't waste time getting to the desired location (draglines)
- Smoothness
  - A continuous function with a finite first derivative (duty cycle)





# Different Trajectory Types

- Cubic Polynomial Trajectories
- Quintic Polynomial Trajectories
- Linear Segments with Parabolic Blends (LSPB)
- Minimum Time Trajectories (Bang-Bang)

# Cubic Polynomial

- Describing (joint) position as a cubic polynomial gives **spatial** and **temporal** accuracy, and **smoothness**.

$$q(t) = a_0 + a_1t + a_2t^2 + a_3t^3$$

$$\dot{q}(t) = a_1 + 2a_2t + 3a_3t^2$$

$$\ddot{q}(t) = 2a_2 + 6a_3t$$

- Find constants by setting initial and final positions and velocities and choosing a trajectory time.
- **4 variables, need to write 4 independent equations**

# Cubic Polynomial - Example

- Use a cubic polynomial to describe motion from 0 to 30 degrees in 3 seconds, with zero start and stop velocities.

- First write the two position equations:

$$q(t) = a_0 + a_1t + a_2t^2 + a_3t^3$$

$$0 = a_0$$

$$30 = a_0 + 3a_1 + 9a_2 + 27a_3$$

- Then write the two velocity equations:

$$\dot{q}(t) = a_1 + 2a_2t + 3a_3t^2$$

$$0 = a_1$$

$$0 = a_1 + 6a_2 + 27a_3$$

- Solve for  $a_0$ ,  $a_1$ ,  $a_2$  and  $a_3$ .

# Cubic Polynomial - Example

- Solve for  $a_0$ ,  $a_1$ ,  $a_2$  and  $a_3$ .

$$0 = a_0$$

$$30 = a_0 + 3a_1 + 9a_2 + 27a_3$$

$$0 = a_1$$

$$0 = a_1 + 6a_2 + 27a_3$$

- $a_0 = 0$ ,  $a_1 = 0$

$$30 = 9a_2 + 27a_3$$

$$0 = 6a_2 + 27a_3$$

- $a_2 = 10$ ,  $a_3 = -60/27 = -2.22$

# Cubic Polynomial - Example

- $a_0 = 0, a_1 = 0$
- $a_2 = 10, a_3 = -60/27 = -2.22$
- Write the equations for  $q_i(t), \dot{q}_i(t)$

$$q(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3$$

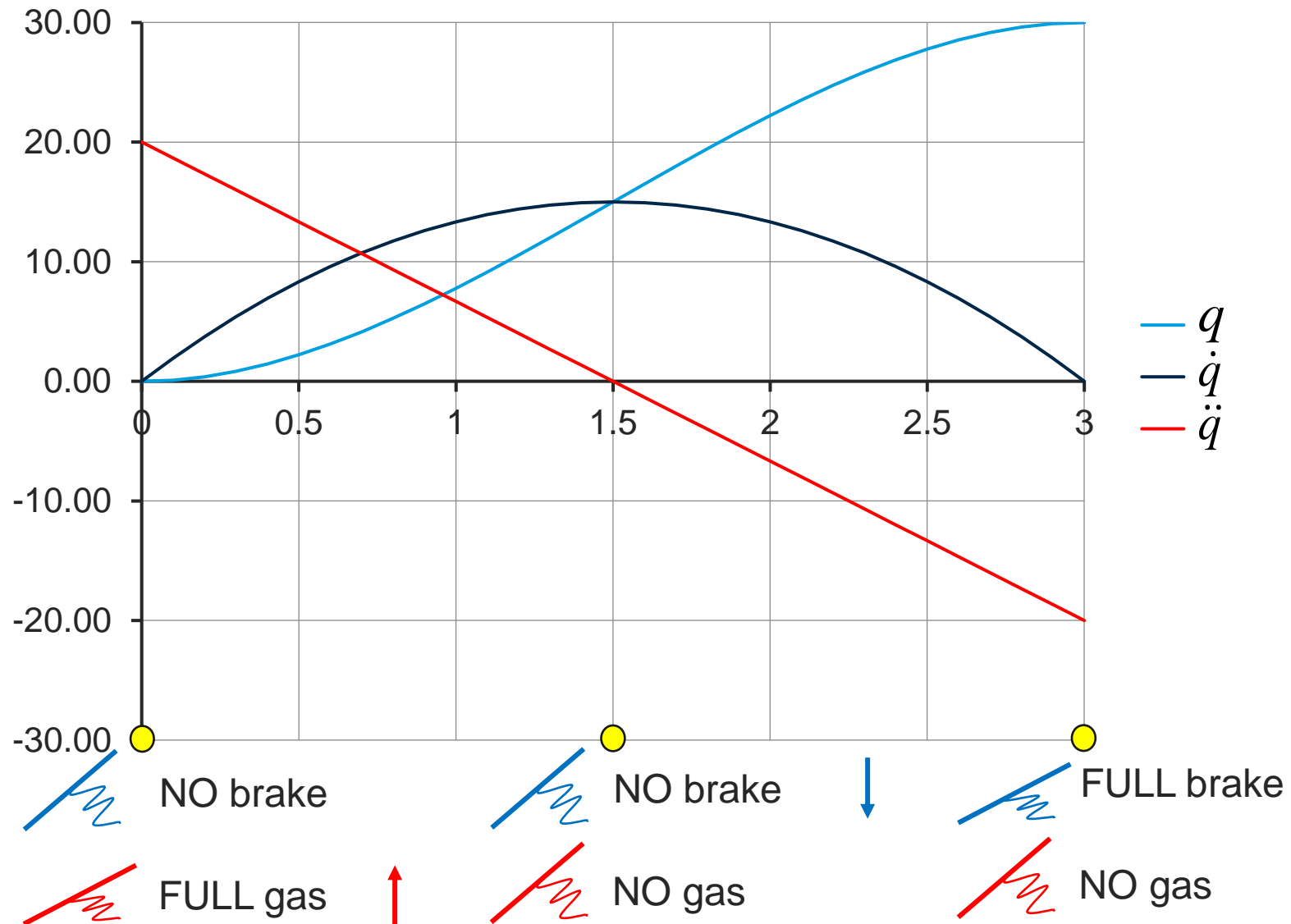
$$\dot{q}(t) = a_1 + 2a_2 t + 3a_3 t^2$$

$$q(t) = 10t^2 - 2.22t^3$$

$$\dot{q}(t) = 20t - 6.66t^2$$



# Cubic Polynomial - Example



# Cubic Polynomial Drawbacks

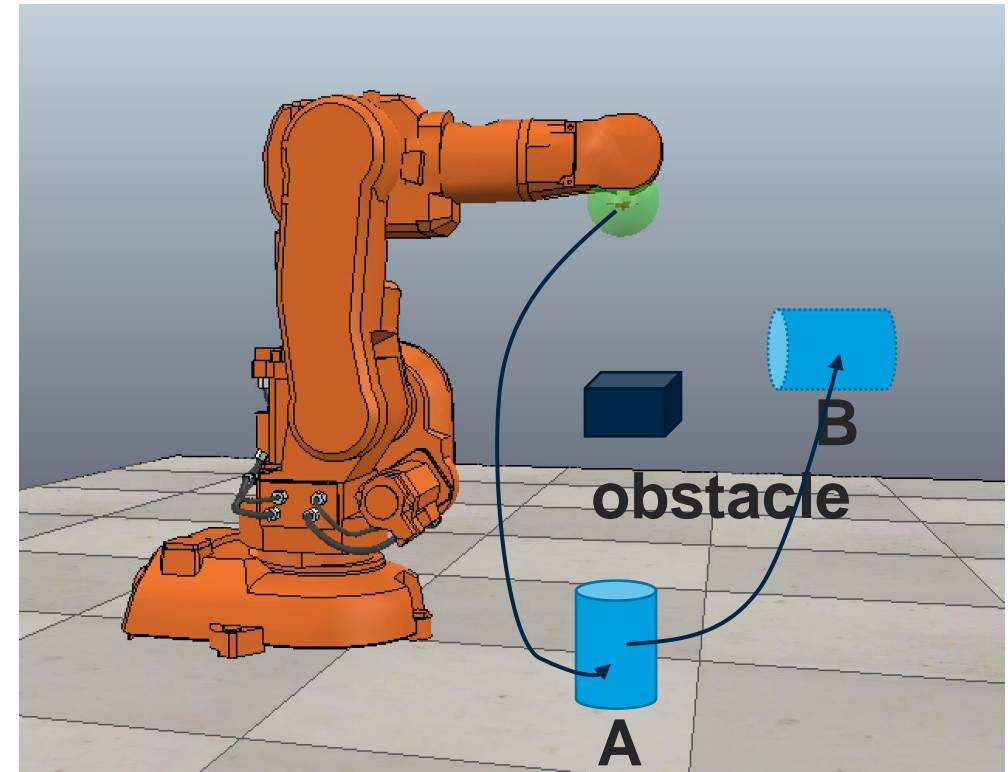
- Doesn't readily facilitate **minimum time** operations.
  - Not using full actuator capability.
- Need to choose **end and start velocities** when **linking** trajectories (to avoid discontinuities).
  - Can use higher order polynomial to ensure smooth connections.
- **Infinite jerk** (derivative of acceleration) at start and end.
  - Some loss of smoothness.

# Summary

- **Path planning** is to find a **collision free** path in the **joint space** to connect the starting configuration to the goal configuration.
- **Artificial potential field** method can formulate the problem in the **task space** then convert it into the **joint space** with the Jacobian matrix.
- **Trajectory** is path associated with **time**.
- **Cubic polynomial trajectory** or even higher order trajectories can be used.

# Motivating Problem

- Imagine one of your arms is replaced by a robotic arm. You are supposed to move an object from A to B.
- Now you know where the object is in front of you (homogeneous transformation).
- You know where your “hand” is with respect to your “body” (forward kinematics).
- You know how to move your “hand” to reach the object (inverse kinematics) at a certain speed (velocity kinematics).
- **Can you find a path to move the object while avoiding the obstacle?**
  - Path and trajectory planning



# Final Remarks

- Acknowledgements
  - Some material of the slides was developed by the previous lecturers of EGB339 - Introduction to Robotics (Michael Milford, Peter Corke, and Leo Wu)