# ECE 470: Lab 5 Inverse Kinematics

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## 1 Introduction

In this lab, we are required to calculate and implement the inverse kinematics on the UR3e robot arm. Considering the complexity of computing the setting of robot arm, the lab simplified the condition, fixing some part of the joint with constant angle. All we need to do is to do senior-high school level calculation and implement the process in Python. After that, we need to verify the correctness with Forward Kinematics code in Lab4.

# 2 written derivation of the inverse kinematics solution

Below image is the hand-derivation of IK:

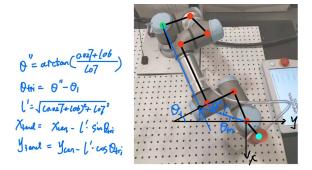


Figure 1: Physical Layout of the UR3e, we calculate the angle based on the given limb length

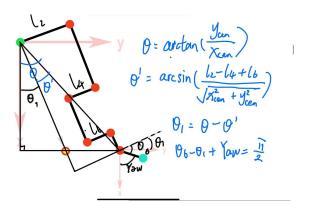


Figure 2: Top View Stick Pictorial of UR3e and Calculation

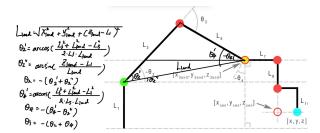


Figure 3: Side View Stick Pictorial of UR3e and Calculation

### 3 Source of Errors

As we learned in **CS450:** Numerical Analysis, several error can affect the accuracy and reliability of the joint angle calculations.

### 3.1 Numerical Approximations

The inverse kinematics solutions rely on numerical methods that approximate continuous mathematical operations. Finite-precision arithmetic in digital computers can introduce small discrepancies in calculations, particularly in trigonometric functions and matrix operations critical for robotic transformations.

#### 3.2 Measurement and Calibration Errors

Physical discrepancies in the robot's structure, such as variations in link lengths and joint calibration, can lead to errors in the kinematics model. Misalignments and measurement inaccuracies directly impact the calculated positions of the robot's end-effector.

### 3.3 Assumptions in Kinematic Model

The kinematic model simplifies the robot by assuming rigid links and perfect joints. In reality, mechanical play, joint elasticity, and non-linear friction characteristics can deviate the actual movement from the computed paths, particularly under different load conditions.

## 3.4 Algorithmic Limitations

The algorithms used for solving inverse kinematics may not account for all possible configurations of the robot. For example, configurations near or at kinematic singularities can result in solutions that are not feasible or not unique, thus impacting the performance of the robot in executing precise movements.

## 3.5 Precision of Input Parameters

Finally, the precision of input parameters such as the desired position and orientation of the end-effector can significantly influence the output. Inaccuracies in specifying these inputs might be magnified through the kinematic chain, resulting in larger positional errors at the end-effector.

These factors must be carefully considered and managed to improve the accuracy of the inverse kinematics solutions implemented in robotic systems.

## 4 Conclusion

This lab exercise successfully derived and implemented inverse kinematics for the UR3e robot arm, integrating theoretical calculations with practical Python coding. The accuracy of our results was confirmed through comparison with forward kinematics, highlighting the effectiveness of our approach.

The project underscored the importance of addressing potential sources of error such as numerical inaccuracies and structural deviations, which are crucial for understanding both the limitations and capabilities of robotic kinematics.

Overall, this lab not only helped our understanding of robotics mechanics and control but also demonstrated the critical nature of precision in robotic engineering. The insights gained here will be valuable for future robotics and automation projects, emphasizing the practical application of theoretical knowledge.

# 5 Appendix: Code Implementation

Here, we merged the lab 4 and lab 5 code together into 'lab5\_test.py' to fasten the verification. The code passed all test cases and got 80/80 grade in the lab demo.

```
import numpy as np
import math
from scipy.linalg import expm
import sys
import time
PI = np.pi
11 11 11
You may write some helper functions as you need
Use 'expm' for matrix exponential
Angles are in radian, distance are in meters.
def Get MS():
        # Fill in the correct values for S1~6, as well as the M matrix
        M = np.eye(4)
        S = np.zeros((6,6))
        S[0] = np.array([0,1,1,1,0,1])
        S[1] = np.array([0,0,0,0,-1,0])
        S[2] = np.array([1,0,0,0,0,0])
        S[3] = np.array([0,0,0,0,0.152,0])
        S[4] = np. array([0, 0.152, 0.152, 0.152, 0.152])
        S[5] = np.array([0,0,0.244,0.457,-0.131,0.542])
        M[0] = np.array([0,0,1,0.293])
        M[1] = np.array([0,1,0,-0.542])
        M[2] = np.array([-1,0,0,0.152])
        M[3] = np.array([0,0,0,1])
        return M, S
Function that calculates encoder numbers for each motor
def lab fk(theta1, theta2, theta3, theta4, theta5, theta6):
        theta = np.array([theta1, theta2, theta3, theta4, theta5, theta6])
        M, S = Get MS()
        print (f "M_ is  : \ n\{M\} \ n\ " )
        print (f "S_i is i : n\{S\} \setminus n \setminus n")
        T = np.eve(4)
        for i in range (6):
                 T = np.dot(T, expm(skew(S[:,i])*theta[i]))
        T = np.dot(T,M)
```

```
print("Forward_kinematics_calculated:\n")
         print (f "T \cup is \cup: \n{T}\n\n")
         return T
def skew(s):
         return np.array([
                  [0, -s[2], s[1], s[3]],
                  [s[2], 0, -s[0], s[4]],
                 [-s[1], s[0], 0, s[5]],
                 [0,0,0,0]
        ])
11 11 11
Function that calculates an elbow up Inverse Kinematic solution for the UR3
def lab invk(xWgrip, yWgrip, zWgrip, yaw WgripDegree):
    \# theta1 to theta6
         thetas = [0.0, 0.0, 0.0, 0.0, 0.0, 0.0]
         101 = 0.152
         102 = 0.120
         103 = 0.244
         104 = 0.093
         105 = 0.213
         106 = 0.104
         107 = 0.085
         108 = 0.092
         109 = 0
         110 = 0.07
                      \# thickness of aluminum plate is around 0.01
         yaw WgripDegree = yaw WgripDegree * PI / 180
         xgrip = -yWgrip
         ygrip = xWgrip
         zgrip = zWgrip
         print("xyzgrip", xgrip, ygrip, zgrip)
         xcen = xgrip - np.cos(yaw WgripDegree) * 109
         ycen = ygrip - np.sin(yaw WgripDegree) * 109
         zcen = zgrip
        \# theta1
         thetas [0] = \text{math.atan2}(\text{ycen}, \text{xcen}) - \text{math.asin}((0.027 + 106)) / \text{math.sqrt}(\text{xcen} *
\# Default value Need to Change
        \# print(xcen, ycen, zcen, l02 - l04 + l06)
        \# print(0.027 + 106)
    \# \ l02 - l04 + l06 = 0.027 + l06 = 0.131
         \# theta6
         thetas [5] = PI - (PI / 2 - thetas [0]) - yaw WgripDegree # Default value Need
        1 \log = \text{math.sqrt} ((106 + 0.027) ** 2 + 107 ** 2)
```

```
theta tri = math.atan2(107, (106 + 0.027))
                     x3end = xcen - l_long * math.sin(theta_tri - thetas[0])
                     y3end = ycen - 1 long * math.cos(theta tri - thetas[0])
                     z3end = zcen + 110 + 108
                     temp = math.sqrt((z3end - 101) ** 2 + x3end ** 2 + y3end ** 2)
                     print("xyz3end", x3end, y3end, z3end)
                     thetas[1] = -(math.atan2((z3end - 101), math.sqrt(x3end ** 2 + y3end ** 2)) + math.sqrt(x3end ** 2 + y3end ** 2))
# Default value Need to Change
                     thetas [2] = PI - math.acos ((103 ** 2 + 105 ** 2 - temp ** 2) / (2 * 103 * 105))
# Default value Need to Change
                     thetas[3] = -thetas[1] - thetas[2] # Default value Need to Change
                                                                                  # Default value Need to Change
                     thetas [4] = -PI / 2
                     \mathbf{print}("theta1\_to\_theta6:\_\backslash n\backslash n" + \mathbf{str}(thetas) + "\backslash n")
                     return thetas
def main():
           if len(sys.argv) != 5:
                     \mathbf{print}("\n")
                     print("rosrun_lab5pkg ik_lab5 exec.py_xWgrip_yWgrip_zWgrip_yaw WgripDegree_\n")
           else:
                     \mathbf{print} \left( \text{"} \backslash \text{nxWgrip} : \text{\_"} + \text{sys.argv} [1] + \text{"}, \text{\_yWgrip} : \text{\_"} + \text{sys.argv} [2] + \text{\lor} \right)
                                     ", \_zWgrip: \_" + sys.argv[3] + ", \_yaw\_WgripDegree: \_" + sys.argv[4] + " \ ""
                     \# Assuming lab\_invk and lab\_fk are defined elsewhere and correctly imported
                     new_dest = lab_invk(float(sys.argv[1]), float(sys.argv[2]), float(sys.argv[3]),
                    T = lab_fk(new_dest[0], new_dest[1], new_dest[2], new_dest[3], new_dest[4], new_d
                     \mathbf{print}("\mathbf{check}_T[0][3], _T[1][3], _T[2][3]  with _\mathbf{your}_\mathbf{input}_\mathbf{destination}")
                     print ( f "T_i is_i \setminus n \setminus n\{T\}")
                     \mathbf{print}(" \setminus n \setminus n")
                     end effct = T[0:3,3]
                     \mathbf{print}(f"end effct_is_\backslash n \setminus \{end effct\}")
                     \mathbf{print}(" \setminus n \setminus n")
# Ensure the following functions are defined and imported correctly:
\# lab invk() and lab fk()
if __name__ == "__main__":
                     try:
                                         main()
                     except:
                                         print("error")
```