

# 5ELEN018W - Robotic Principles

## Lecture 5: Inverse Kinematics

Dr Dimitris C. Dracopoulos

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where  $\theta_i, \alpha_i, r_i, d_i$  are the DH parameters.

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For example, for 6-joint manipulator:

$${}^0T_6 = {}^0T_1 \cdot {}^1T_2 \cdot {}^2T_3 \cdot {}^3T_4 \cdot {}^4T_5 \cdot {}^5T_6 \quad (2)$$



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3. It is possible that no solutions exist

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- ▶ Manufacturing and assembly

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- ▶ To calculate, they take advantage of particular geometric features of specific robot mechanisms.
- ▶ As the number of joints increases, this becomes increasingly difficult.
- ▶ For some serial-link robot manipulators, no analytical (closed form) solution exists!

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1. Identify the significant equations containing the joint variables.
  2. Manipulate them into a soluble form.
- A common strategy is reduction to a equation in a single variable, e.g.

$$C_1 \cos \theta_i + C_2 \sin \theta_i + C_3 = 0$$

where  $C_1, C_2, C_3$  are constants.











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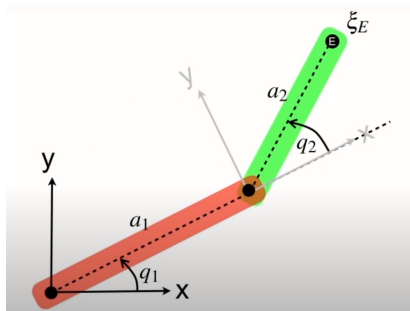
—→ often results to the decomposition of the spatial problem into separate planar problems.

- ▶ Decomposition of the full problem into inverse position kinematics and inverse orientation kinematics.
- ▶ The solution is derived by rewriting equation (2) as:

$${}^0T_6 \cdot {}^6T_5 \cdot {}^5T_4 \cdot {}^4T_3 = {}^0T_1 \cdot {}^1T_2 \cdot {}^2T_3 \quad (3)$$

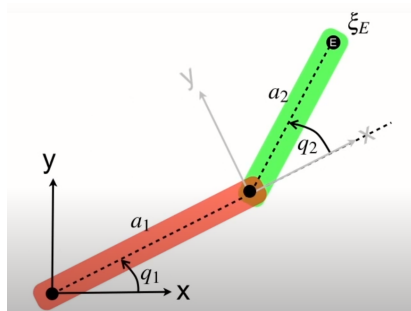
# Calculating Analytical Solutions in the Python Robotics Toolbox

Consider a 2-joint Planar (2D) Robot



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Given the position of the end-effector  $(x_E, y_E)$  calculate the required joint angles to achieve this position.

# Calculating Analytical Solutions in Python (con'd)

*# create symbols for lengths of the 2 links*

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>>> TE = e.fkine([q1, q2])
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# translation part of matrix gives the position (x_fk, y_fk) of the  
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>>> x_fk, y_fk = TE.t
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>>> print(x_fk)
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# then  $x\_fk**2 + y\_fk**2 - x**2 - y**2 = 0$ 
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>>> eq1 = (x_fk**2 + y_fk**2 - x**2 - y**2).trigsimp()
```

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a1**2 + 2*a1*a2*cos(q2) + a2**2 - x**2 - y**2

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>>> print(q2_sol)
[-acos(-(a1**2 + a2**2 - x**2 - y**2)/(2*a1*a2)) + 2*pi,
 acos((-a1**2 - a2**2 + x**2 + y**2)/(2*a1*a2))]
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# expand the two equations x_fk=x, y_fk=y
>>> eq2 = tuple(map(sympy.expand_trig, [x_fk - x, y_fk - y]))
>>> print(eq2)
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$$(a1 \cos(q1) + a2(-\sin(q1) \sin(q2) + \cos(q1) \cos(q2)) - x, \\ a1 \sin(q1) + a2(\sin(q1) \cos(q2) + \sin(q2) \cos(q1)) - y)$$

*# solve for sin(q1), cos(q1)*

```
>>> q1_sol = sympy.solve(eq2, [sympy.sin(q1), sympy.cos(q1)])
>>> print(q1_sol) # dictionary containing sin(q1) and cos(q1)
```

# Calculating Analytical Solutions in Python (con'd)

```
#  $\tan(q1) = \sin(q1)/\cos(q1)$   
>>> print(q1_sol[sin(q1)]/q1_sol[cos(q1)])
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>>> print(q1_sol[sin(q1)]/q1_sol[cos(q1)])

# solve for  $q_1$ 
>>> sympy.atan2(q1_sol[sin(q1)], q1_sol[cos(q1)]).simplify()
```



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**How?** Minimise the error between the forward kinematics solution and the desired end-effector pose  $\xi_E$ :

$$\mathbf{q}^* = \arg \min_{\mathbf{q}} (FK(\mathbf{q}) - \xi_E)$$

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Various classical numerical methods can be applied, including among others:

- ▶ Newton-Raphson: first order approximation of original equations
- ▶ Levenberg–Marquardt optimisation: using the second order derivative for the approximation of the original system.

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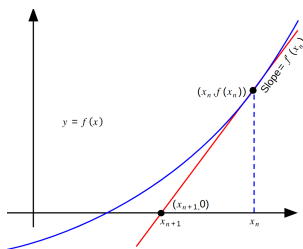
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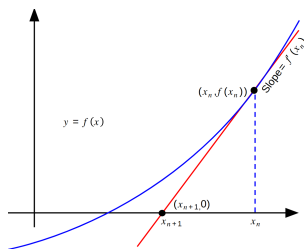
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Then the approximated solution for finding the root of  $f$  (where  $f(x) = 0$ ) can be calculated iteratively by:

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)} \quad (5)$$

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>>> def E(q):
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t = 0.5, 0.4; -32.7°
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