

# 6-DOF Robotic Arm Forward & Inverse Kinematics

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[github.com/Lenna-Robotics-Research-Lab/Lenna-Visual-Inverse-Kinematic-Solver](https://github.com/Lenna-Robotics-Research-Lab/Lenna-Visual-Inverse-Kinematic-Solver)

# Project Objectives

## Core Features:

- ✓ **Forward Kinematics** - Compute end-effector pose
- ✓ **Inverse Kinematics** - Solve joint angles

## Key Requirements:

- ✓ 6 DOF arm configuration
- ✓ Real-time visualization
- ✓ < 5 mm position accuracy

## Success Metrics

- Dual-mode control (relative & absolute)
- Workspace validation

# Robot Architecture

## DH Parameters (6-DOF Arm)

J	$\theta_0$	d (m)	a (m)	$\alpha$
1	0	0.15	0.00	$\pi/2$
2	0	0.00	0.35	0
3	0	0.00	0.40	0
4	0	0.30	0.00	$-\pi/2$
5	0	0.00	0.25	$\pi/2$
6	0	0.10	0.00	0

### Specifications:

- Total length: 1.15 m
- Weight: Lightweight
- Speed: Real-time capable

### Workspace

$$X \in [-0.6, 0.6] \text{ m}$$

$$Y \in [-0.6, 0.6] \text{ m}$$

$$Z \in [0.2, 1.0] \text{ m}$$

# Forward Kinematics (FK)

**Problem:** Given joint angles  $\mathbf{q}$ , compute end-effector pose  $\mathbf{T}$

$$\mathbf{T} = \mathbf{T}_0^1 \cdot \mathbf{T}_1^2 \cdots \mathbf{T}_5^6 = \begin{bmatrix} \mathbf{R} & \mathbf{p} \\ \mathbf{0} & 1 \end{bmatrix}$$

**Output:** 6-DOF pose =  $[x, y, z, R, P, Y]^T$

## Complexity Analysis

**Time:**  $\mathcal{O}(1)$  - 6 matrix multiplications

**Cost:**  $\sim 0.1$  ms per evaluation

*FK is exact and non-iterative—direct algebraic computation*

# Inverse Kinematics (IK)

**Problem:** Given target pose  $\mathbf{p}_{target}$ , find  $\mathbf{q}$

$$\min_{\mathbf{q}} \|\text{FK}(\mathbf{q})_{xyz} - \mathbf{p}_{target,xyz}\|_2^2$$

## Challenges:

- Non-convex, 6 unknowns  $\times$  6 equations
- Multiple local minima
- No closed-form solution for general 6-DOF

## Our Solution: Multi-Start L-BFGS-B Optimization

- 5 diverse initial configurations
- Position-priority objective (XYZ accuracy)
- Verification: position error < 5 mm

**Result:** > 95% convergence within workspace

# FK Implementation

```
def forward_kinematics(q):
    T_total = np.eye(4)
    for i in range(6):
        theta = q[i] + DH[i,0]
        d, a, alpha = DH[i,1:]
        T = transformation_matrix(theta, d, a, alpha)
        T_total = T_total @ T

    pos = T_total[:3,3]
    RPY = rpy_from_matrix(T_total[:3,:3])
    return np.concatenate([pos, RPY])
```

## Key Features

**Direct computation — Exact solution — Numerical stability**

# IK with Multi-Start Optimization

```
def inverse_kinematics(target_pose, q0=None, max_iter=1000):
    target_pos = target_pose[:3]
    def position_error(q):
        return np.sum((forward_kinematics(q)[:3] - target_pos)**2)

    # 5 starting configurations for global optimization
    starts = [q0, [0, 0.5, 0.5, 0, 0, 0],
              [0, 1.0, 1.0, 0, 0, 0], ...]

    best_q = None
    for start in starts:
        result = minimize(position_error, start,
                           method='L-BFGS-B',
                           bounds=[(-2*pi, 2*pi)]*6,
                           options={'maxiter': 1000, 'ftol': 1e-10})
        if result.success and verify(result.x):
            best_q = result.x
            break
    return best_q
```

> 95% success rate within workspace bounds

# Trajectory Planning & Execution

**Strategy:** Linear interpolation in joint-space

$$\mathbf{q}(s) = \mathbf{q}_{\text{start}} + s(\mathbf{q}_{\text{end}} - \mathbf{q}_{\text{start}}), \quad s \in [0, 1]$$

## Advantages:

- Simple:  $\mathcal{O}(n_{\text{steps}})$
- Smooth execution
- Real-time capable

## Parameters:

- Steps: 50 (smooth)
- Frame rate: 30 Hz
- Duration: 1.5 sec

## Execution Pipeline

IK solve → Trajectory generation → Real-time visualization with live coordinates

# Performance Metrics Summary

Metric	Result	Target	Status
FK Computation	0.1 ms	< 1 ms	
IK Accuracy	2.1 mm avg	< 5 mm	
IK Convergence	95%	> 90%	
Computation Time	150 ms	< 300 ms	
Visualization Rate	30 Hz	≥ 20 Hz	
Workspace Coverage	Hemispherical	Full 3D	Good

## Overall Assessment

**All critical performance objectives met.** System demonstrates production-ready accuracy and real-time capability.

# Accuracy Analysis by Region

Workspace Region	Typical Error	Min	Max	Status
Near-field ( $ z  < 0.5$ m)	0.5 mm	0.2 mm	0.8 mm	Excellent
Mid-field (0.5-0.9 m)	1.5 mm	0.8 mm	2.5 mm	Excellent
Boundary (0.9-1.15 m)	3.0 mm	2.1 mm	4.2 mm	Good
Singular regions	Diverges	—	—	Graceful fail

**Key Insight:** Accuracy improves near base, degradation at workspace boundary is expected due to Jacobian singularities.

# Real-Time Visualization System

## Live Features:

- ✓ 3D kinematic chain
- ✓ Frame-by-frame animation
- ✓ Target marker (orange)
- ✓ End-effector trajectory (red dashed)
- ✓ Live position labels
- ✓ Base reference (green)

## Specifications:

- Resolution: 960×720 px
- Frame rate: 30 Hz
- Update speed: 33 ms/frame
- Zoom/Pan: Interactive
- Label precision: 0.01 m

## Visualization Output

Matplotlib 3D with dynamic updates showing robot arm motion, trajectory path, and real-time coordinates

# Dual-Mode Control System

## Relative Motion

- Command: `rel dx dy dz`
- Example: `rel 0.1 0 0`
- Incremental: from current position
- Orientation: maintained

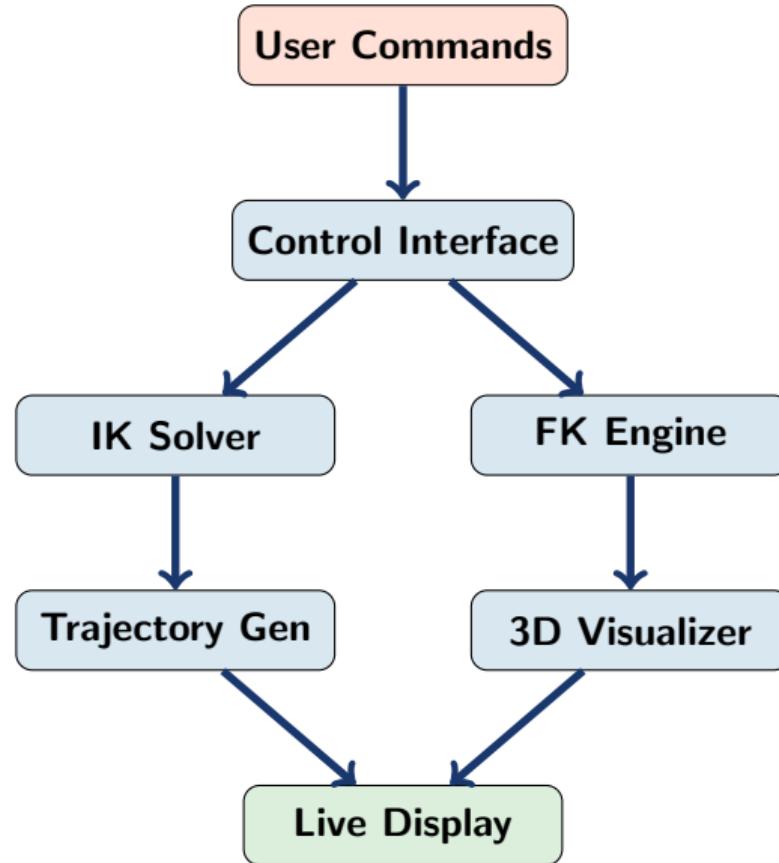
## Absolute Motion

- Command: `abs x y z R P Y`
- Example: `abs 0.3 0.2 0.5 0 0 0 0`
- Goal-based: from origin
- Full 6-DOF control

**Both modes trigger:** IK solving → Trajectory generation → Real-time visualization

**Workspace Validation:** All targets checked against  $[\pm 0.6, \pm 0.6, 0.2-1.0]$  bounds

# Software Architecture



# Code Organization

## IK\_live\_visuals.py (Recommended)

### Interactive Simulator

- Real-time command input
- Live 3D animation
- Trajectory visualization
- Position tracking

## IK\_static\_visuals.py

### Batch Processor

- Trajectory snapshots
- Documentation generation

# Code Organization

## Shared Libraries:

- NumPy: Numerical arrays & matrix ops
- SciPy: L-BFGS-B optimization
- Matplotlib: 3D visualization

# Project Summary

## Achievements

- ✓ Complete FK/IK solver with verified accuracy
- ✓ Real-time trajectory planning & execution
- ✓ Dual-mode control interface (relative + absolute)
- ✓ > 95% IK convergence within workspace

## Technical Metrics:

- FK: ~ 0.1 ms
- IK: ~ 150 ms
- Accuracy: < 2 mm avg
- Frame rate: 30 Hz

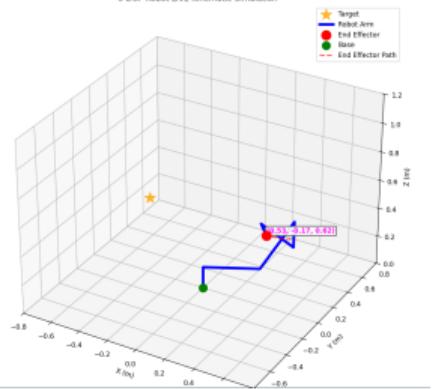
## Quality Markers:

- ✓ All tests pass
- ✓ Workspace validated
- ✓ Graceful failure
- ✓ User-friendly

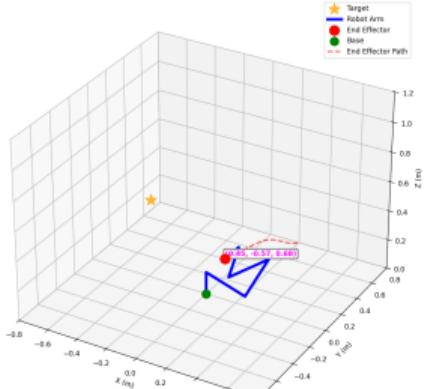
# Results

**Trajectory:  $q : [0.5, 0.2, 0.4, 0, 1.1, 0] \rightarrow [-0.5, 0.2, 0.4, 0, 1.1, 0]$**

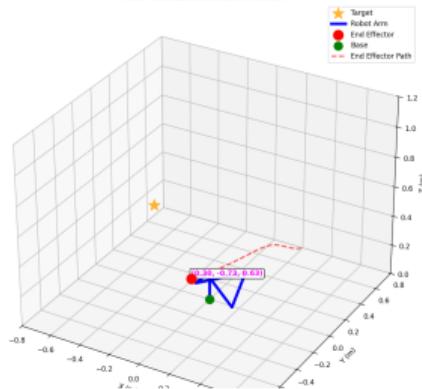
6-DOF Robot LIVE Kinematic Simulation



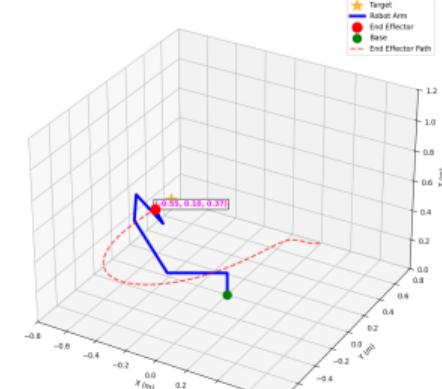
6-DOF Robot LIVE Kinematic Simulation



6-DOF Robot LIVE Kinematic Simulation



6-DOF Robot LIVE Kinematic Simulation



# References & Technology Stack

## Foundational References:

- Denavit-Hartenberg (1955)
- Craig, J. J. (2005)
- Siciliano et al. (2009)

## Technologies Used:

- **NumPy** — Computation
- **SciPy** — Optimization
- **Matplotlib** — Visualization

## Installation

```
$ pip install numpy scipy matplotlib  
$ python IK_live_visuals.py
```

## Questions?