

Figure-8 Trajectory Tracking for Franka Emika using Reinforcement Learning

Natcha Jengjirapas

ETH Zurich, Switzerland

System Architecture

- **Robot Hardware:** Franka Emika Panda
- **Physics Engine:** MuJoCo
- **Framework:** Gymnasium
- **RL Algorithm:** PPO
- **CPU:** AMD Ryzen 9 7950x
- **OS:** Window 11 with WSL2 (Ubuntu 22.04/24.04)



RL: State & Observation Space

$$[q, \dot{q}, \Delta p, \Delta R, v_{des}, w_{des}, \sin\phi, \cos\phi, a, b]$$

- `_get_obs` function
- Total: 30-dim float32 vector

RL: State & Observation Space

$$[\underline{q}, \dot{q}, \Delta p, \Delta R, v_{des}, w_{des}, \sin\phi, \cos\phi, a, b]$$

Body Schema

- `_get_obs` function
- Total: 30-dim float32 vector

RL: State & Observation Space

$$[q, \dot{q}, \underline{\Delta p, \Delta R}, v_{des}, w_{des}, \sin\phi, \cos\phi, a, b]$$

Tracking Errors

- `_get_obs` function
- Total: 30-dim float32 vector

RL: State & Observation Space

$$[q, \dot{q}, \Delta p, \Delta R, \underline{v_{des}, w_{des}}, \sin\phi, \cos\phi, a, b]$$

Reference Dynamics

- `_get_obs` function
- Total: 30-dim float32 vector

RL: State & Observation Space

$$[q, \dot{q}, \Delta p, \Delta R, v_{des}, w_{des}, \sin\phi, \cos\phi, a, b]$$

Phase Embedding

- `_get_obs` function
- Total: 30-dim float32 vector

RL: State & Observation Space

$$[q, \dot{q}, \Delta p, \Delta R, v_{des}, w_{des}, \sin\phi, \cos\phi, \underline{a}, b]$$

Task Parameters

- `_get_obs` function
- Total: 30-dim float32 vector

RL: Action Space

$$q_{cmd}^{(t+1)} = \text{clip} \left(q_{act}^{(t)} + (a^{(t)} \cdot \lambda), q_{min}, q_{max} \right)$$

- Action: Normalized in [-1 , 1]
- Scale: 0.02 - 0.08 rad/step
- Frequency: 40 Hz control loop

RL: Action Space

$$q_{cmd}^{(t+1)} = \text{clip} \left(q_{act}^{(t)} + \underbrace{(a^{(t)} \cdot \lambda)}_{\text{Action Vector}}, q_{min}, q_{max} \right)$$

- Action: Normalized in [-1 , 1]
- Scale: 0.02 - 0.08 rad/step
- Frequency: 40 Hz control loop

RL: Action Space

$$q_{cmd}^{(t+1)} = \text{clip} \left(q_{act}^{(t)} + \underbrace{(a^{(t)} \cdot \lambda)}_{\text{Scaling Factor}}, q_{min}, q_{max} \right)$$

Scaling Factor

- Action: Normalized in [-1 , 1]
- Scale: 0.02 - 0.08 rad/step
- Frequency: 40 Hz control loop

RL: Action Space

$$q_{cmd}^{(t+1)} = \text{clip} \left(\underbrace{q_{act}^{(t)} + (a^{(t)} \cdot \lambda)}_{\text{Delta Integration}}, q_{min}, q_{max} \right)$$

- Action: Normalized in [-1 , 1]
- Scale: 0.02 - 0.08 rad/step
- Frequency: 40 Hz control loop

RL: Action Space

$$q_{cmd}^{(t+1)} = \text{clip} \left(q_{act}^{(t)} + (a^{(t)} \cdot \lambda), q_{min}, q_{max} \right)$$

Safety Guard

- Action: Normalized in [-1 , 1]
- Scale: 0.02 - 0.08 rad/step
- Frequency: 40 Hz control loop

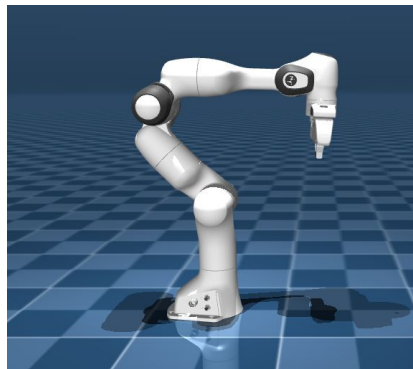
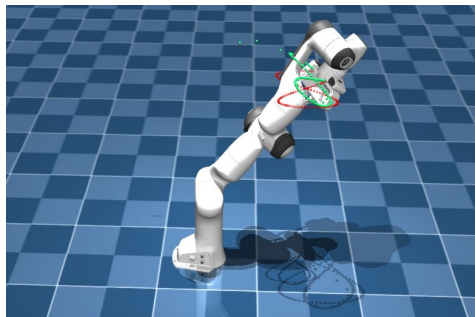
RL: Reward Function

$$R = -(w_p \|\Delta p\|^2 + w_R \|\Delta R\|^2 + w_u \|a\|^2)$$

- **Position**: tracking accuracy
- **Rotation**: orientation alignment
- **Action Penalty**: encourage smooth torque profiles

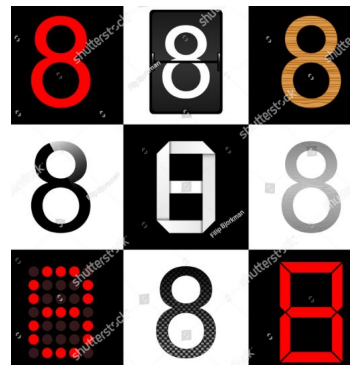
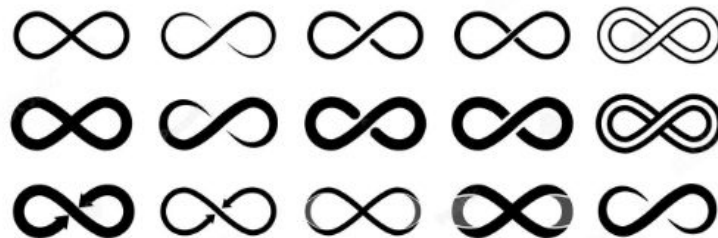
RL: Environment Initialization & Constraints

- Athletic Home Pose: $\mathbf{q}_{home} = [0, -0.785, 0, -2.356, 0, 1.571, 0.785]$
- Termination Criteria:
 - `Terminate_pos_err_m` = 1.2
 - `Termination_grace_steps` = 25



RL: Reset

- Randomization:
 - Width (a): [0.10, 0.17] m
 - Height (b): [0.08, 0.12] m
 - X-Axis: [0.25, 0.40] m
 - Y-Axis: [0.00, 0.10] m
 - Phase: [0.00, 2π] m



Training & Refinement

- PPO:

- MlpPolicy
- n step: 2048
- batch: 256
- ent_coef: 0.005
- Total Time Step: 500,000

- Env:

- Position Weight : 25 -> 100 -> 1000
- Orientation Weight: 0.5
- Action Weight: 0.02
- action_scale = 0.02 -> 0.05 -> 0.08
- F_hz = 0.75 -> 1 -> 1.3

Performance & Results

- Training Seed: 0
- Testing Seed: 24
- Success: < 5cm

Table 1: Frequency 1.0Hz Results Across 10 Randomized Episodes ($N = 10$)

| Test Case | Size ($a \times b$) | Pos. Error (cm) | Ori. Error (rad) | Success Rate |
|------------------|-----------------------|-----------------|------------------|--------------|
| In-Distribution | 0.14×0.10 | 2.62 | 0.3418 | 100% |
| Small Scale | 0.10×0.06 | 2.17 | 0.3435 | 100% |
| Kinematic Limit* | 0.22×0.15 | 4.99 | 0.3472 | 40% |

Table 2: Frequency 1.3Hz Results Across 10 Randomized Episodes ($N = 10$)

| Test Case | Size ($a \times b$) | Pos. Error (cm) | Ori. Error (rad) | Success Rate |
|------------------|-----------------------|-----------------|------------------|--------------|
| In-Distribution | 0.14×0.10 | 2.86 | 0.2277 | 100% |
| Small Scale | 0.10×0.06 | 2.11 | 0.1965 | 100% |
| Kinematic Limit* | 0.22×0.15 | 6.8 | 0.2493 | 0% |

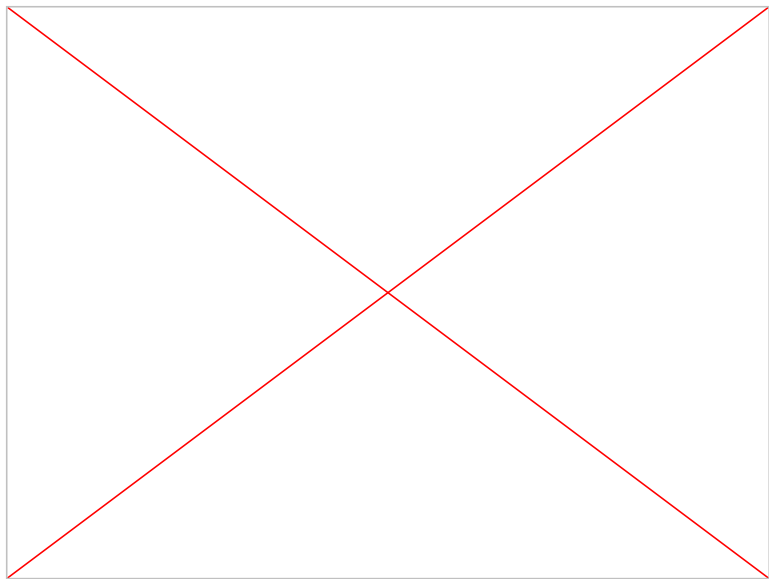
Performance & Results

Table 3: In-Distribution Validation Stability (Specialized Task)

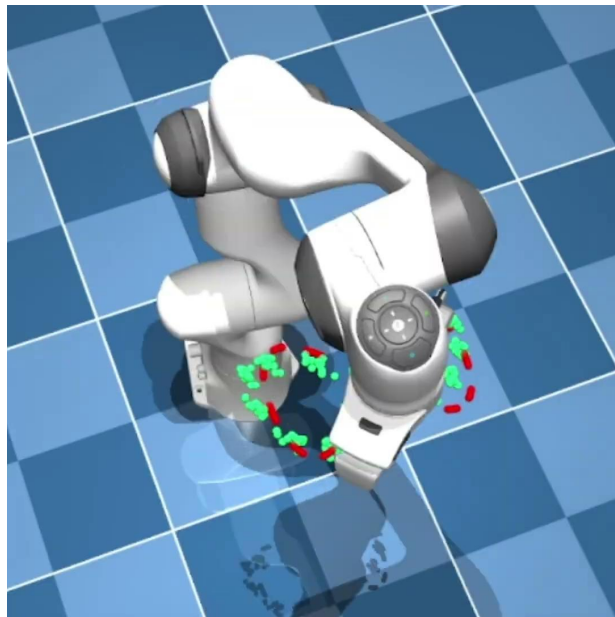
| Metric | Unit | 10-Run Validation | 50-Run Validation |
|------------------------|---------|-------------------|-------------------|
| Avg. Position Error | cm | 1.79 | 1.84 |
| Avg. Orientation Error | rad (°) | 0.1280 (7.33°) | 0.1291 (7.40°) |
| Success Rate | % | 100% | 100% |

Small Out of Distribution

1.0 Hz

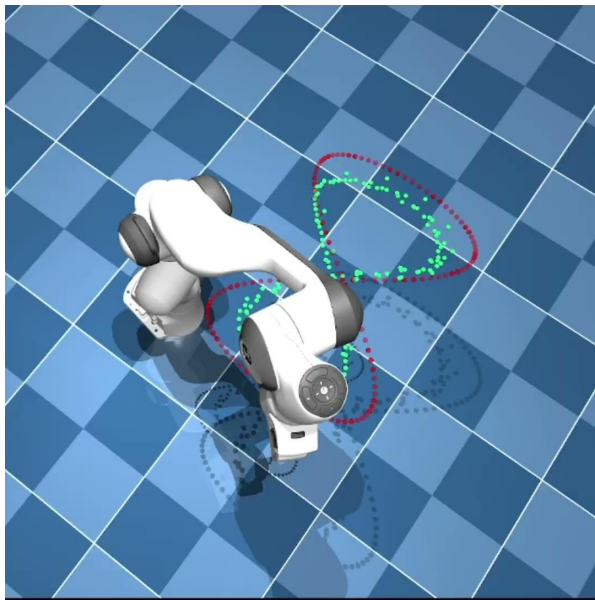


1.3 Hz

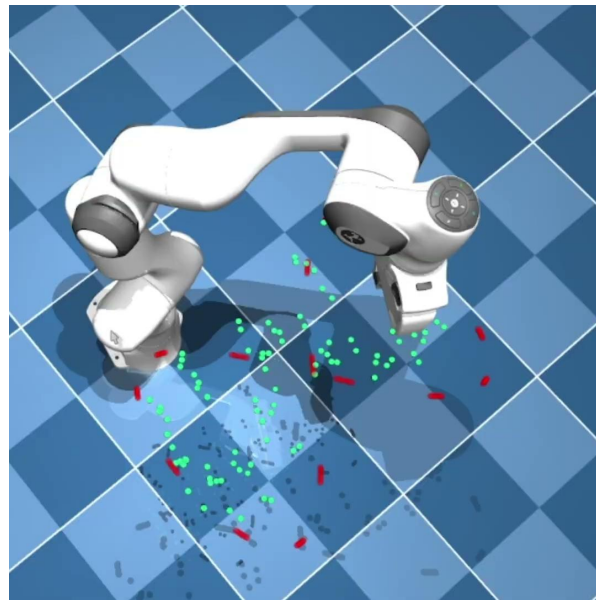


Large Out of Distribution

1.0 Hz

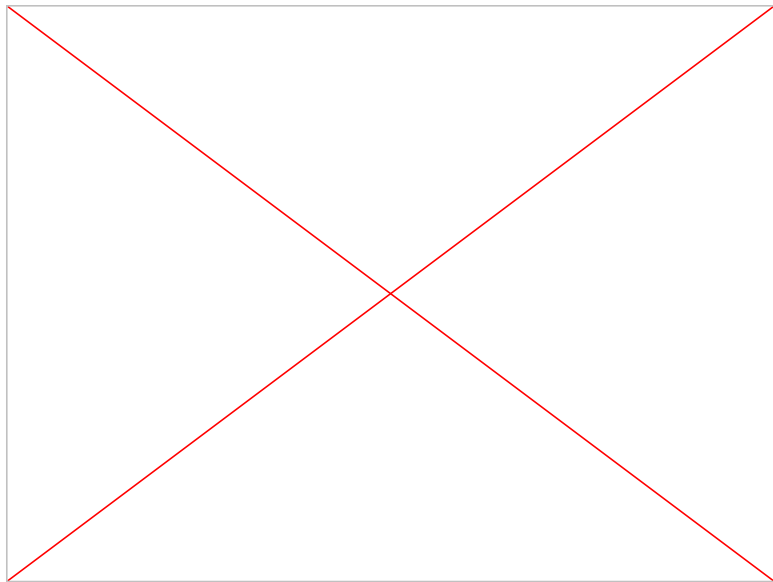


1.3 Hz

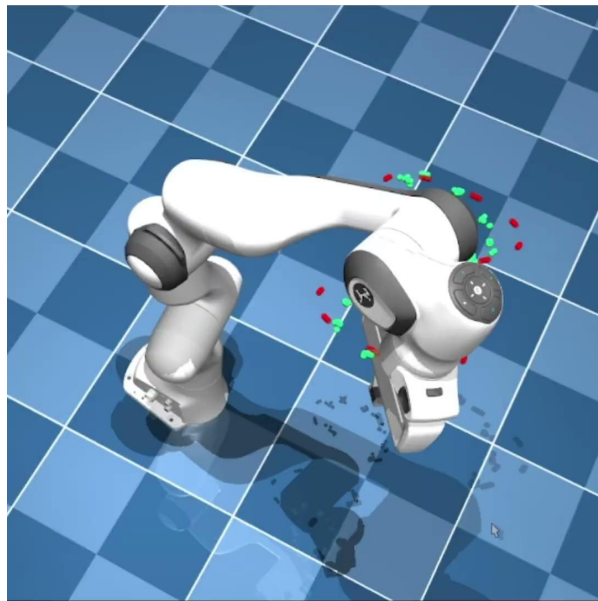


In Distribution

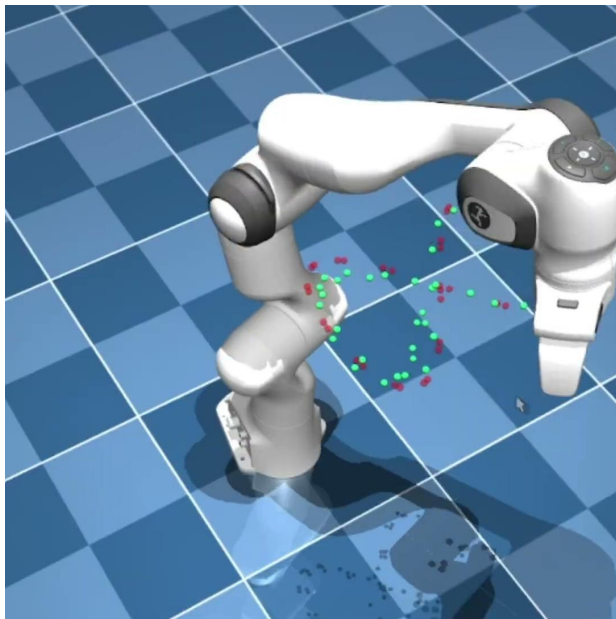
1.0 Hz



1.3 Hz



Specialized Version



Conclusion

- Achieved a consistent 2.62 cm avg tracking error
- 100% success rate on specialized tasks
- Generalized policy adapted to small OOD
- Identified the physical limit of the policy at $a = .22$ m, where Joint 4 and Joint 5 reaching its mechanical stop.

Joint Limitation

Joint 0: -2.90 to 2.90 radians

Joint 1: -1.76 to 1.76 radians

Joint 2: -2.90 to 2.90 radians

Joint 3: -3.07 to -0.07 radians

Joint 4: -2.90 to 2.90 radians

Joint 5: -0.02 to 3.75 radians

Joint 6: -2.90 to 2.90 radians

