

HIERARCHICAL REINFORCEMENT LEARNING FOR AUTONOMOUS PARKING

Combining A* Path Planning with PPO Control



MSML 642 ROBOTICS

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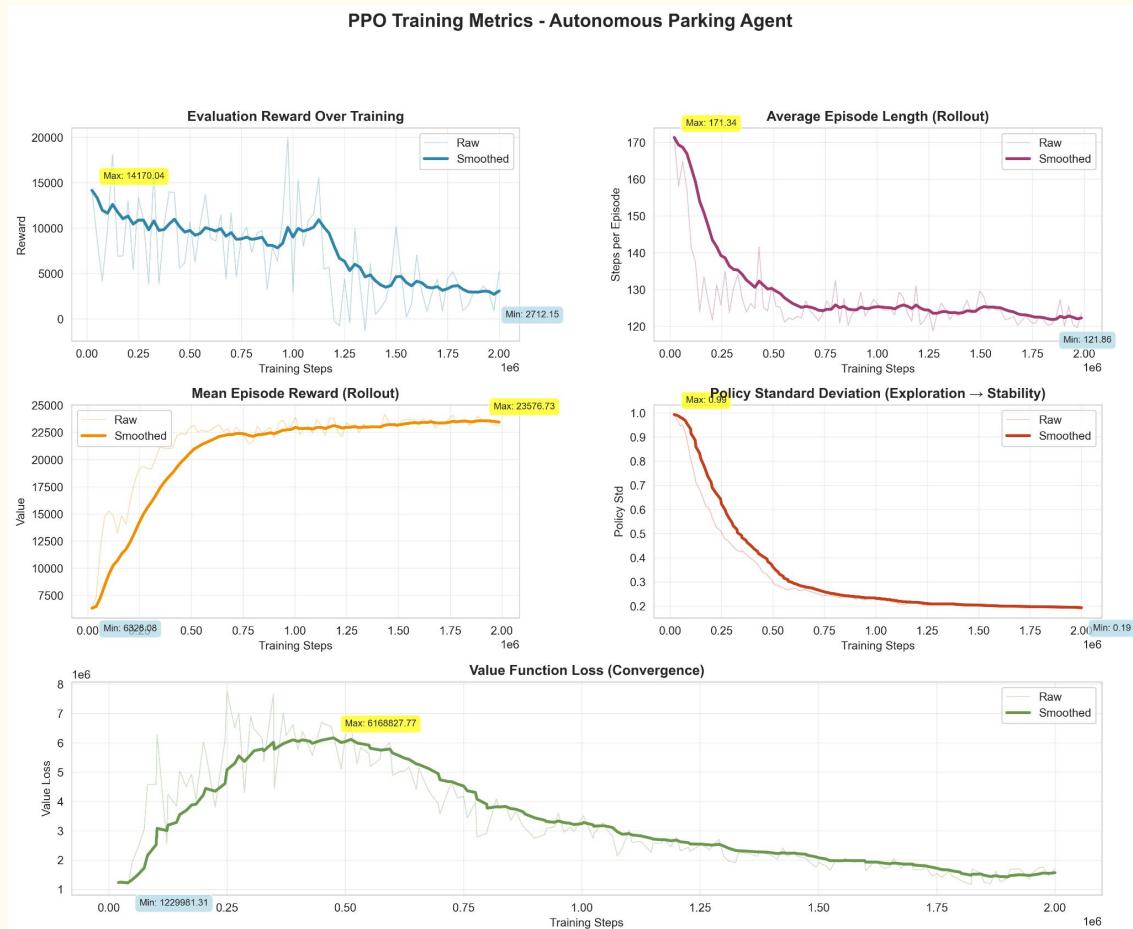
CURRENT RESULTS

Training Summary (2M steps, ~12 hours)

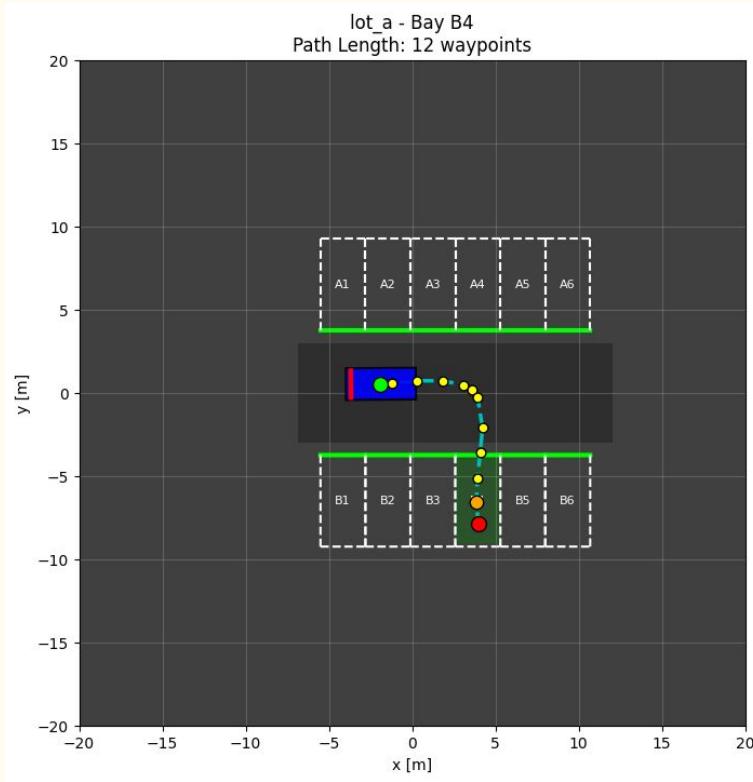
- Episode Reward increased from 6K → 23K
- Episode Length stable at ~125 steps (efficient navigation)
- Policy Std decreased 0.99 → 0.19 (exploration → control)
- Value Loss converged from 6M → 1.7M

Interpretation:

Agent is consistently improving, stabilizing, and following A* waypoint plans with high precision.



AUTONOMOUS PARKING IN LOTS



- INPUT:
 - Robot state (x, y, yaw, v)
 - Goal bay location, yaw
 - 64-ray Lidar scan ($360^\circ, 20\text{m}$)
- OUTPUT:
 - Steering $\in [-1, 1]$
 - Throttle $\in [-1, 1]$
- CHALLENGES
 - Long Horizon: 10-25 meter navigation, 500-2000 timesteps/episode
 - Multi-Orientation: Bays face different directions, Rotation-invariant parking, Progressive curriculum
 - Precision Parking: Position: $<0.5\text{m}$ error, Orientation: $<10^\circ$ error, Must stop completely

HIERARCHICAL ARCHITECTURE

HIGH-LEVEL PLANNER (A*)

- 0.25 m grid-based search
- Collision + corridor checks
 - Prevent cutting through parking bays
 - Forces path to stay inside driving lanes
- B-Spline smoothing
 - Removes sharp A* corners
 - Produces drivable curvature (car-like motion)

LOW-LEVEL CONTROLLER (PPO)

- Observation (~80 dims):
 - 64-ray lidar
 - Vehicle state (x , y , yaw, v)
 - Bay geometry (cx , cy , yaw error)
 - Waypoint tracking (distance + heading)
 - Corridor offset + progress signals
- Actions (continuous):
 - Steering $\in [-1, 1]$
 - Throttle $\in [-1, 1]$

REWARD FUNCTION DESIGN

NAVIGATION PHASE (r_nav) (dist > 3.0m)	PARKING PHASE (r_park) (dist < 3.0m)
<ul style="list-style-type: none"> Waypoint Progress: +0.1 per meter (only if moving forward) 	<ul style="list-style-type: none"> Lateral Alignment: +20.0 $(1 - \text{cy} /0.8)$
<ul style="list-style-type: none"> Path Velocity: $+0.5 \times v_{\text{path}}$ (reward approaching goal) 	<ul style="list-style-type: none"> Yaw Alignment: +15.0 $(1 - \Delta\theta /1.5)$
<ul style="list-style-type: none"> Backward Penalty: $-2.0 \times v$ (strong discouragement) 	<ul style="list-style-type: none"> Depth Progress: $+10.0 \times \Delta cx$ (drive deeper into bay)
<ul style="list-style-type: none"> Anti-freeze: -0.01 (keep moving) 	<ul style="list-style-type: none"> Success Bonus: +500.0 (dominates episode return)
<ul style="list-style-type: none"> Distance Penalty: $-0.002 \times \text{dist}$ 	
<ul style="list-style-type: none"> Corridor Drift Penalty: $-\text{corridor_weight} \times \text{lateral_offset}$ 	

UNIVERSAL PENALTIES

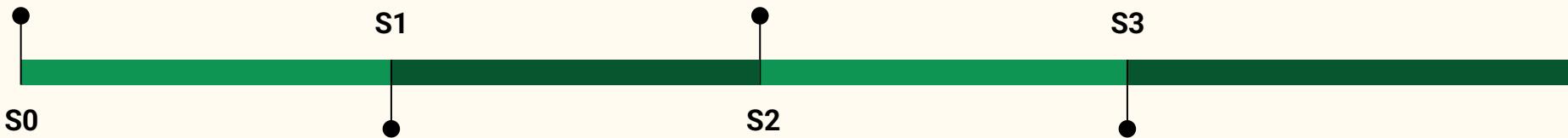
- Collision: -100
- Timeout: -50

Phase Blending:
 $r_{\text{total}} = \alpha(d) \times r_{\text{nav}} + (1 - \alpha(d)) \times r_{\text{park}}$
 where $\alpha(d) = d / 3.0$ (weight decreases as approach)

CURRICULUM LEARNING

Stage 0 - Aligned Spawn

- Aligned Start,
- Lot A
- 6 Bays
- Learn Basics and Park



Stage 2 - Lot A

- Stage 1 + Other side
- Spawns Along the road
- Random spawn
- Navigate the space and Park

Stage 1 - Lot A - Side A

- Add Cars in Other Bays
- Spawns along the road
- Random Spawn 22m max
- Navigate the space and Park

Stage 2 - Lot A + Lot B

- Stage 2 + Lot B
- Added Lot B, a t-shaped lot
- Random spawn along the roads
- Navigate the space and Park

FUTURE WORK & BROADER IMPACT

IMMEDIATE NEXT STEPS

- Complete curriculum, Expected: 10-30% success rate
- Sim-to-real transfer with TurtleBot3
- Multi-agent coordination

KEY LESSONS

- Hierarchical RL works (20x speedup)
- Curriculum essential (300x improvement)
- Reward engineering critical

REFERENCES

- <https://arxiv.org/html/2407.01333v1>
- https://github.com/Harish-Balaji-B/Ros2_humble_Autonomous_Parking_System_using_Turtlebot3_Waffle/
- <https://github.com/aatmaj28/Autonomous-Car-Parking-RL>
- <https://arxiv.org/abs/1707.06347>
- <https://arxiv.org/abs/1805.08296>
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