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## Task Hierarchical Control via Null-Space Projection and Path Integral Approach

Apurva Patil (UT Austin) Riku Funada (Institute of Science Tokyo) Takashi Tanaka (Purdue University) Luis Sentis (UT Austin)



#### Outline

What is Task Hierarchical Control?

Null-Space Projection

Proposed Approach

Simulation Results

Future Work



## Outline

What is Task Hierarchical Control?





Drone swarm: A Large degrees of freedom system



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  - 1. Each drone should avoid collision with obstacles
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How to achieve this?



#### Outline

**Null-Space Projection** 



## Null-Space Projection

- The top priority task is executed by employing all the capabilities of the system
- The second priority task is then applied to the null space of the top priority task
- The task on level three is then executed without disturbing the two higher-priority tasks, and so forth ...

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$$\dot{\sigma}_k = J_k(q)\dot{q}, \qquad J_k(q) = rac{\partial h_k(q)}{\partial q}$$







- Task 1: Avoid collision with the obstacle
- Task 2: Reach the goal

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## **Current Limitations**



- The individual controllers for the tasks are designed using simple low-level controllers such as PID controllers
- ► Desired signals  $\sigma_{i,d}(t)$  and control gains  $K_{p,i}, K_{d,i}$  are often chosen manually
- There is no systematic way to optimize the overall performance of the hierarchy of controllers



#### **Our Contributions**

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- We introduce a new framework for hierarchical task control that combines the null-space projection technique with the path integral control method
- This leverages Monte Carlo simulations for real-time computation of optimal control inputs
- This allows for the seamless integration of simpler PID-like controllers with a more sophisticated optimal control technique

## What is Path Integral Control?



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**Proposed Approach** 



## Proposed Approach



- Integration of null-space projection and path integral control
- In this example, the middle-priority task controller is replaced by a path integral controller

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- 2. Steer the platoon's centroid towards a goal position
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- Low-level controllers (such as PID) can't perform optimally
- What if we design an optimal controller using only the path integral method for each robot in the platoon?  $\Rightarrow$  scalability challenges
- We propose using low-level controllers for tasks 1 and 3, and apply the path integral controller to the more complex task 2
- The path integral controller optimizes the centroid's movement, while simpler tasks are handled by low-level controllers



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- It makes the path integral control scalable to complex systems with large DOFs
- It allows the hierarchical control architecture to seek the optimal control input
- It achieves a better sample efficiency than naïve applications of path integral controllers to high-DOF systems



## Outline

Simulation Results

- A unicycle system
- Two tasks to accomplish in the descending order of importance:
  - 1. obstacle avoidance
  - 2. move-to-goal



Figure: Results of single-agent example without the path integral controller









(b) Mean distance from goal  $\pm$  standard deviation

Figure: Results of single-agent example using the path integral controller

- A system of two unicycles
- > Three tasks to accomplish in the descending order of importance:
  - 1. obstacle avoidance
  - 2. steering the centroid of the robots toward a goal position
  - 3. maintaining a specific distance between the two unicycles



Figure: Results of two-agents example without the path integral controller



(a) Paths followed with path integral controller



(b) Mean distance from goal  $\pm$  standard deviation

Figure: Results of two-agents example using the path integral controller





(a) Without path integral controller: Distance between agents

(b) With path integral controller: Mean distance between agents  $\pm$  standard deviation

Figure: Distance between agents over time



#### Outline

Future Work

#### **Future Work**

Multi-agent red-team blue-team competition



**Figure:** Differential game extension. The red team attempts to maneuver past the blue team in order to occupy the target region. Meanwhile, the blue team maneuvers in order to block the red team.

# Questions?

apurvapatil@utexas.edu # patil-apurva.github.io/portfolio/

