

IROS 2025 Workshop: Advancements in Aerial Physical Interaction

Safety-Critical Aerial Physical Interaction

Postdoc. Jeonghyun Byun

Laboratory for Autonomous Robotics Research (led by Professor H. Jin Kim)
Autonomous Systems and Research Institute
Seoul National University

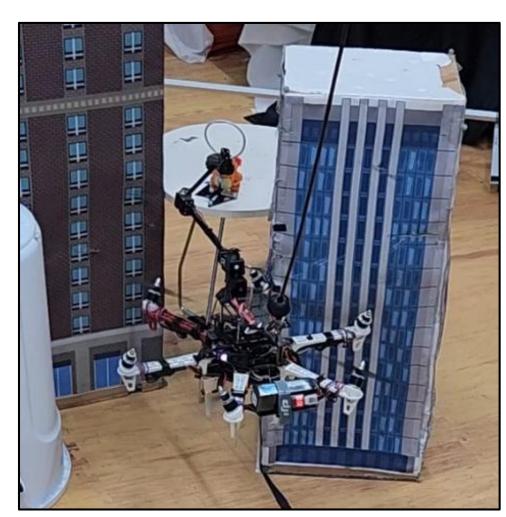




Motivation

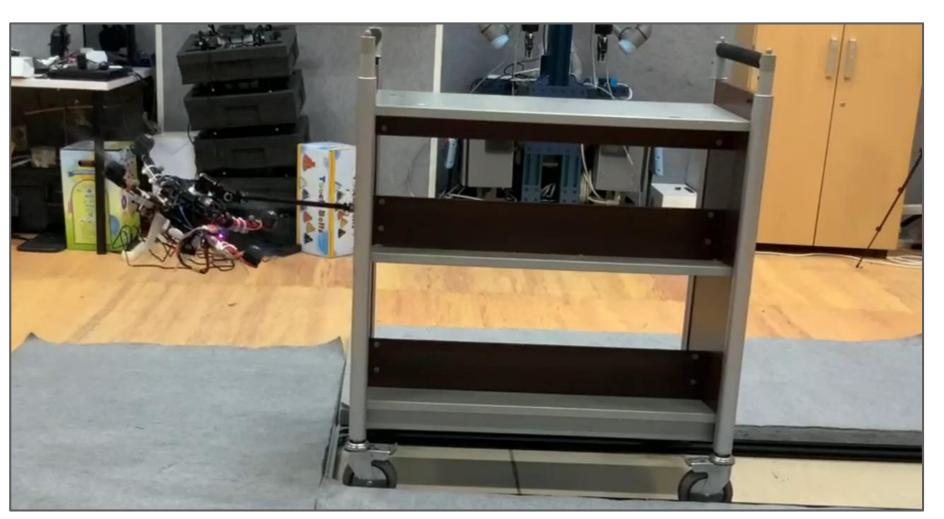
Safety in Aerial Physical Interaction

• Well-known robot safety: Collision avoidance



[Video] Failure case due to the collision [1]

Specific safety for Aerial Physical Interaction:
 Motor saturation avoidance



[Video] Failure case due to motor saturation [2]

• Other safety constraints: Power flow limit [3], speed limit, compliance, etc.

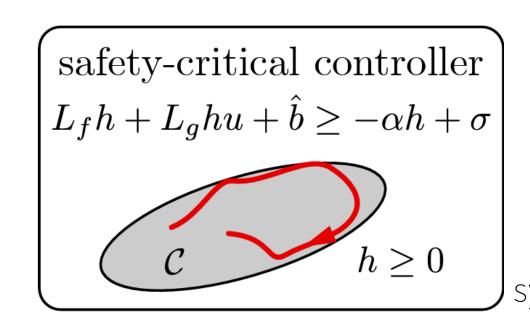




Motivation

Safety-" Critical" Aerial Physical Interaction

- Safety-critical control
 - ✓ Controller with mathematical proven safety constraints.
 - ✓ Examples: Reachable Forward Set (RFS) [4], Control Barrier Function (CBF) [5], Model Predictive Controller (MPC) [6], etc ···
 - ✓ Our choice: Control Barrier Function
 - > Fast computation with quadratic programming (QP)-based optimization with linear inequalities
 - > No need for explicit physical interaction model
 - > Rigorous guarantee on system safety and dynamic feasibility



[Fig] Illustration of the formal behavior of a system under safety-critical control framework. [5]



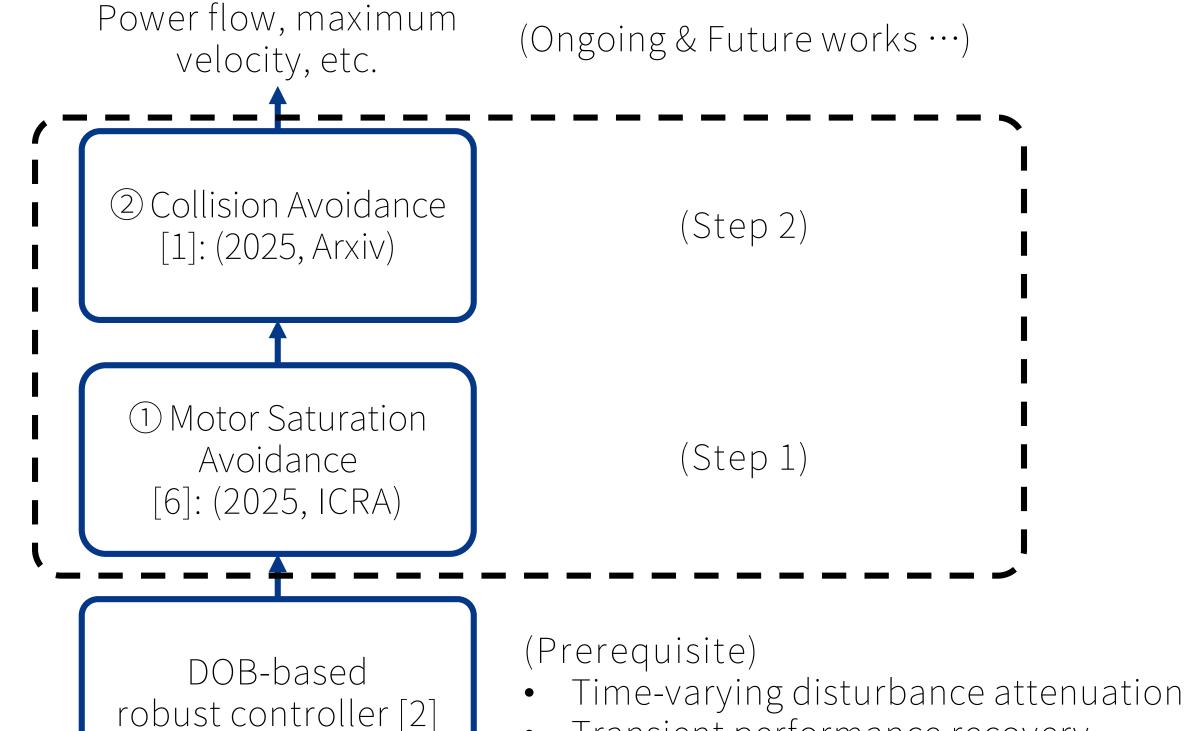


Motivation

Steps Towards Safer Aerial Physical Interaction

Safety-critical control leveraging prestabilization ability of existing robust control framework

Cornerstone



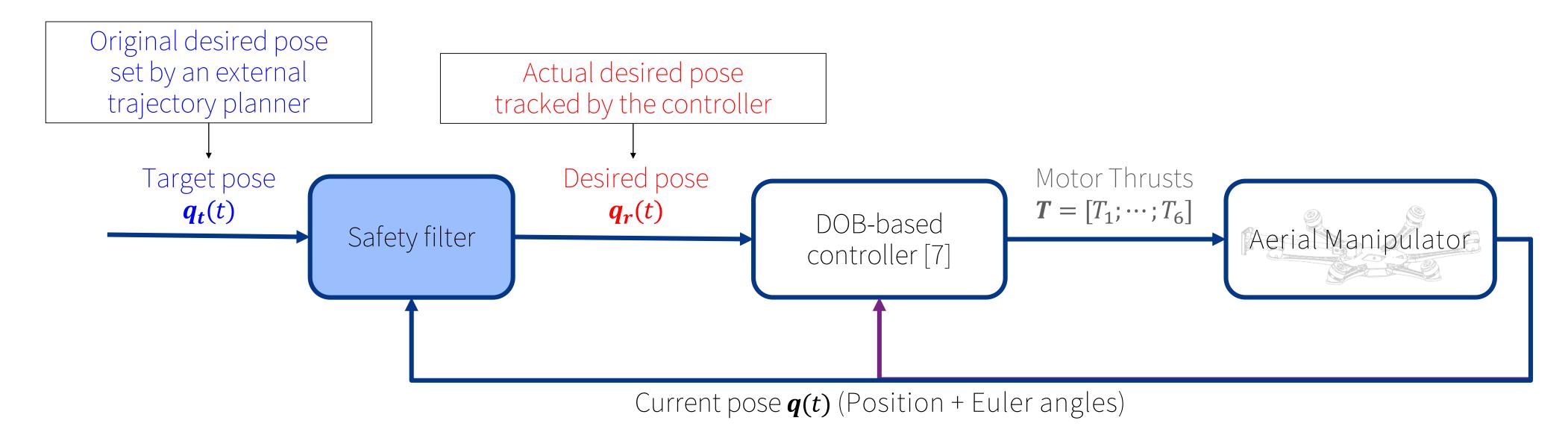




Transient performance recovery

Motor Saturation-Aware Safety-Critical Controller

- Safety filter (Proposed) + DOB-based controller [7]
 - Divide the entire control law into the outer-loop (safety filter) and inner-loop (DOB-based controller) to leverage high performance of the DOB-based controller







Motor Saturation-Aware Safety-Critical Controller

Safety filter (Proposed)

① Integrated system of control framework & nominal dynamics

$$\dot{x} = f(x) + g\ddot{q}_r + w(\tilde{d})$$

- $x \in \mathbb{R}^{36}$: Current pose & twist / desired pose & twist / DOB variables
- $w(\tilde{d})$: Model uncertainty arisen by imperfect disturbance attenuation

② Motor saturation CBFs: "T = T(x)"

$$h_{T,i}(x) = \left(\frac{T_{\max} - T_{\min}}{2}\right)^2 - \left(T_i(x) - \frac{T_{\max} + T_{\min}}{2}\right)^2$$
, $i = 1, \dots, 6$

• Formal formulation of robust CBF-QP [5]

$$\min_{\ddot{\boldsymbol{q}_r}} \|\ddot{\boldsymbol{q}_r} - \ddot{\boldsymbol{q}_t}\|^2$$
 $s.t.$ $\sigma_{T,1} - \hat{\beta}_{T,1}(\boldsymbol{x}) \leq \mathcal{L}_f h_{T,1}(\boldsymbol{x}) + \mathcal{L}_g h_{T,1}(\boldsymbol{x}) \ddot{\boldsymbol{q}_r} + \gamma_{T,1} h_{T,1}(\boldsymbol{x})$ \vdots
$$\sigma_{T,6} - \hat{\beta}_{T,6}(\boldsymbol{x}) \leq \mathcal{L}_f h_{T,6}(\boldsymbol{x}) + \mathcal{L}_g h_{T,6}(\boldsymbol{x}) \ddot{\boldsymbol{q}_r} + \gamma_{T,1} h_{T,6}(\boldsymbol{x})$$
 Estimations of Nominal CBF constraint: $\boldsymbol{w}(\boldsymbol{d})$ -related $0 \leq h_{T,i}(\boldsymbol{x}, \dot{\boldsymbol{x}}) + \gamma_{T,i}(\boldsymbol{x}) h_{T,i}(\boldsymbol{x})$ Positive parameters for robustness

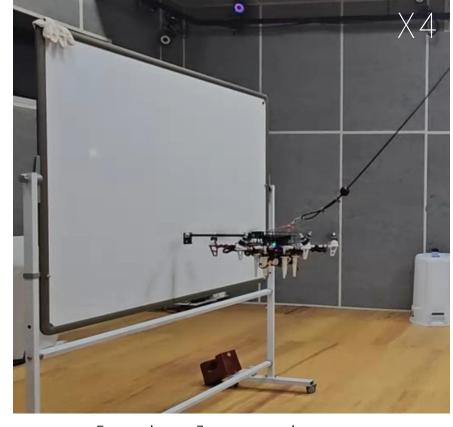


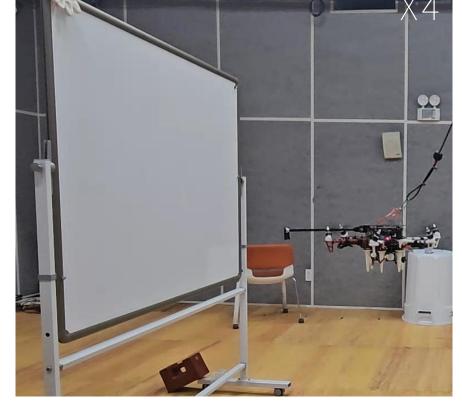


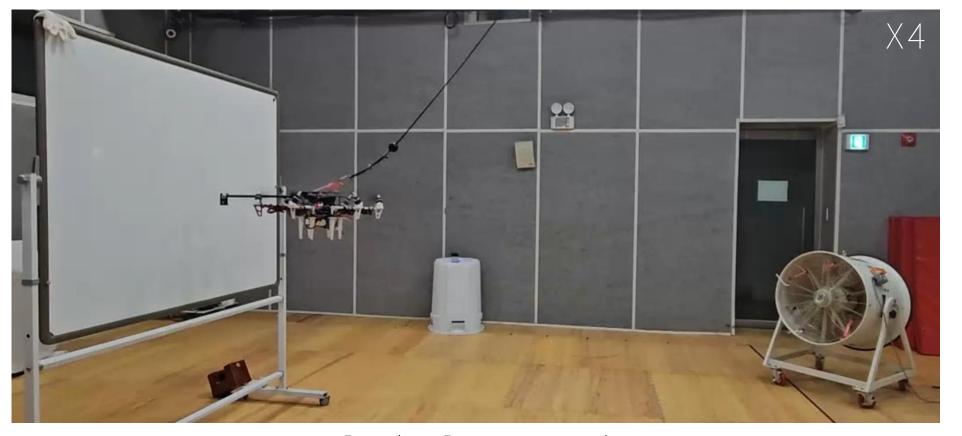
Comparative Experiments

- What if the target pose is located in unreachable regions?
 - 1) Pushing a static wall: Target pose beyond the wall

	Method	Remarks	
Baseline 1	DOB + Thrust clipping	$T_i = \min(\max(T_{d,i}, T_m), T_M)$	
Baseline 2	DOB + Thrust adjustment by CBF	Control affine system with $\dot{\boldsymbol{T}}$ as an input	
Proposed	DOB + Reference adjustment by CBF	Control affine system with \ddot{q}_r as an input	







[Video] Baseline 1

[Video] Baseline 2

[Video] Proposed

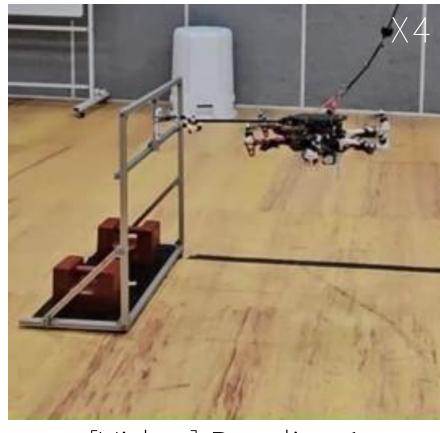


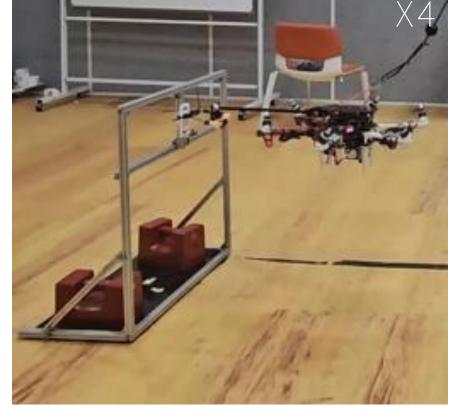


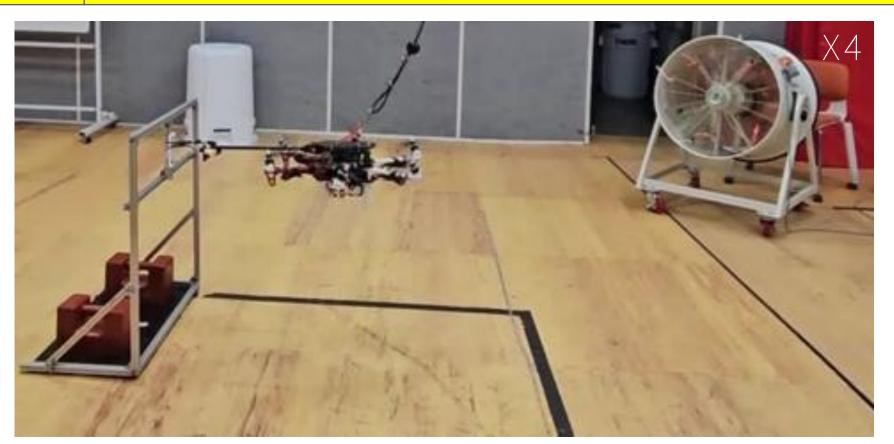
Comparative Experiments

- What if the target pose is not reachable?
 - 2) Pulling a firmly attached plug: Target pose is located away from the socket in the pulling direction.

	Method	Remarks	
Baseline 1	DOB + Thrust clipping	$T_i = \min(\max(T_{d,i}, T_m), T_M)$	
Baseline 2	DOB + Thrust adjustment by CBF	Control affine system with $m{\dot{T}}$ as an input	
Proposed	DOB + Reference adjustment by CBF	Control affine system with \ddot{q}_r as an input	







[Video] Baseline 1

[Video] Baseline 2

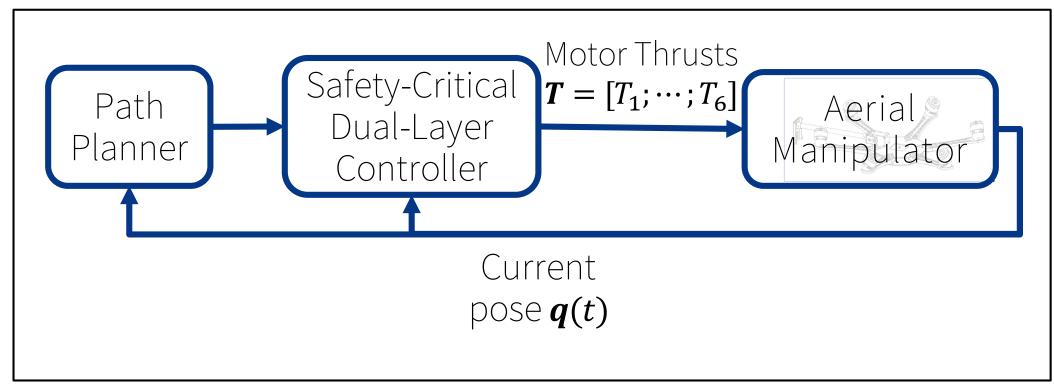
[Video] Proposed



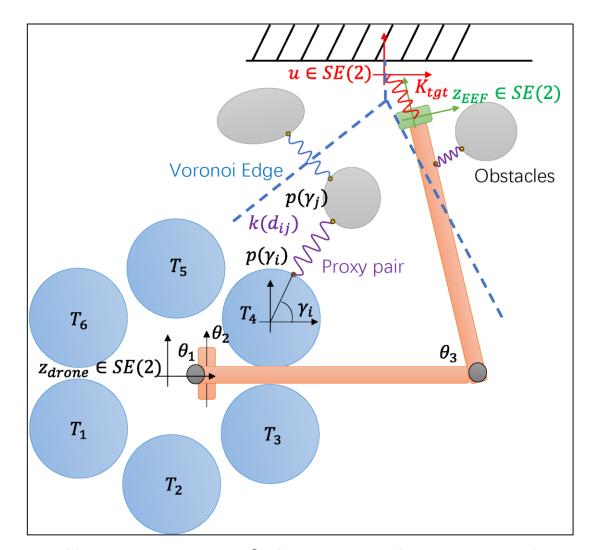


Planning and Control for Collision Avoidance

- Objective
 - ✓ Passing through a narrow gap that can only be traversed by the thin linkages of the robot arm
 - ✓ Still, avoiding motor saturation



[Fig] Controller Diagram



[Fig] Illustration of the aerial manipulator's end-effector reaching its goal position [2]



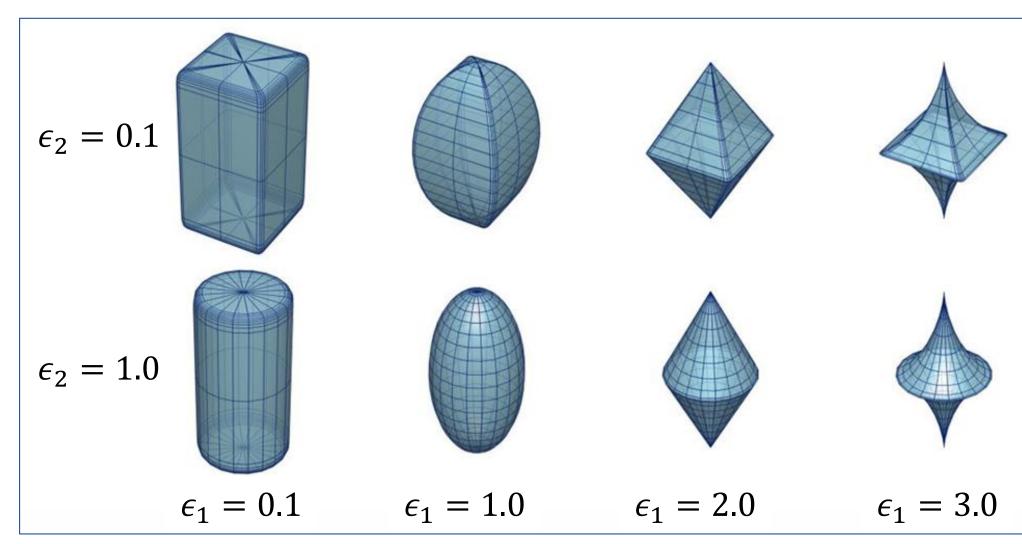


Geometric Representation of Vehicle & Obstacles

- Superquadrics (SQs): Reduced conservativeness
 - ✓ Implicit equation [8]

$$F(x,y,z) = \left(\left(\frac{x}{a_1}\right)^{2/\epsilon_2} + \left(\frac{y}{a_2}\right)^{2/\epsilon_2}\right)^{\epsilon_2/\epsilon_1} + \left(\frac{z}{a_3}\right)^{2/\epsilon_1}$$
If F > 1 (outside), F=1 (on), F < 1 (inside)

- Advantages
 - Compact parameterization: Low Storage requirements & Low computational overhead
 - Wide shape variability: Broad range of shapes Near-spherical to boxlike.
 - Smooth surfaces: Continuously differentiable (\it{C}^{1}) surfaces
 - → Ideal for gradient-based computations and optimization



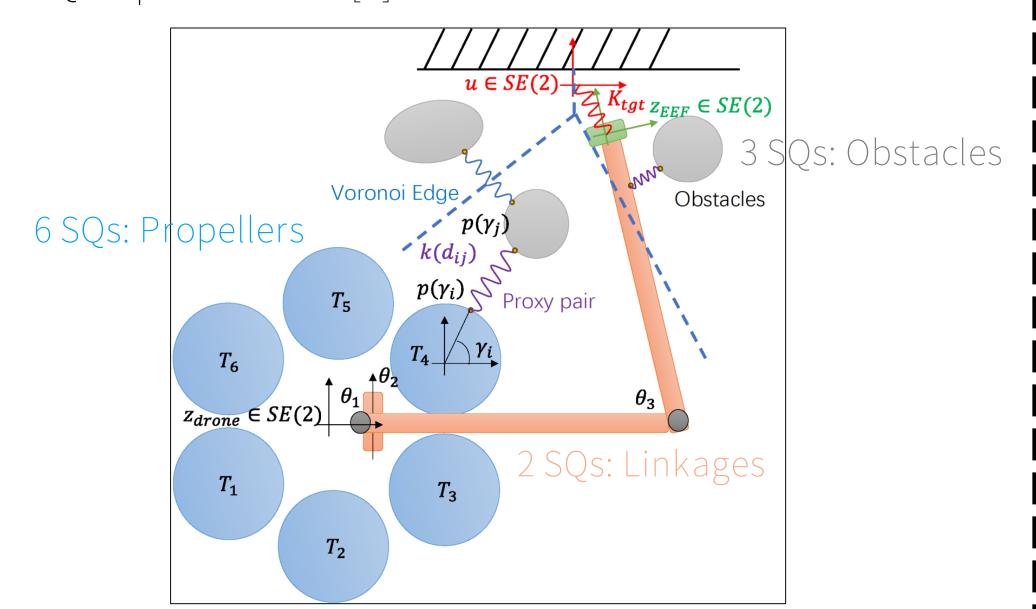
[Fig] Illustration of wide shape variability of superquadrics [8].





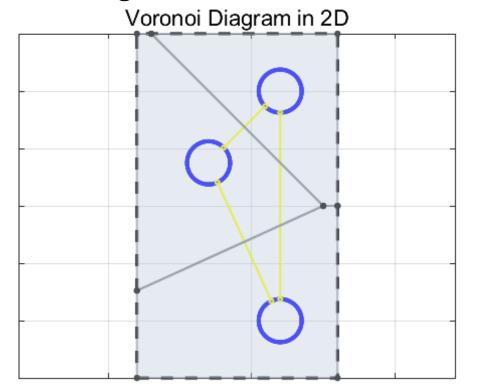
Whole-Body Path Planner

SQs representation [1]



[Fig] SQ representations for our hexarotor-based aerial manipulator with 2-link robot arm and obstacles

- Maximum Clearance Whole-Body Path Planner [2]
 - 1) Voronoi Diagram based on obstacle SQs



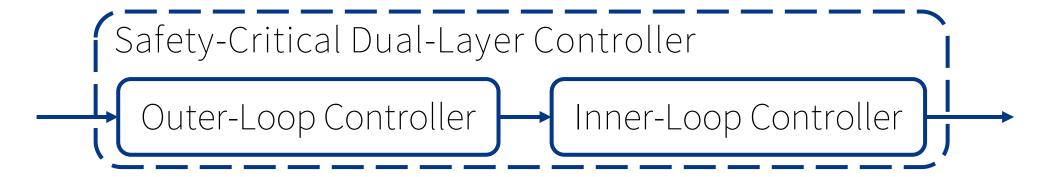
- 2) Generate path on equilibrium manifold [9]
 - → <u>Potential function</u>-based method attracting pose of the end-effector

Collision avoidance potential + Goal-attracting potential





Safety-Critical Dual-Layer Control Architecture

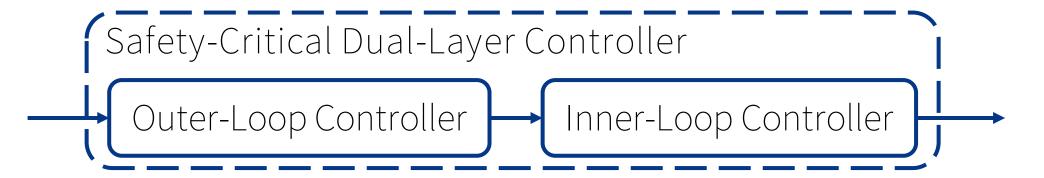


- 1) Outer-Loop Controller (Robust CBF-QP)
 - Constraint 1: Motor saturation-aware CBF [6]





Safety-Critical Dual-Layer Control Architecture



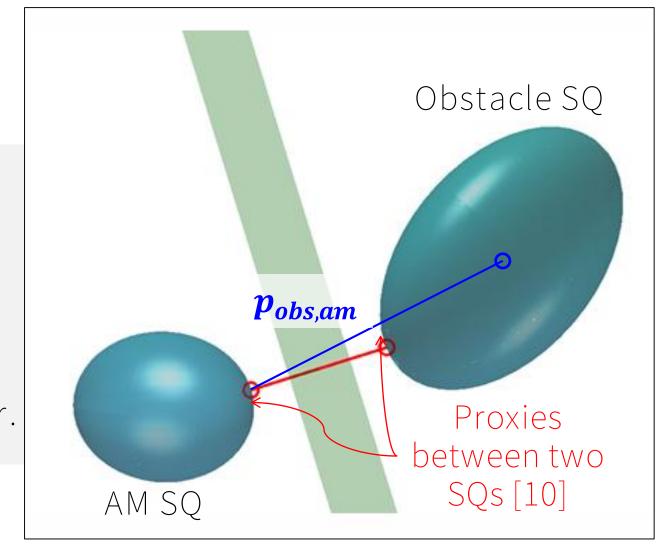
- 1) Outer-Loop Controller (Robust CBF-QP)
 - Constraint 1: Motor saturation-aware CBF [6]
 - Constraint 2: SQ Distance-based CBF [1]

$$\rightarrow h_{co} = \log \left(F(\mathbf{p}_{obs,am}) \right)$$

 $F(x,y,z) = \left(\left(\frac{x}{a_1}\right)^{2/\epsilon_2} + \left(\frac{y}{a_2}\right)^{2/\epsilon_2}\right)^{\epsilon_2/\epsilon_1} + \left(\frac{z}{a_3}\right)^{2/\epsilon_1}$ If F > 1 (outside), F=1 (on), F < 1 (inside)

 \rightarrow Why? $F(\cdot)$ rapidly increases as the distance becomes larger.

$$\rightarrow \ddot{h}_{co} + \gamma_{co}\dot{h}_{co} + \gamma_{co}^2h_{co} \ge -\hat{\beta}_{co} + \sigma_{co} \rightarrow$$
Instantly avoid collision upon path following error.

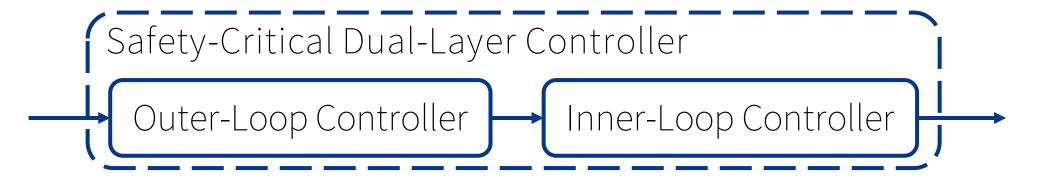


[Fig] Illustration of the proxies and distance between them [10]





Safety-Critical Dual-Layer Control Architecture



- 1) Outer-Loop Controller (Robust CBF-QP)
 - Constraint 1: Motor saturation-aware CBF [6]
 - Constraint 2: SQ Distance-based CBF [1]

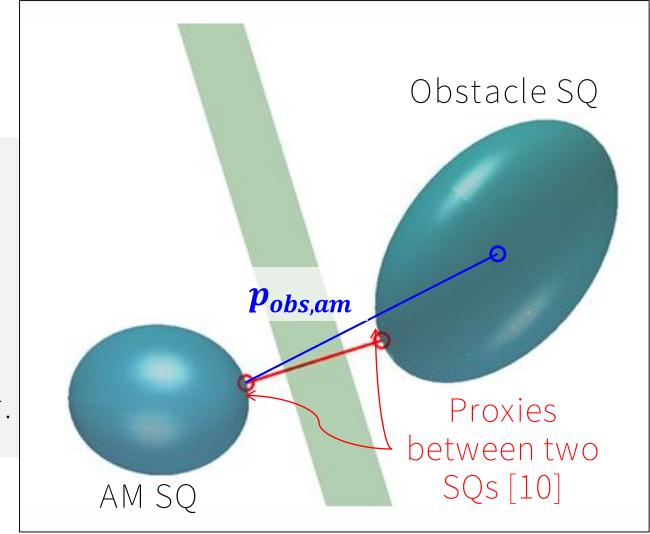
$$\rightarrow h_{co} = \log \left(F(p_{obs,am}) \right)$$

$$\rightarrow \text{Why? } F(\cdot) \text{ rapidly increases after 1.}$$

 $F(x,y,z) = \left(\left(\frac{x}{a_1}\right)^{2/\epsilon_2} + \left(\frac{y}{a_2}\right)^{2/\epsilon_2}\right)^{\epsilon_2/\epsilon_1} + \left(\frac{z}{a_3}\right)^{2/\epsilon_1}$ If F > 1 (outside), F=1 (on), F < 1 (inside)

o $\ddot{h}_{co} + \gamma_{co}\dot{h}_{co} + \gamma_{co}^2h_{co} \geq -\hat{\beta}_{co} + \sigma_{co} \rightarrow$ Instantly avoid collision upon path following error.

2) Inner-Loop Controller: DOB-based controller [2]



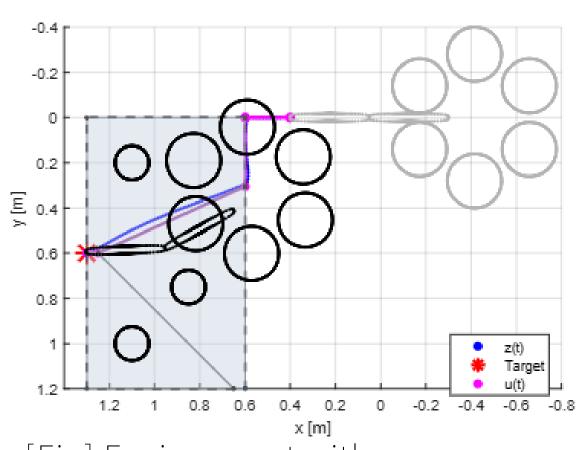
[Fig] Illustration of the proxies and distance between them [10]



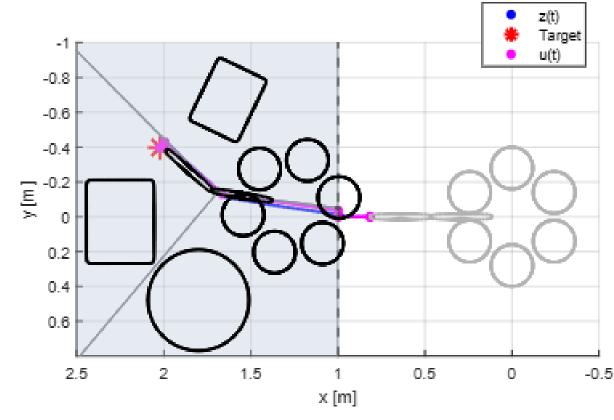


Path Planning Results in Simulation

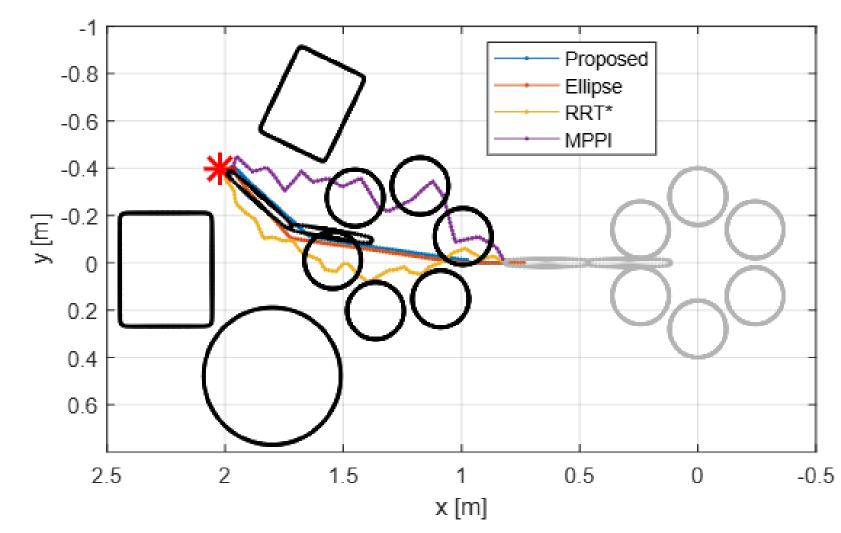
• Picking a target object [1]



[Fig] Environment with a narrow gap



[Fig] Environment with different obstacle shapes



Method	Time	Min-distance	Arc-length	Jerkiness
Proposed	0.128	0.0644	1.375	0.0016
Ellipse	0.113	-0.0049	1.383	0.0001
RRT*	105.8	0.0314	1.565	0.0032
MPPI	0.552	0.0039	1.721	0.0057

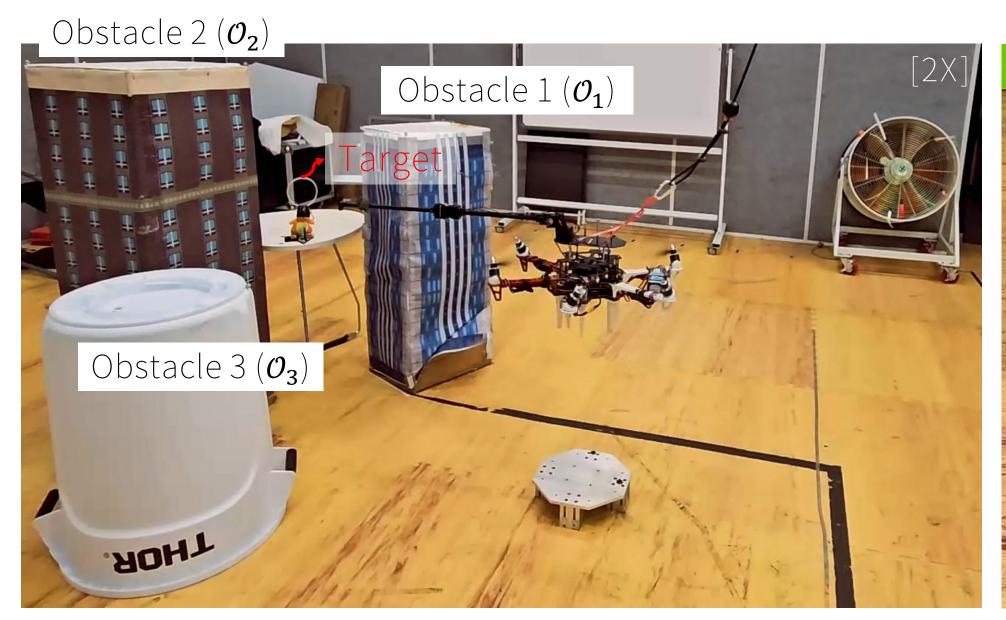
[Fig] Comparative simulations



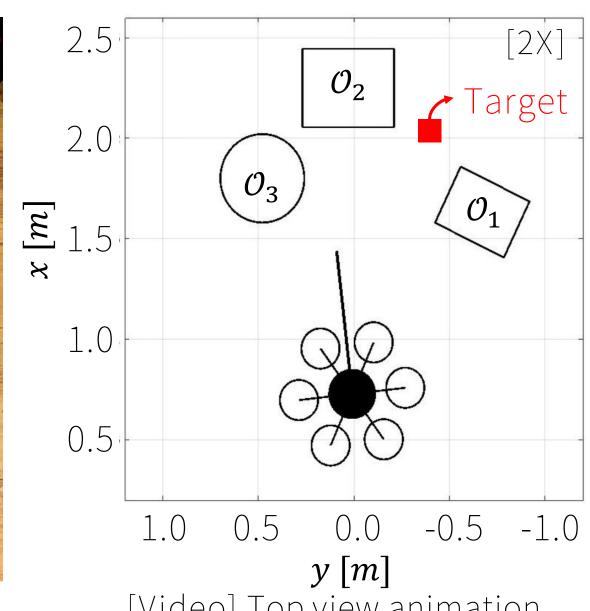


Experimental Result

• Picking a target object [1]



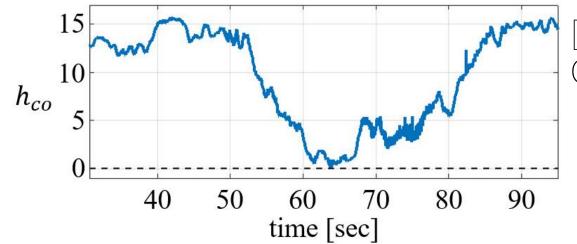




[Video] Side view video

[Video] Front view video

[Video] Top view animation



[Fig] History of collision-avoidance CBF with its minimum value of 0.115

[1] L. Yang, J. Lee, D. Campolo, H. J. Kim, and J. Byun, "Whole-body motion planning and safety-critical control for aerial manipulation, "arXiv preprint arXiv:2511.02342, 2025.





Conclusion

Takeaways

- Can achieve strict enforcement of safety constraints through a CBF-based formulation built on the existing control law.
- Can realize effective collision avoidance in cluttered environments by combining superquadrics geometric representations.

Future Directions

- Further explore stronger mathematical guarantees on safety.
- Find sweet spot between stability and safety.
- Implement and compare existing safety-critical controllers to aerial physical interaction.







IROS 2025 Workshop: Advancements in Aerial Physical Interaction

Thank you

E-mail: quswjdgus97@snu.ac.kr

LinkedIn: 🗖



Special thanks to the collaborators:

Professor H. Jin Kim (SNU, Korea)

Professor Domenico Campolo (NTU, Singapore)

Dr. Dongjae Lee (CMU, USA)

Mr. Lin Yang (NTU, Singapore)

Mr. Dohyun Eom (SNU, Korea)

Mr. Yeonjun Kim (SNU, Korea)

Mr. Jinwoo Lee (SNU, Korea)



