

Technische Universität München

Towards Connecting Control to Perception: High-Performance Whole-Body Collision Avoidance Using Control-Compatible Obstacles

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PROBLEM

Collision avoidance is a crucial part of safe manipulation Requirements (R):

- Self-collision avoidance (see Fig. 1) \rightarrow R1: Whole-body awareness
- Environment collision avoidance (see Fig. 1) \rightarrow R2: Environment awareness (representation) \rightarrow R3: Environment perception (update procedure)
- Flexibility to react safely to unexpected events (see Fig. 2) \rightarrow R4: Real-time capability
- Adaptability, e.g., for keeping the ability to manipulate and avoid collisions with grasped objects \rightarrow R5: Connection to a knowledge base



Fig. 1: The taxonomy of the collision types that must be considered in collision avoidance for safe manipulation Collision types can be divided into two groups: (1) self-collisions and (2) environment (env.) collisions.

Fig. 2: A robot without a real-time environment-aware collision avoidance control is unable to react safely to unexpected events such as an unavoidable human-robot collision. This may lead to further unwanted events, as shown in this figure.



EXPERIMENTAL RESULTS / PIPELINE



Fig. 4: The visualization of our pipeline for connecting control to perception, captured during the experiments. (a) shows the environment replica stored in the digital twin and constantly updated through a multimodal sensory environment perception. Based on this, our new Replica Abstraction Layer (RAL) displayed in (b) creates control-compatible environment primitives for the currently reachable objects. The environment primitives are complemented by those of the robot in 1 kHz real time by our collision avoidance control shown in (c). It then calculates distances between all possible collision pairs (green lines) and determines virtual Cartesian repulsive forces along the connecting distance lines for those (black lines) below a certain threshold (we used 6 cm). The Cartesian forces are transformed into joint torques τ_{Rep} , which are then applied to the real-world robot in (d).

DISCUSSION AND FUTURE WORK



- Average computation time of 7.45 µs in the scenario of Fig. 4c
 - 4 spheres, 12 capsules, and 3 planes result in 126 collision pairs
- Computation time scales with the number of primitives (Fig. 5)

Limitations and future work:

- No globally optimized motions
- Local minima may exist that can lead to convergence issues
- **Combine with a Cartesian space** \rightarrow collision-free motion planning (Table I)

TABLE I: Decision matrix showing why an APF-based environment-aware collision avoidance control is useful and with what of other collision avoidance

	Method Feature	CBF-based self-collision avoidance control	CBF-based envaware collision avoidance control	APF-based self-collision avoidance control	APF-based envaware collision avoidance control	Cartesian space collision-free motion planning	Joint space collision-free motion planning
	Whole-body collision avoidance	\checkmark	\checkmark	\checkmark	\checkmark	×	\checkmark
	Real-time capability	×	×	\checkmark	\checkmark	×	×
tives. se all come robot	Handling of unexpected scenarios	×	×	×	\checkmark	×	×
	Globally optimized paths	×	×	×	×	\checkmark	×
	Local minima-free and convergent	\checkmark	\checkmark	×	×	\checkmark	\checkmark

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	Glossary				
Er	nv.	Environment			
R	AL	Replica Abstraction Layer			
AF	PF	Artificial Potential Field			
CI	BF	Control Barrier Function			