



ETH zürich

EPFL

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Advisor and Co-funder at Qolo Inc.**



Metrics and Safety for Autonomous Robot Navigation



Disclaimer: DP is a shareholder of Qolo Inc.



Passive exoskeleton technology

Supporting sit-to-stand and stand-to-sit transitions



Qolo

Quality of Life
with Locomotion



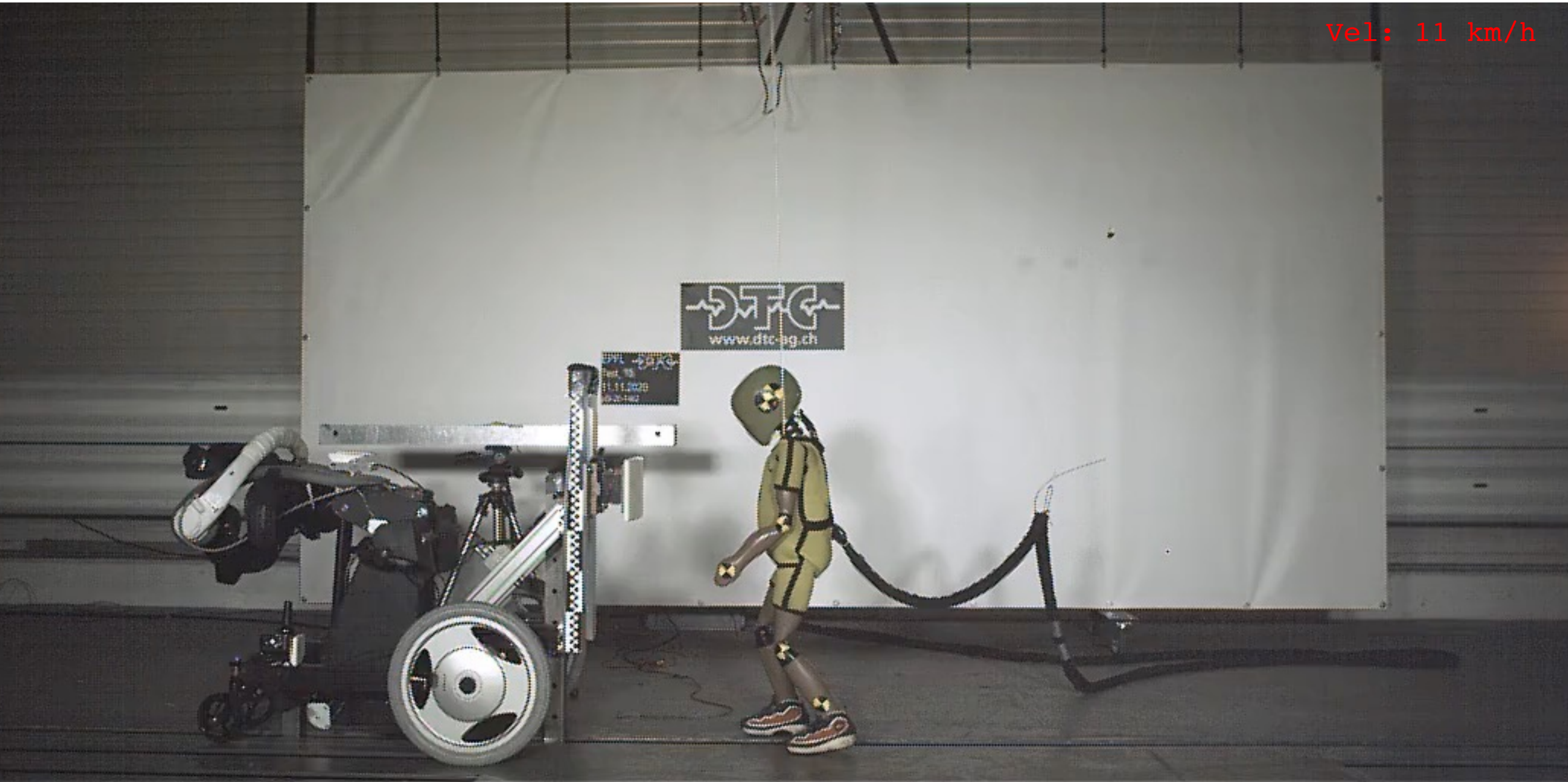
Robots – Could be a risk for humans

Vel: 11 km/h

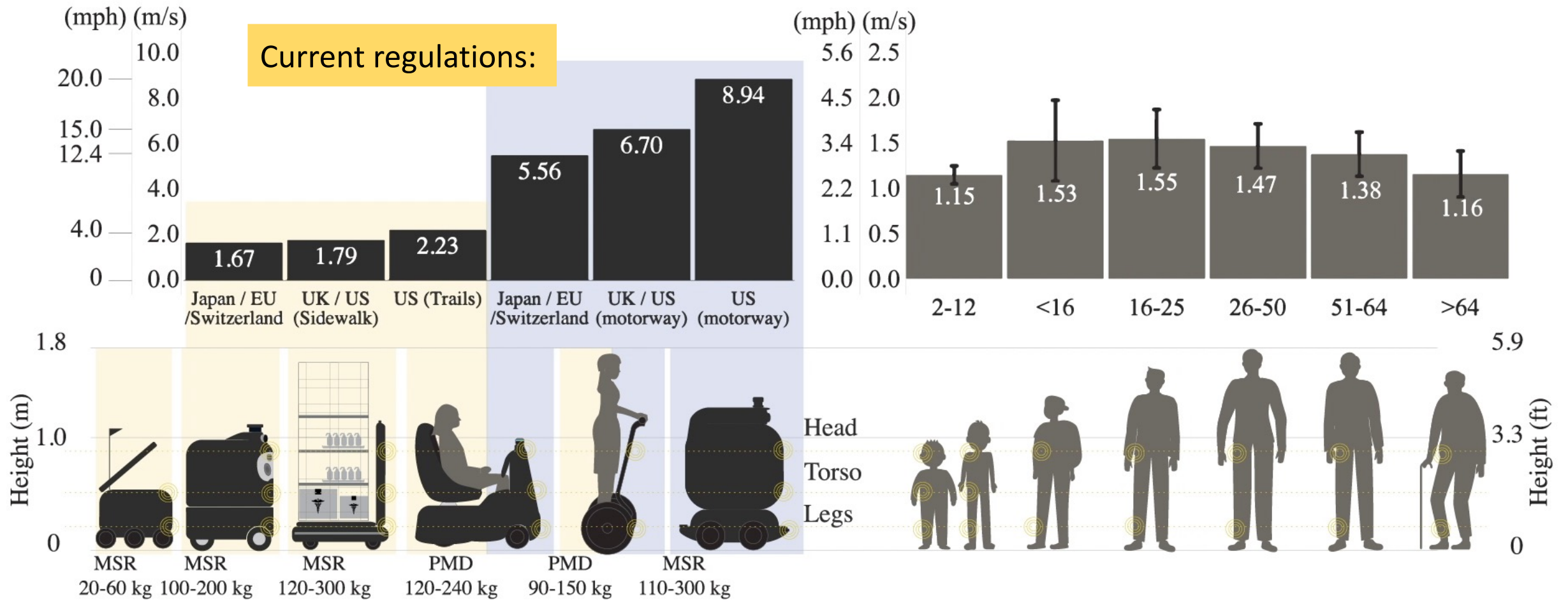


Robots – Could be a risk for humans

Vel: 11 km/h



Robots – Could be a risk for humans





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CROWDBOT

Safety Metrics in Social Navigation

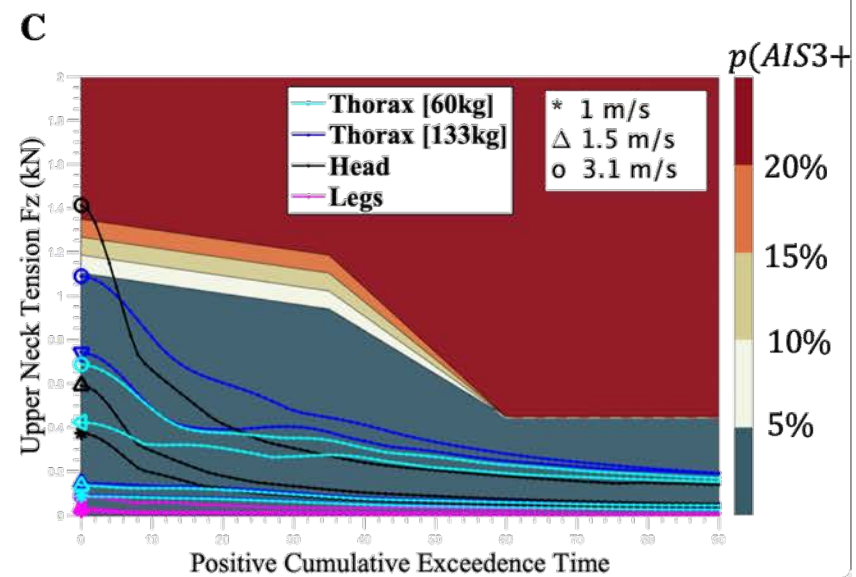
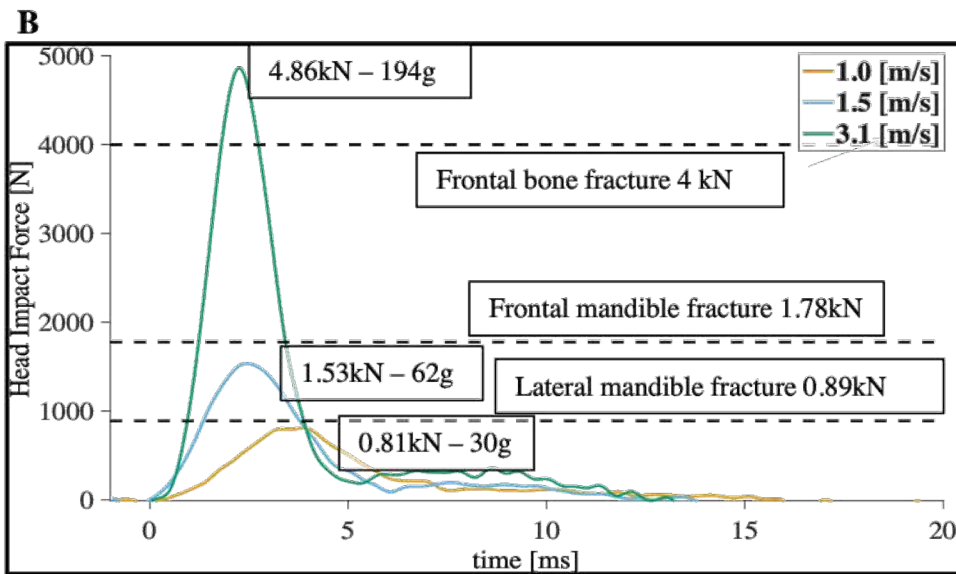
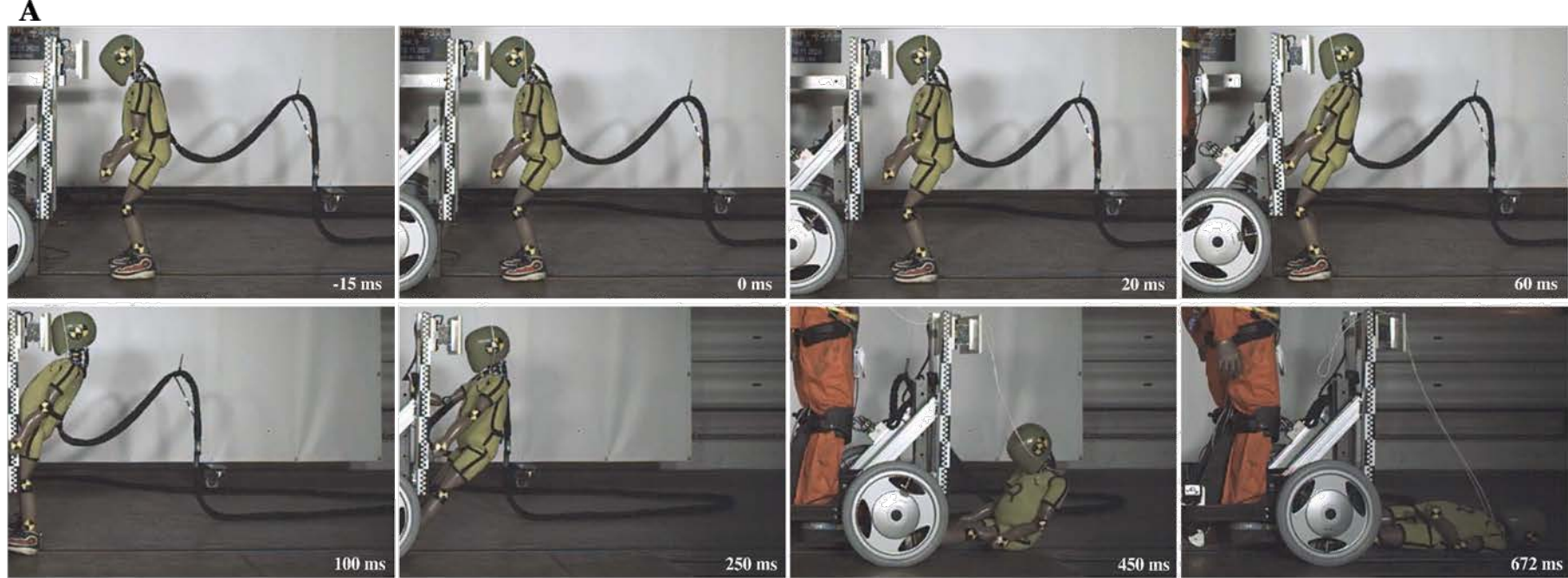


Chest Impact with Child Dummy Q3

Robot weight: 133 kg

Speed: 3.1 m/s (11 km/h) (6.9 mph)

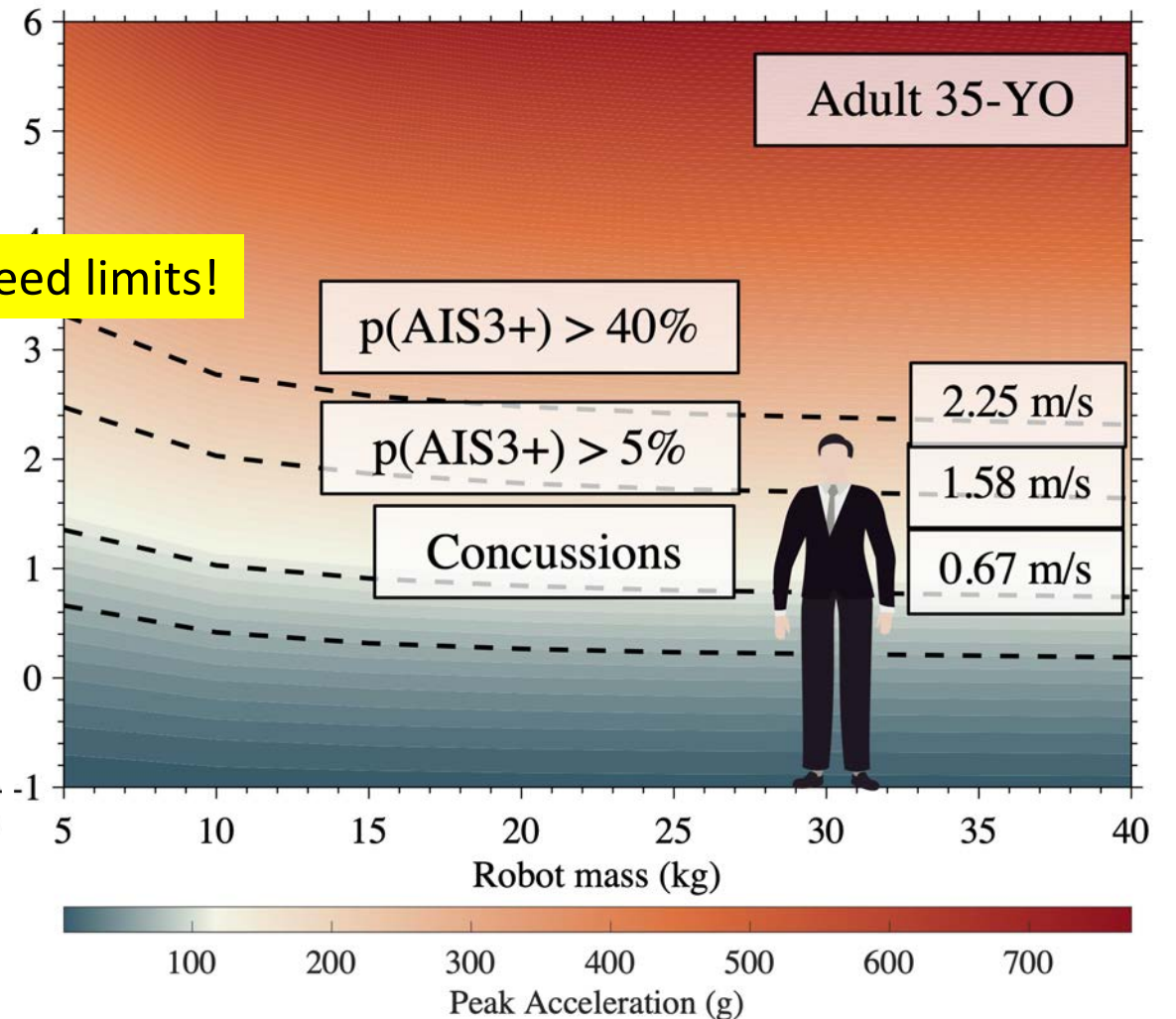
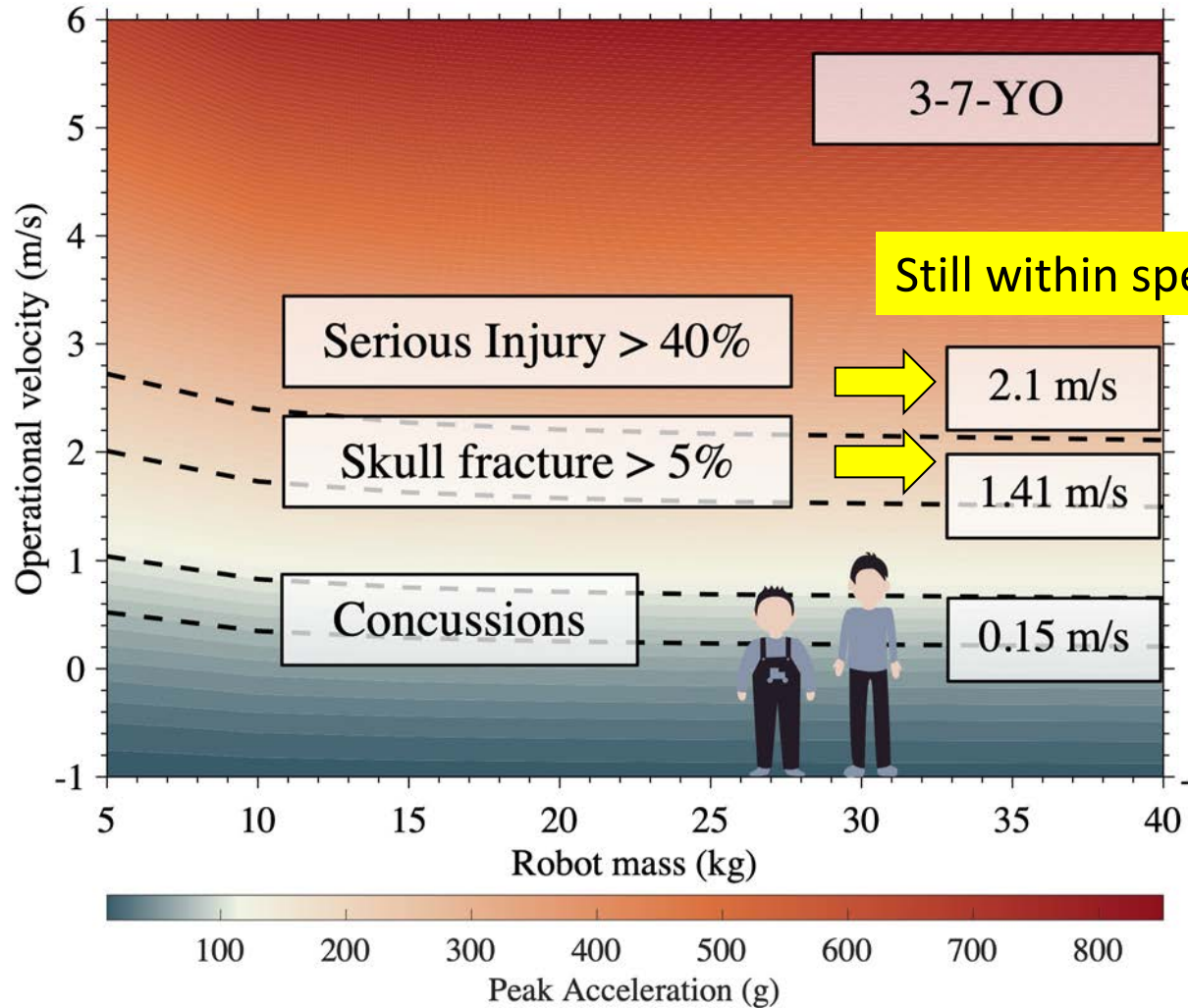




Paez-Granados, D., & Billard, A. (2022). Crash test-based assessment of injury risks for adults and children when colliding with personal mobility devices and service robots. *Nature Scientific Reports*, 12(5285), 1–13. [DOI:10.1038/s41598-022-09349-9](https://doi.org/10.1038/s41598-022-09349-9)

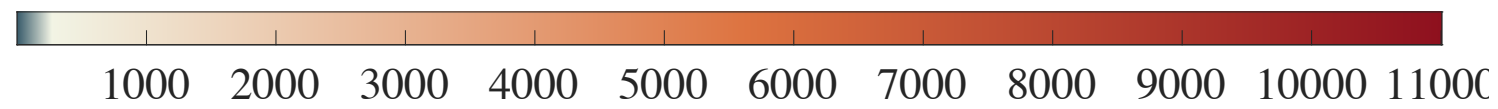
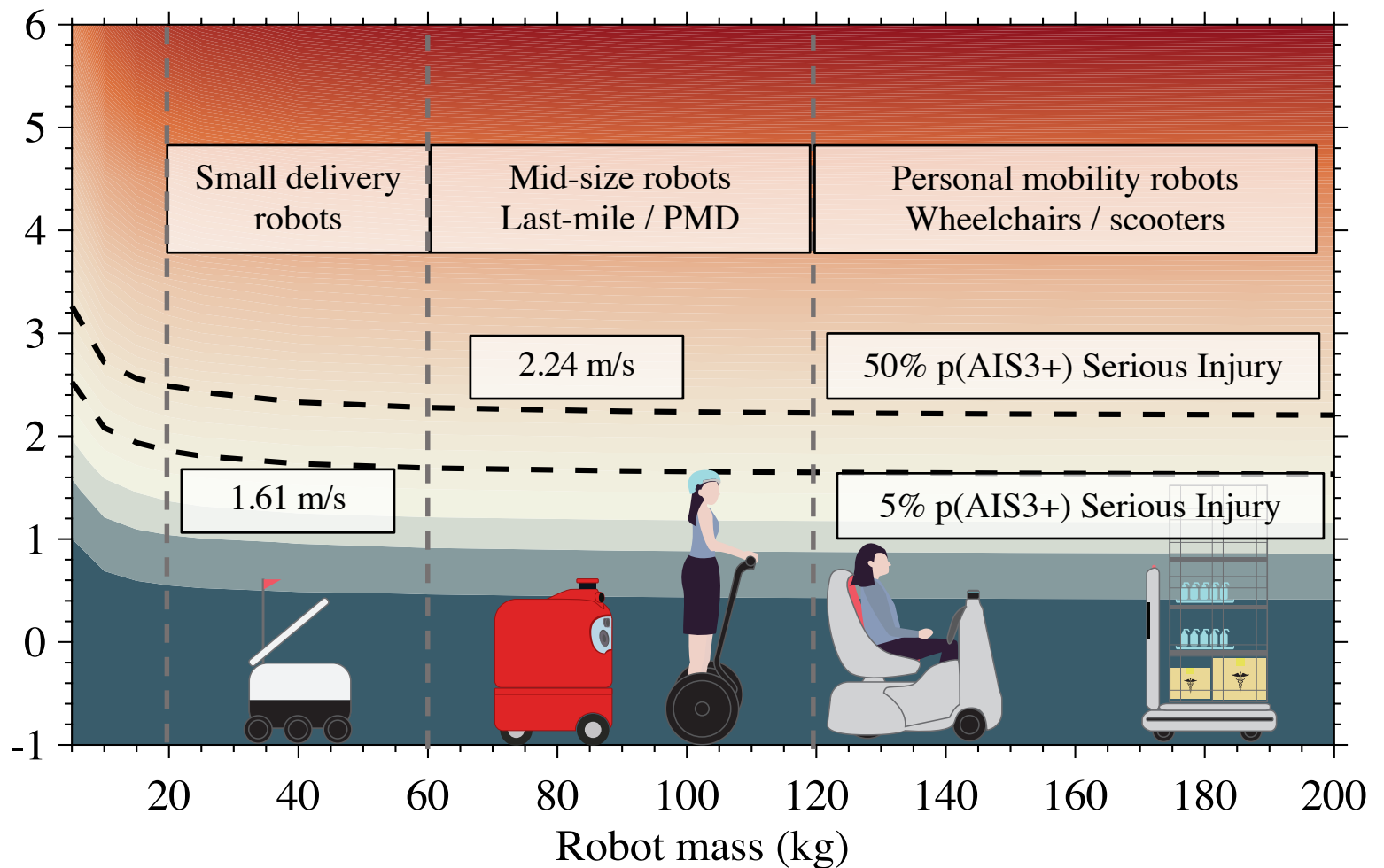


Robots – Could be a risk for humans



*Accounting for a pedestrian speed of 1.5 m/s

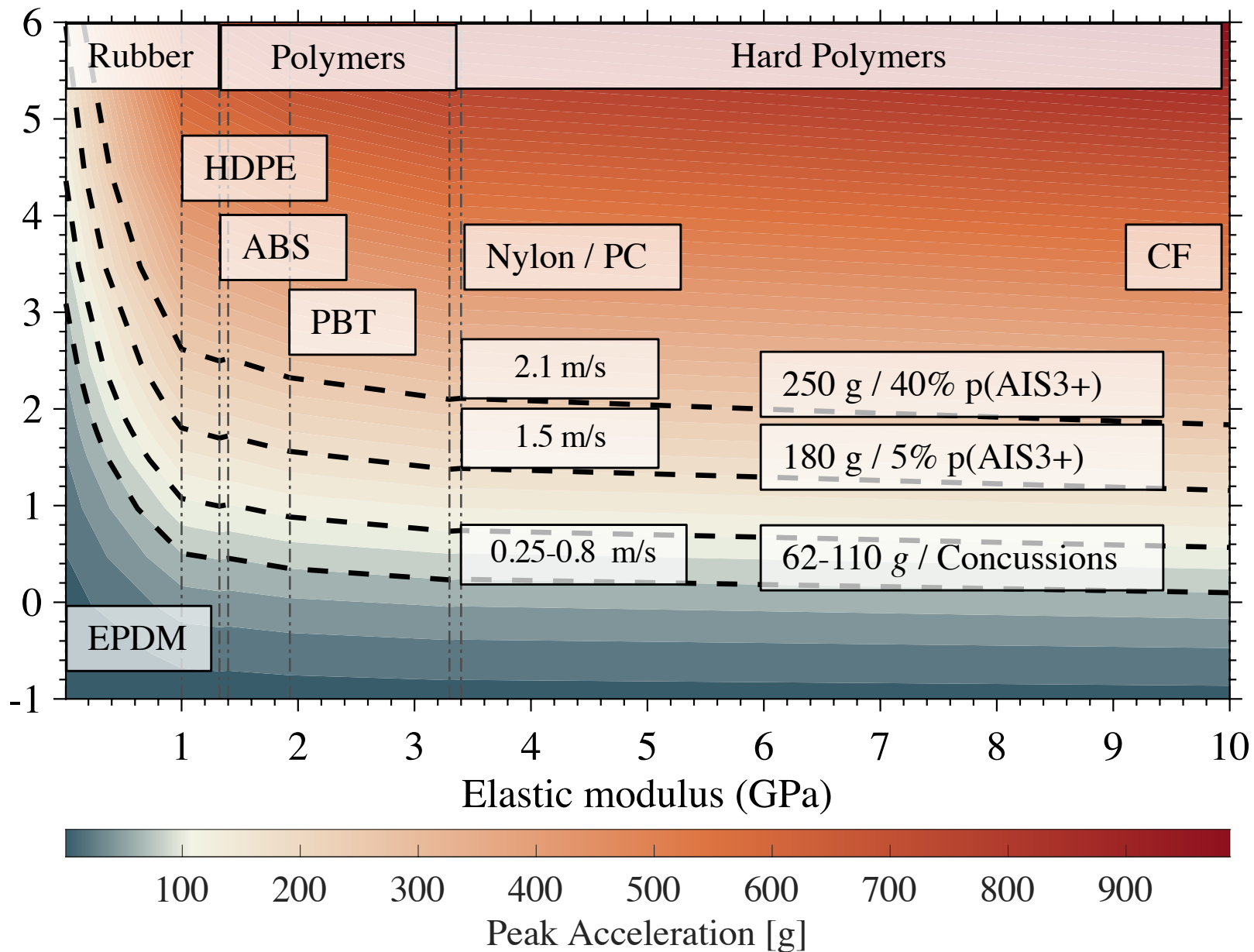
Robots – Could be a risk for humans



HIC

- Starship, USA
- DeliRo - ZMP
Japan
- i2SE - Segway
Inc., Bedford,
NH, USA
- RakuRo, ZMP,
Japan
- TUG - Aethon,
Pittsburgh, USA

Robots – Could be a risk for humans



Secondary Injuries from Ground Impact Child and Adult
Robot weight: 133 kg

- You could limit the speed.
- You could reduce the mass.
- You could use compliant hulls.

But most importantly, you do want to avoid collisions!

To avoid collisions, robots need to understand their environments.





STARBUCKS COFFEE



CLER

Bershka

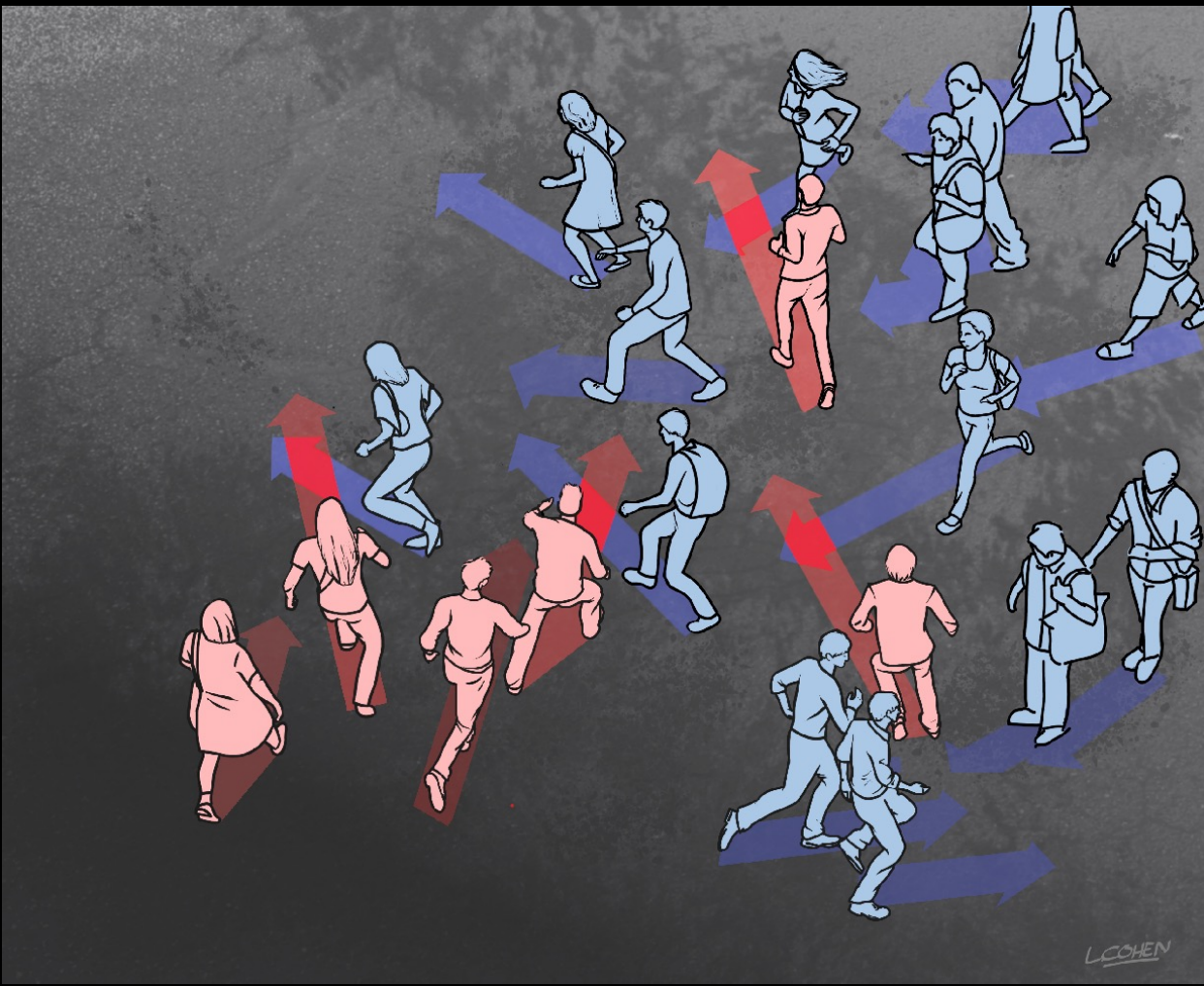
CLER
Bershka

COOP

OCHSMER SHIP

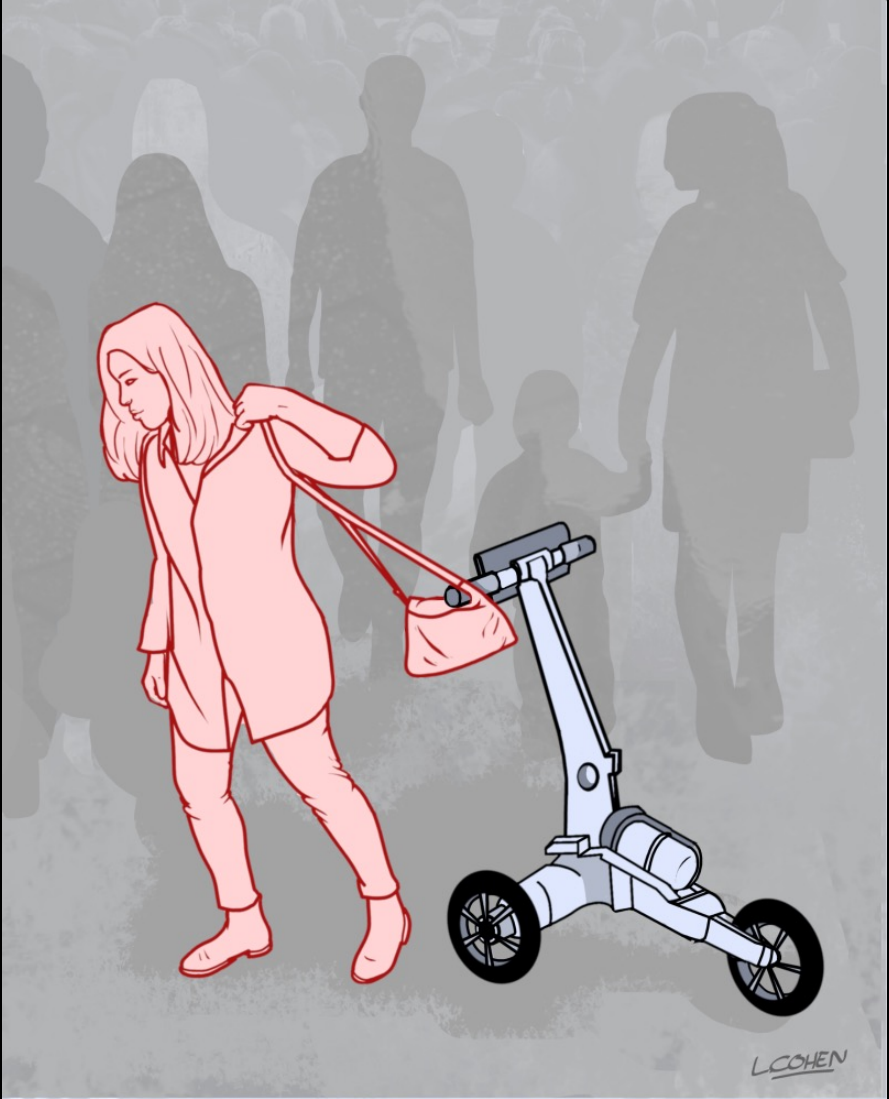
COOP

Understanding the crowd!



Smart robots that can interact with humans safely

Model the behavior and types of pedestrians



Illustrations: Laura Cohen

Smart robots that can interact with humans safely

Bystanders variety: pedestrians, bicycles, scooter, wheelchairs, vehicles



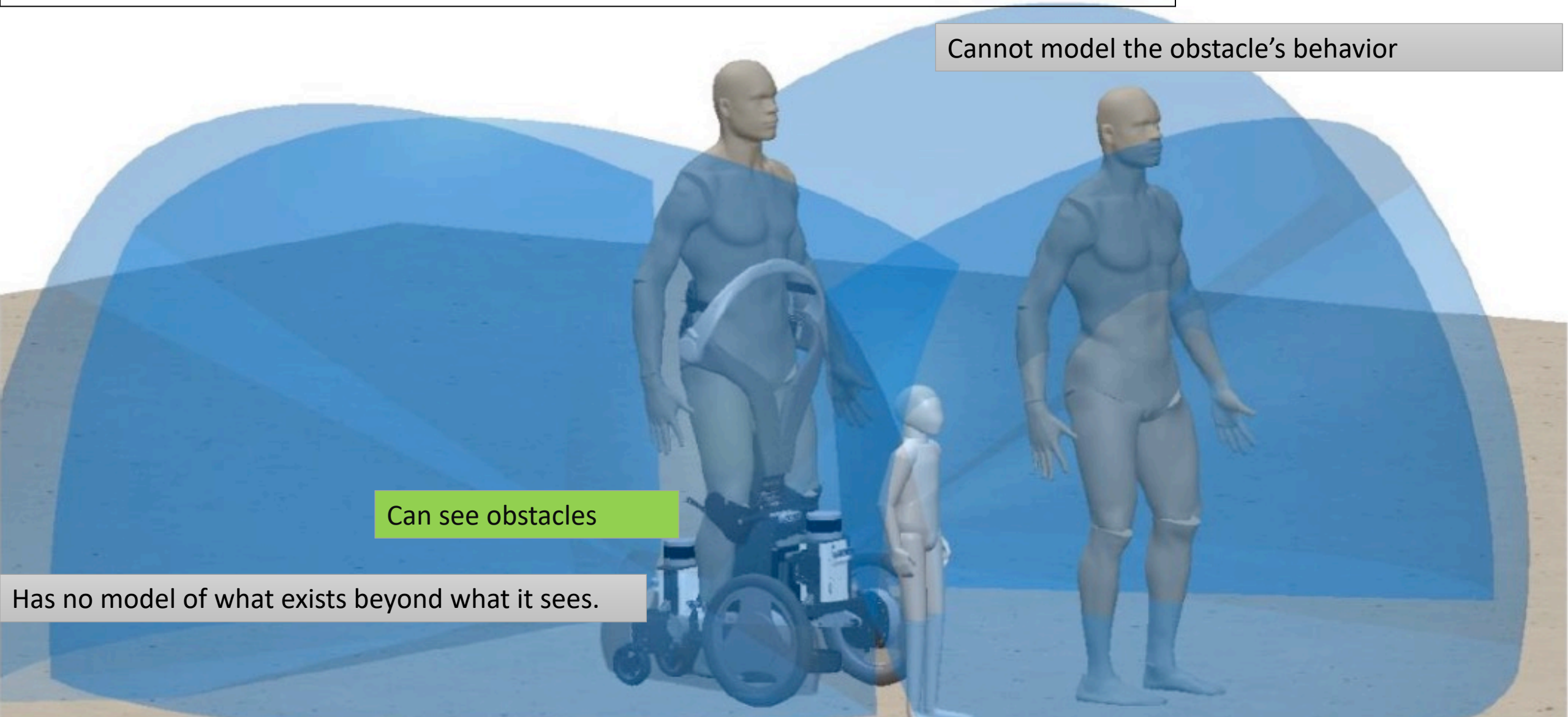
Smart robots that can interact with humans safely

What is the Level of robot's understanding?

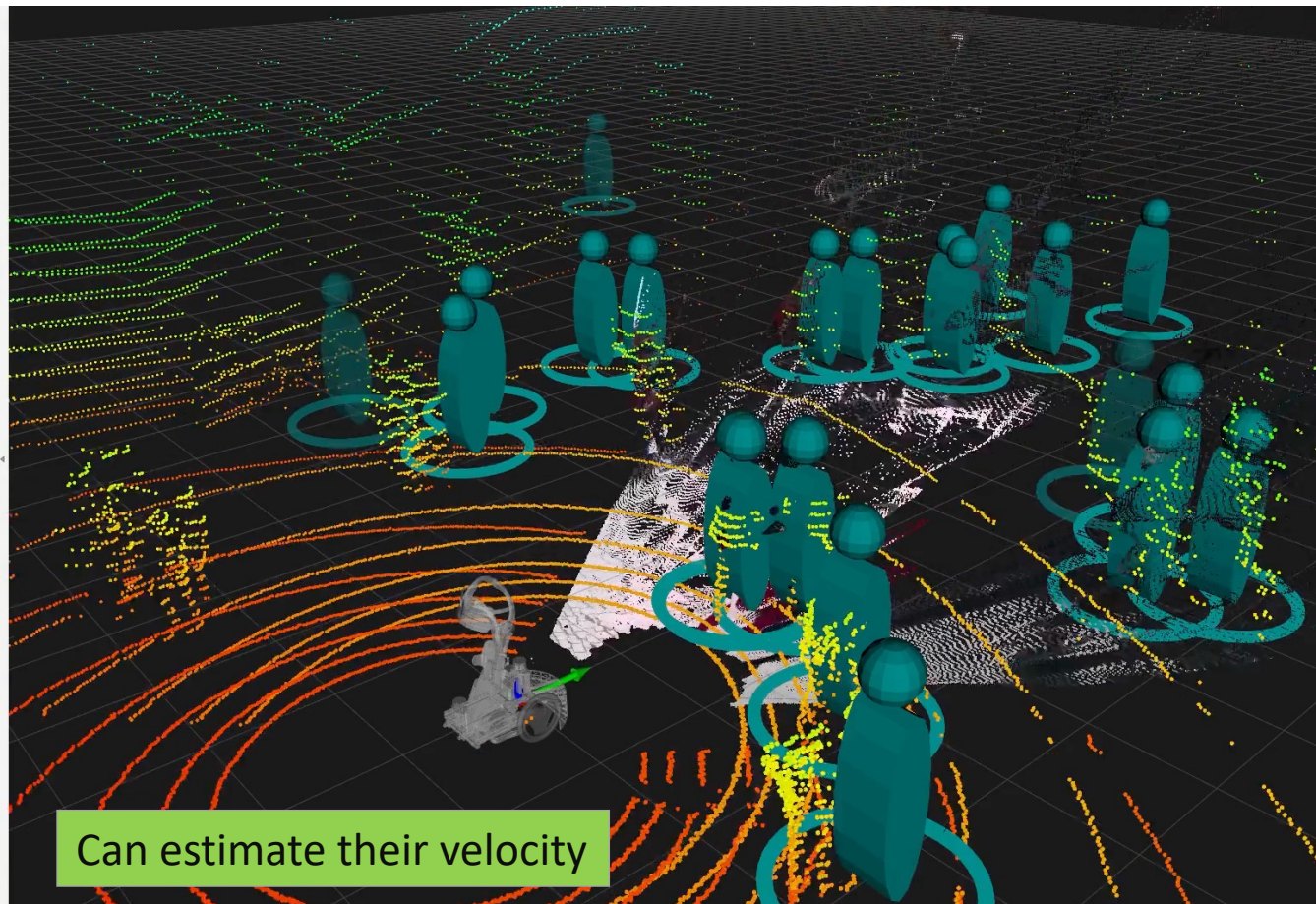
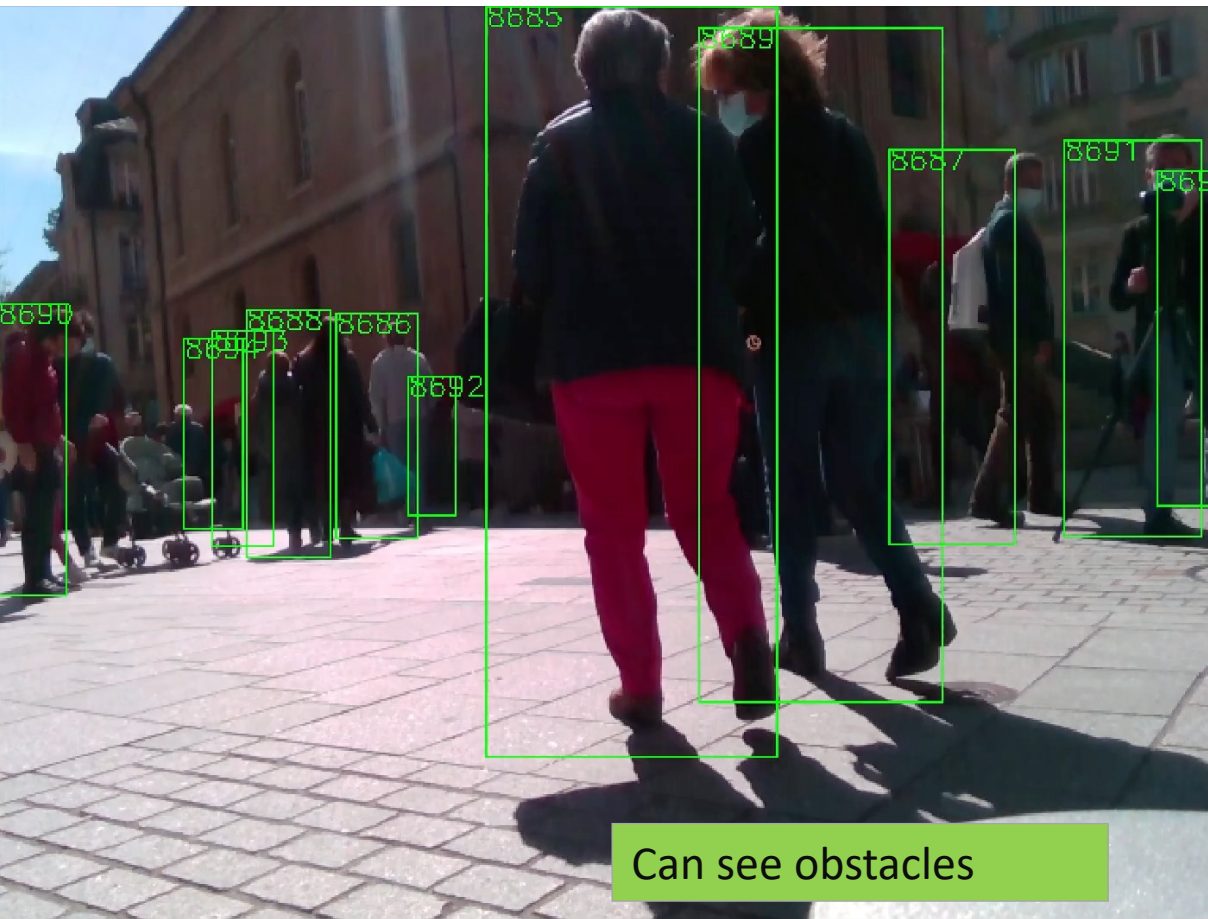
Cannot model the obstacle's behavior

Can see obstacles

Has no model of what exists beyond what it sees.



Smart robots that can interact with humans safely



Real time Robot's Egocentric view:

Smart robots that can interact with humans safely

EPFL

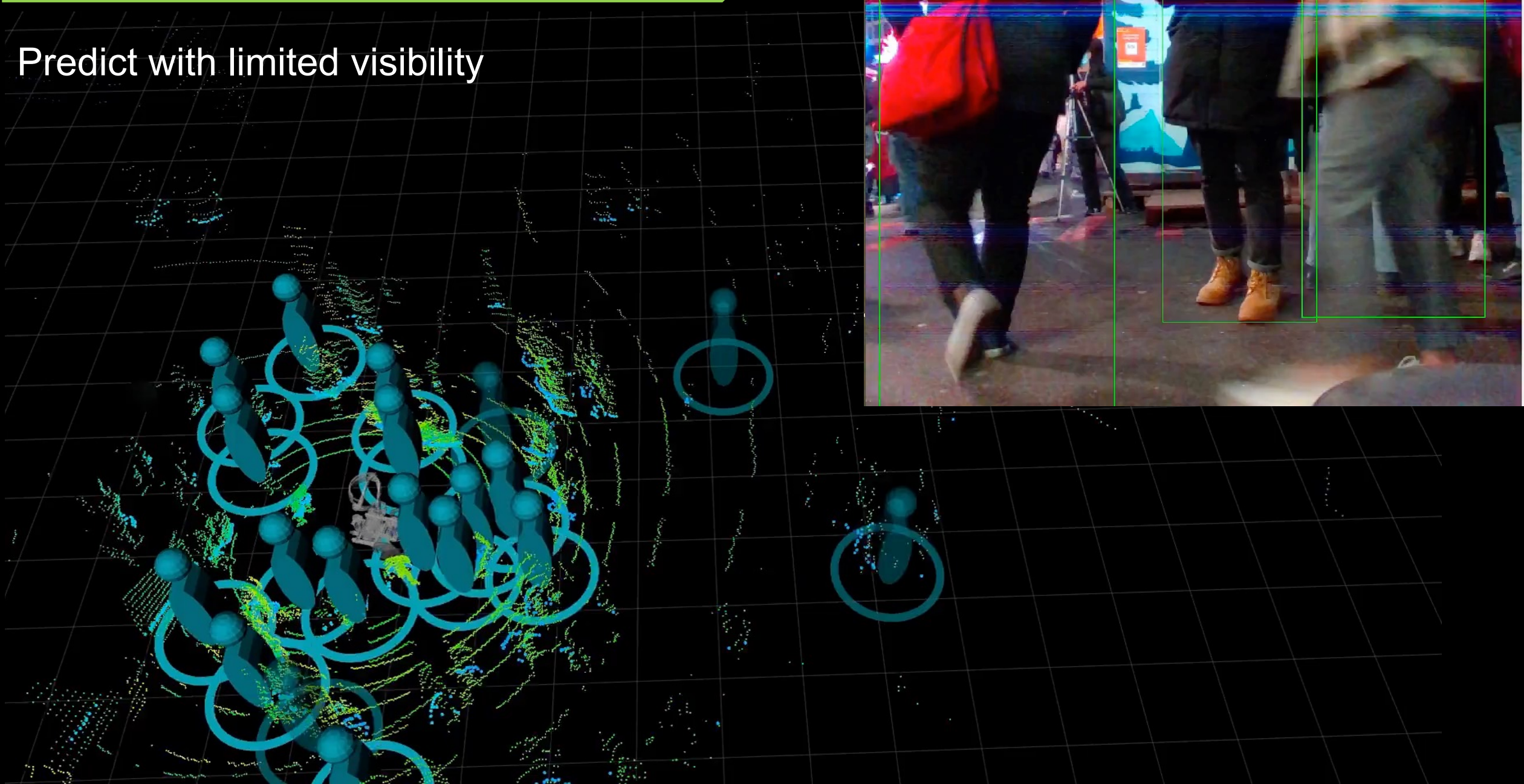
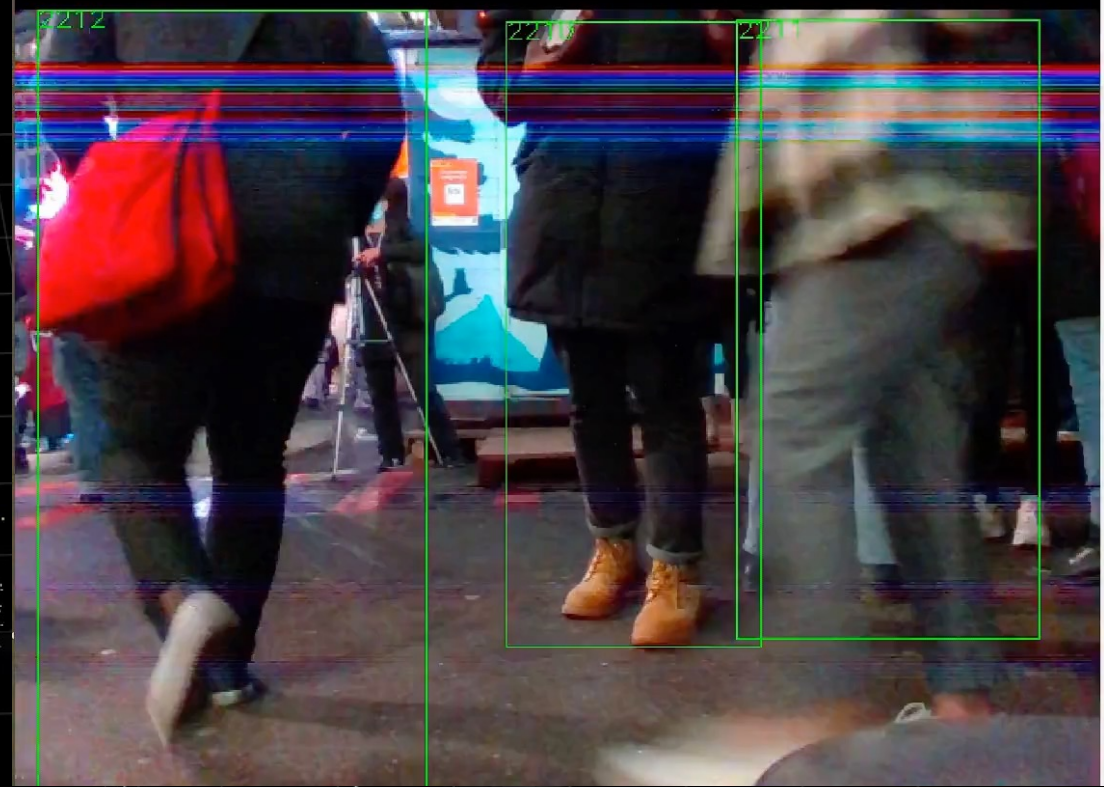
Understand the Crowd!

CROWDBOT



Smart robots that can interact with humans safely

Predict with limited visibility





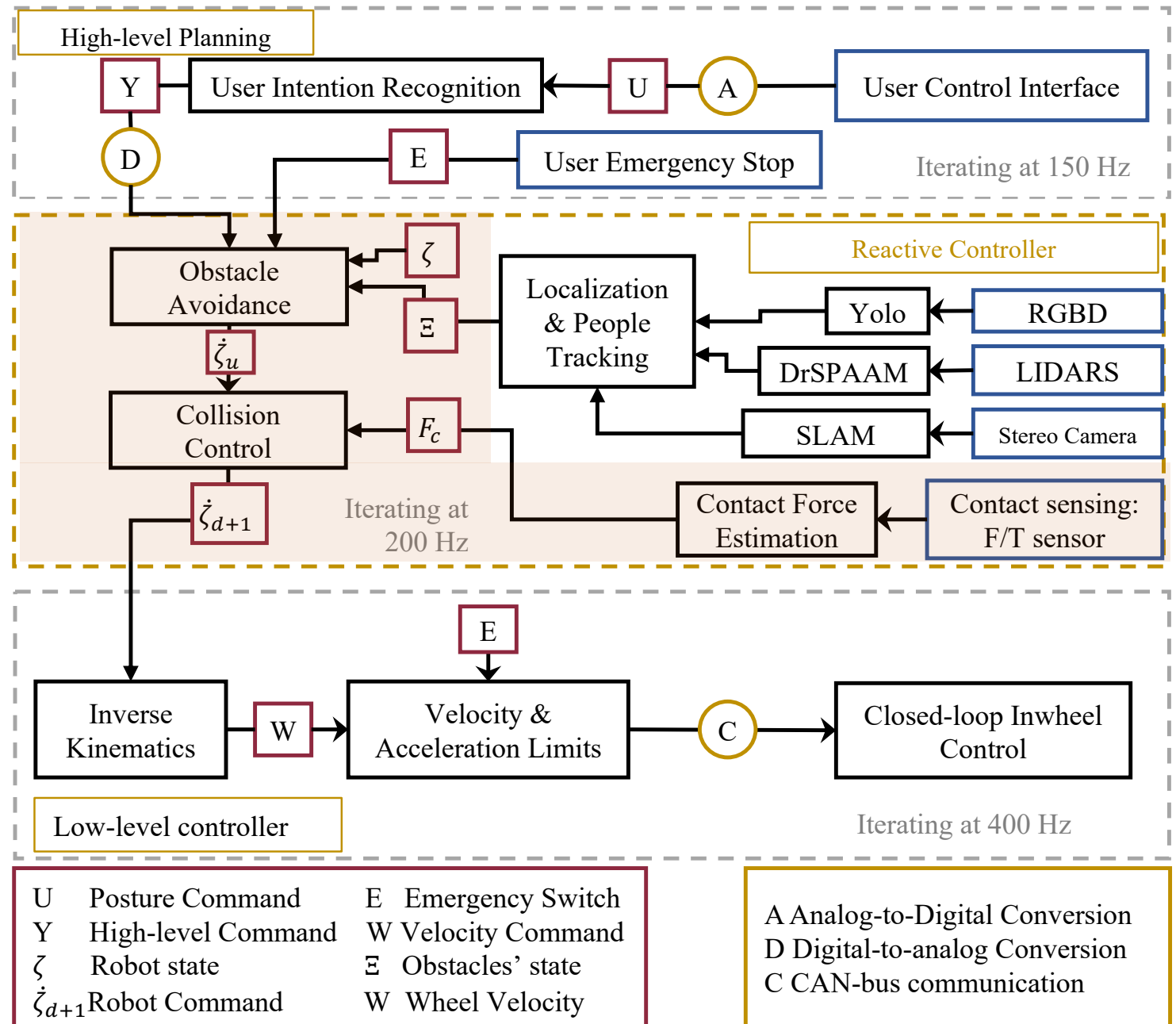
EPFL

Metrics in Social Navigation



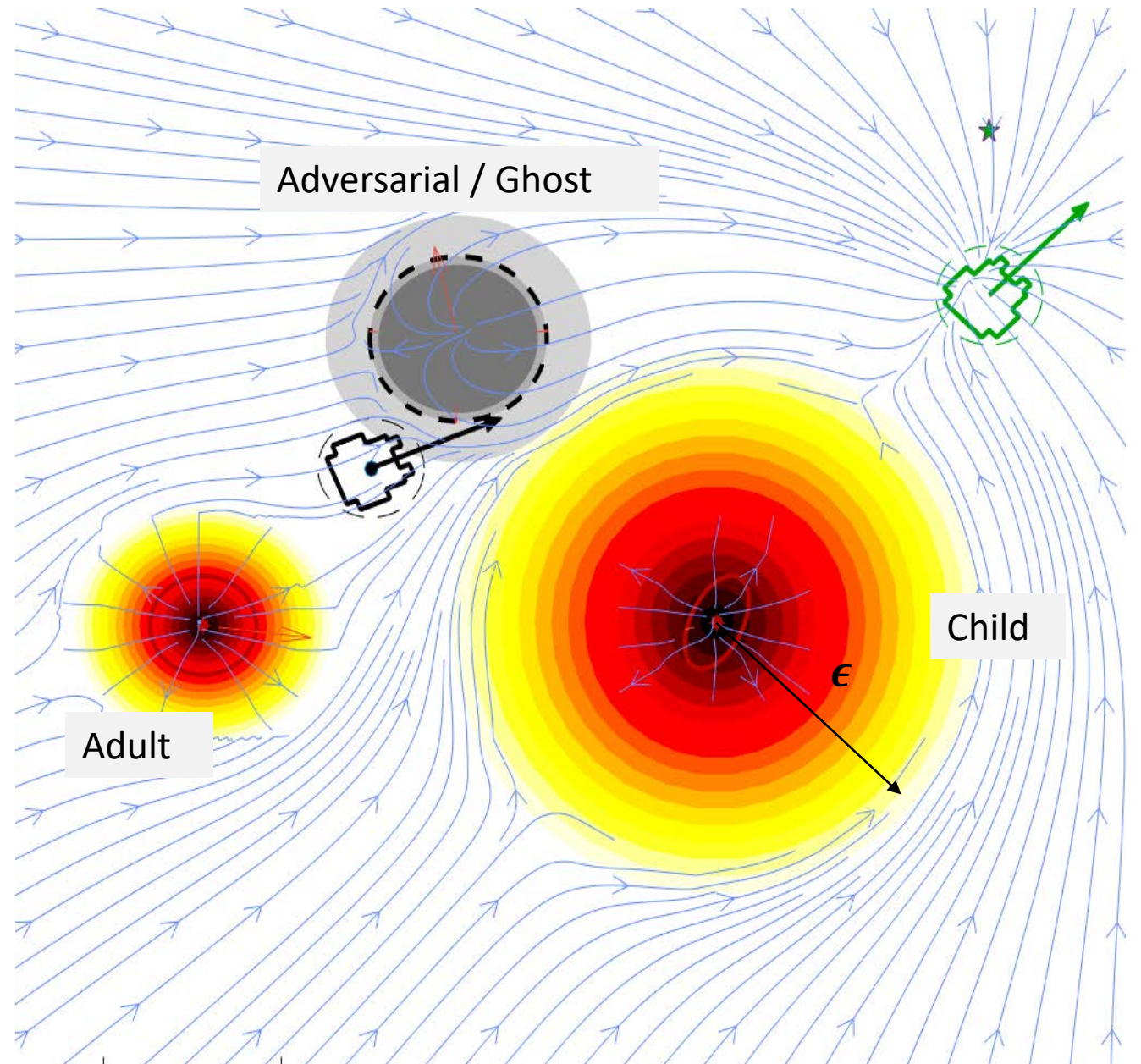
Unified Framework for pre- and post-Collision Control

Paez-Granados, D., Gupta, V., & Billard, A. (2022). Unfreezing Social Navigation : Dynamical Systems based Compliance for Contact Control in Robot Navigation. *IEEE International Conference on Robotics and Automation (ICRA)*, 1(1), 1–7



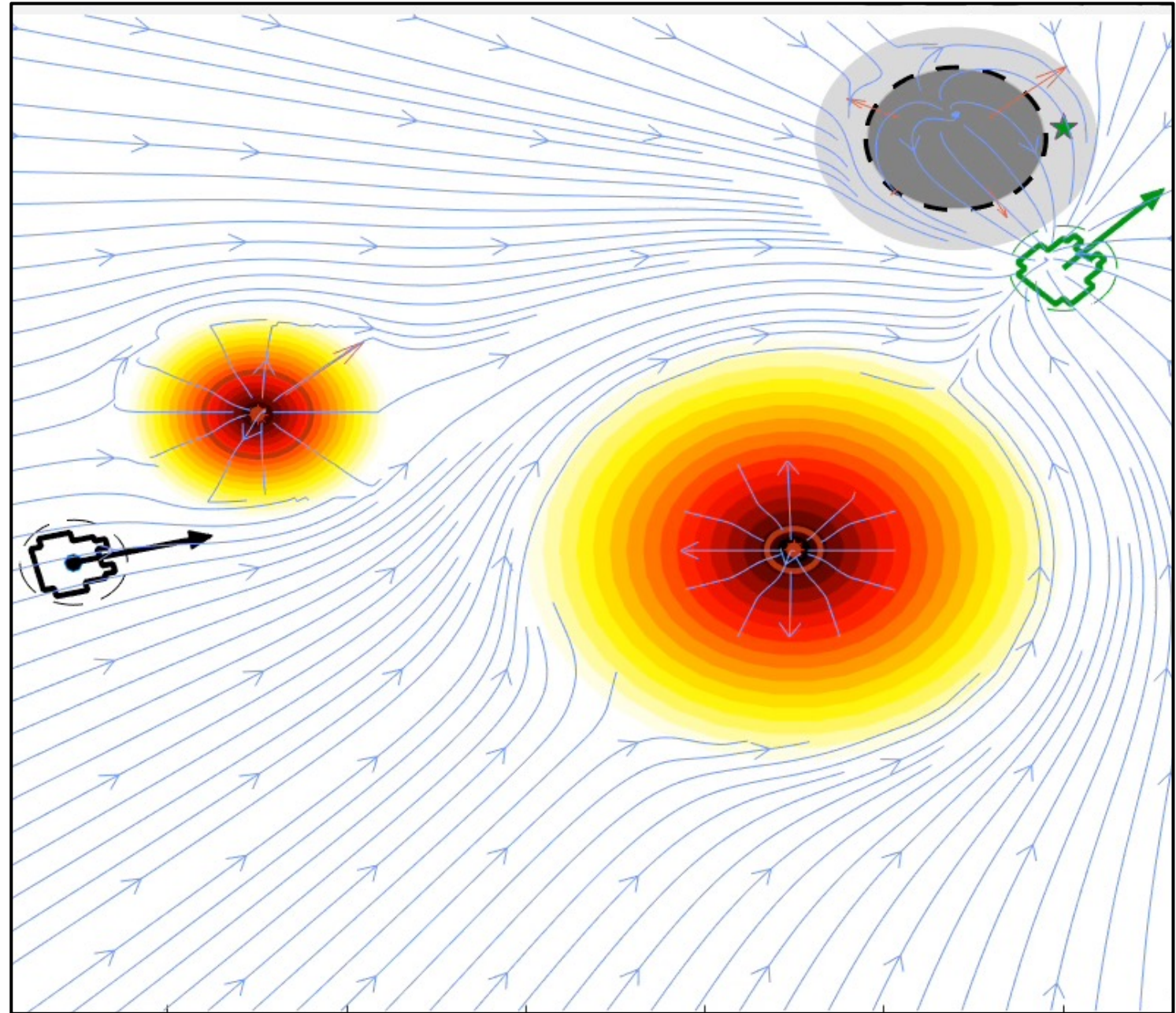
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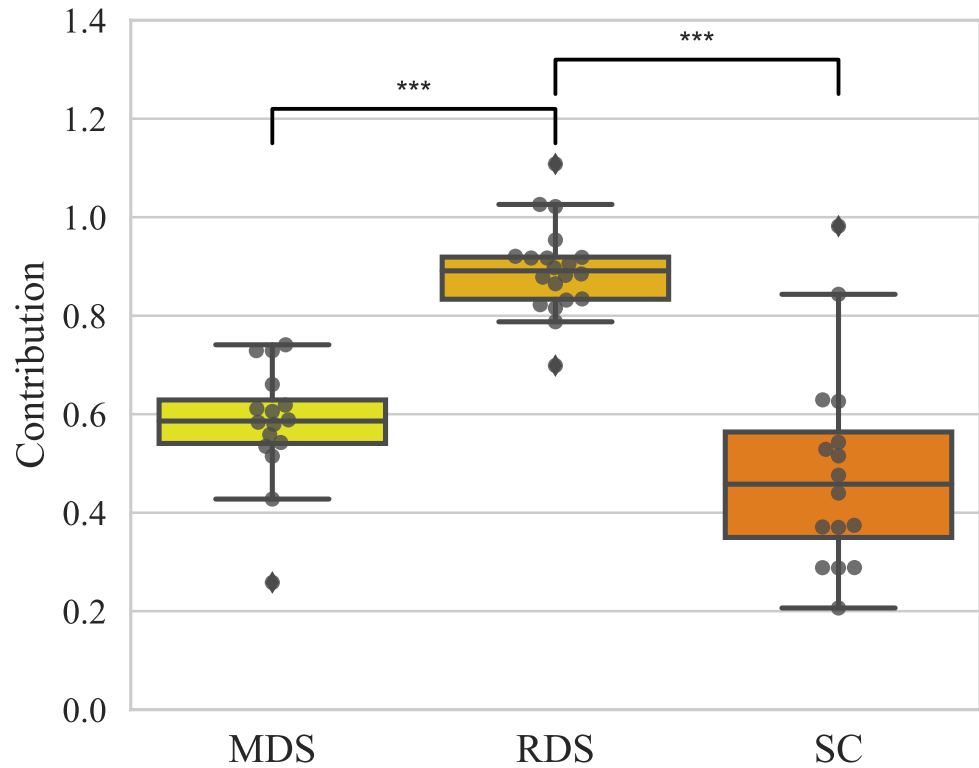
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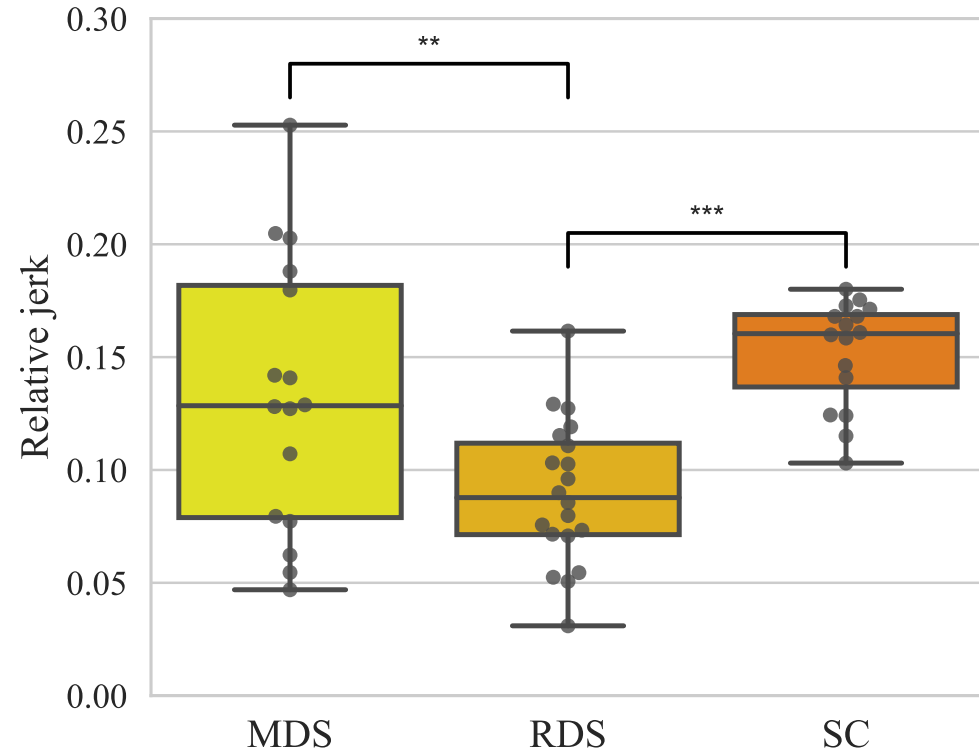


Autonomy versus shared control

CONTROLLER CONTRIBUTION



MOTION JERK



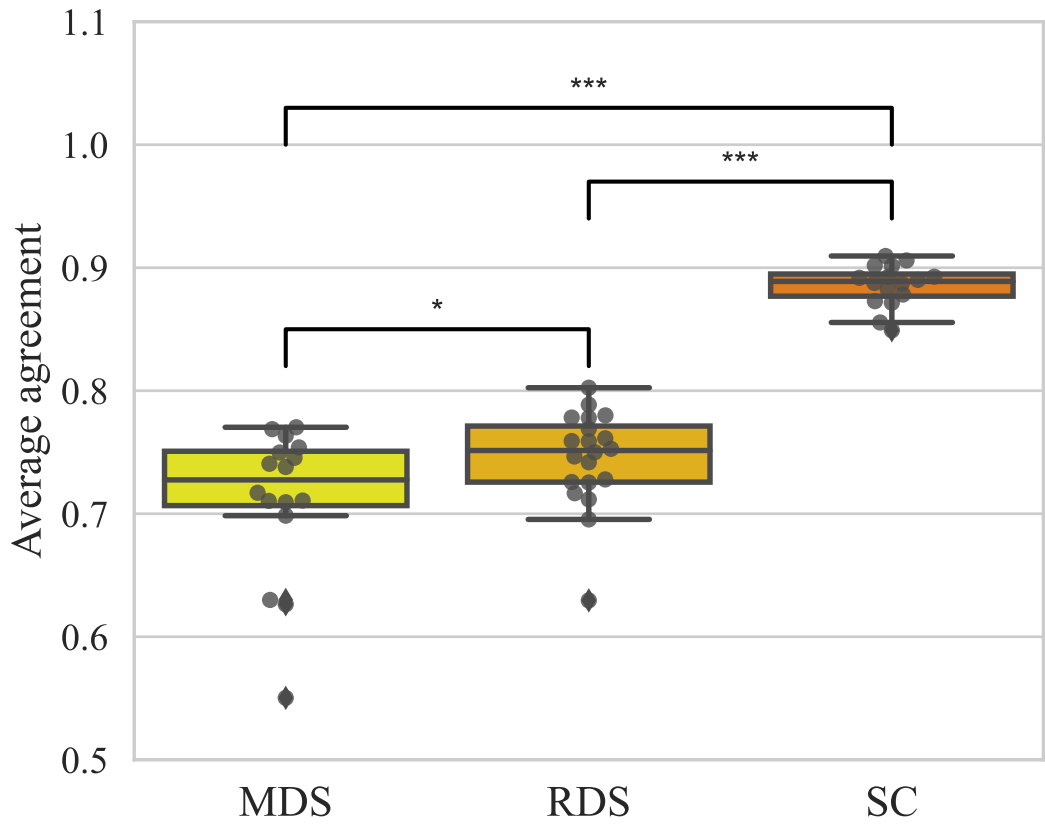
Number of tests compared:

- MDS: **15 / 18**
- RDS: **20 / 30**
- Shared control (SC): **17 / 45**

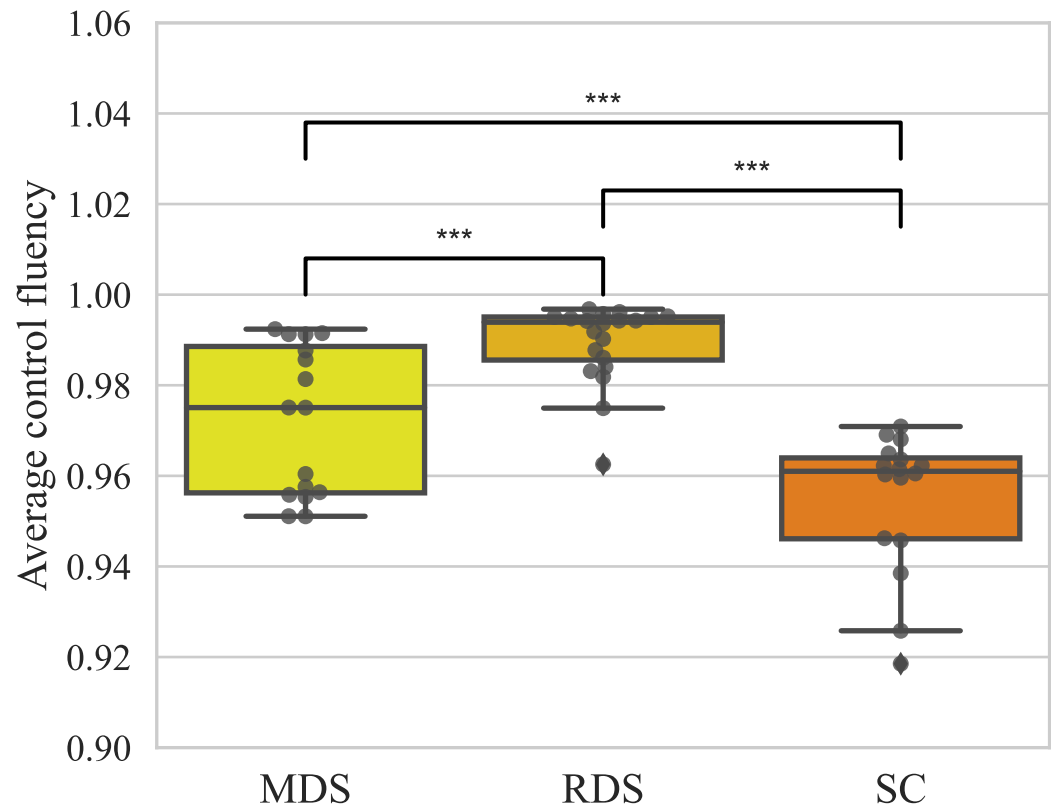
ANOVA results
* → $p < 0.1$
** → $p < 0.05$
*** → $p < 0.05$

Autonomy versus shared control

AGREEMENT



HIGH-LEVEL FLUENCY



Number of collisions:

- MDS: 2
- RDS: 2
- Shared Control (SC): 3

ANOVA results

- * → $p < 0.1$
- ** → $p < 0.05$
- *** → $p < 0.001$

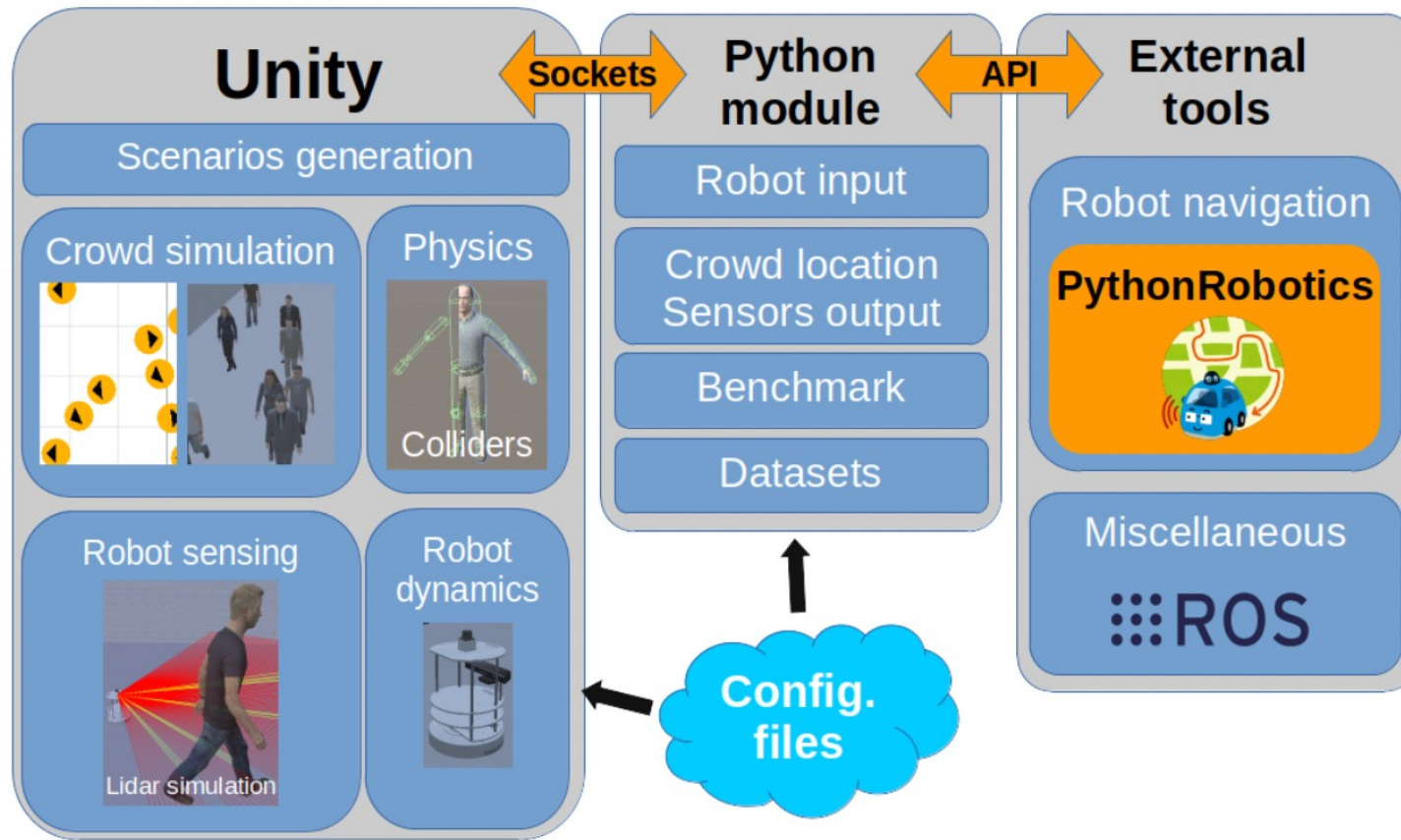
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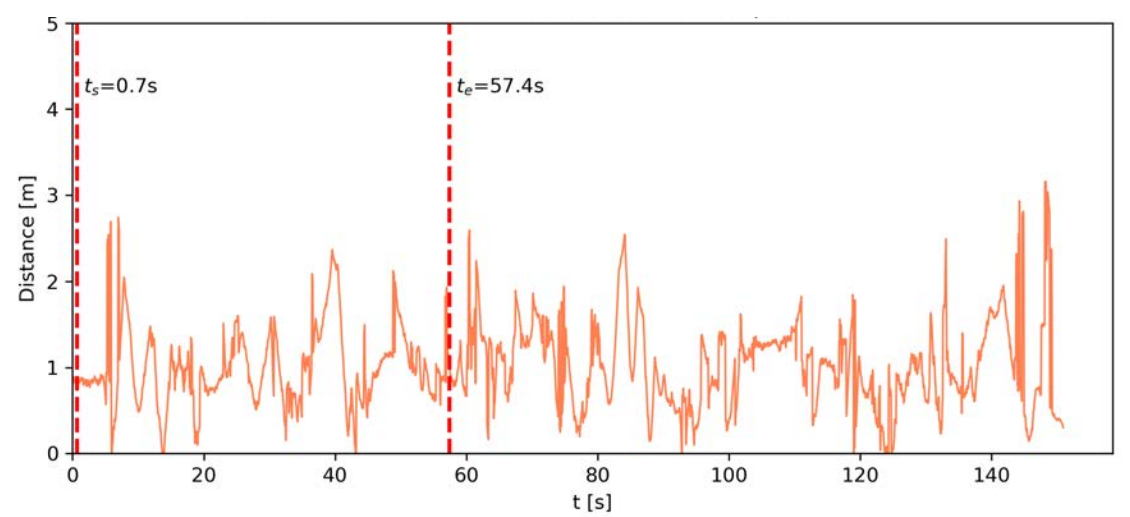
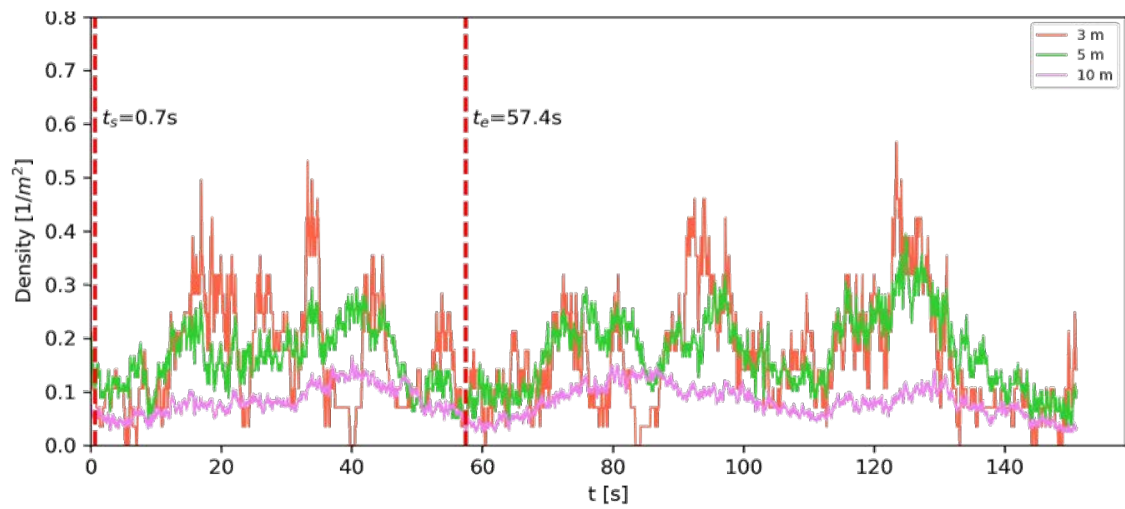
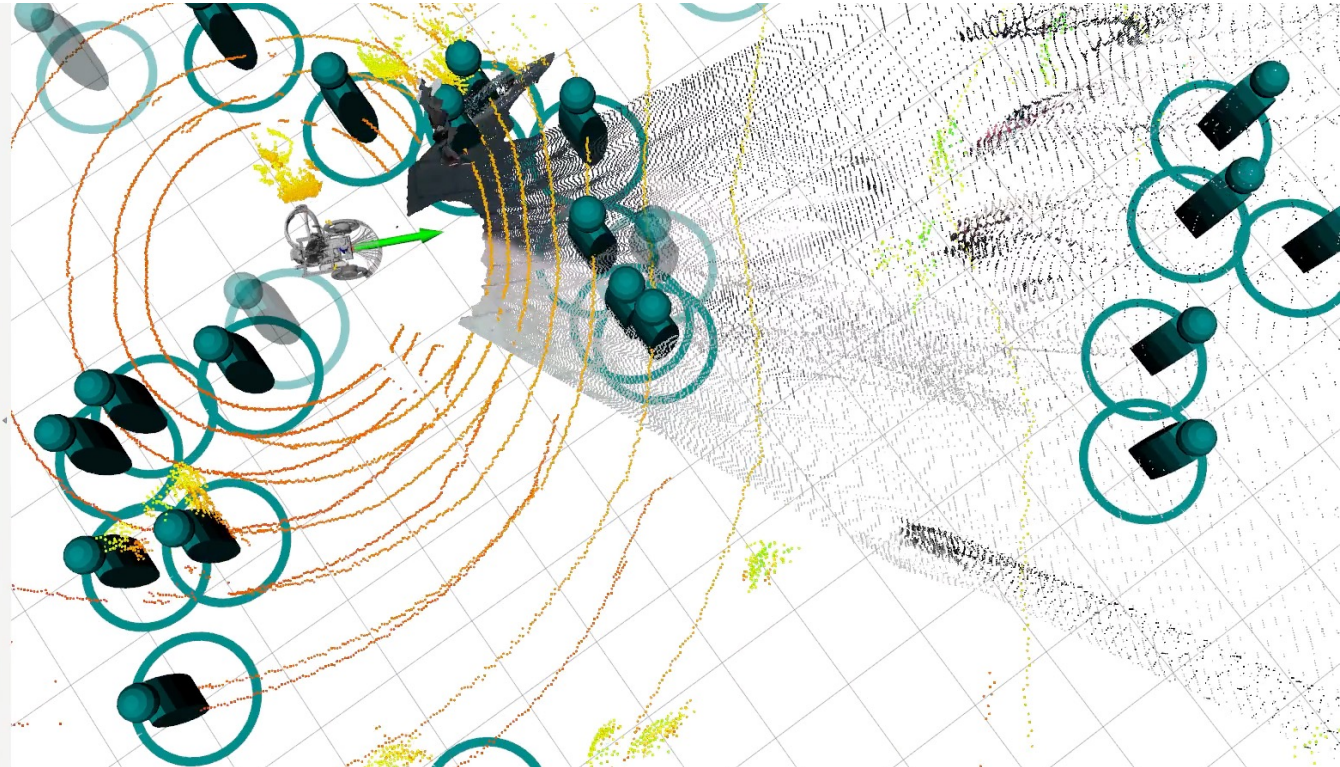
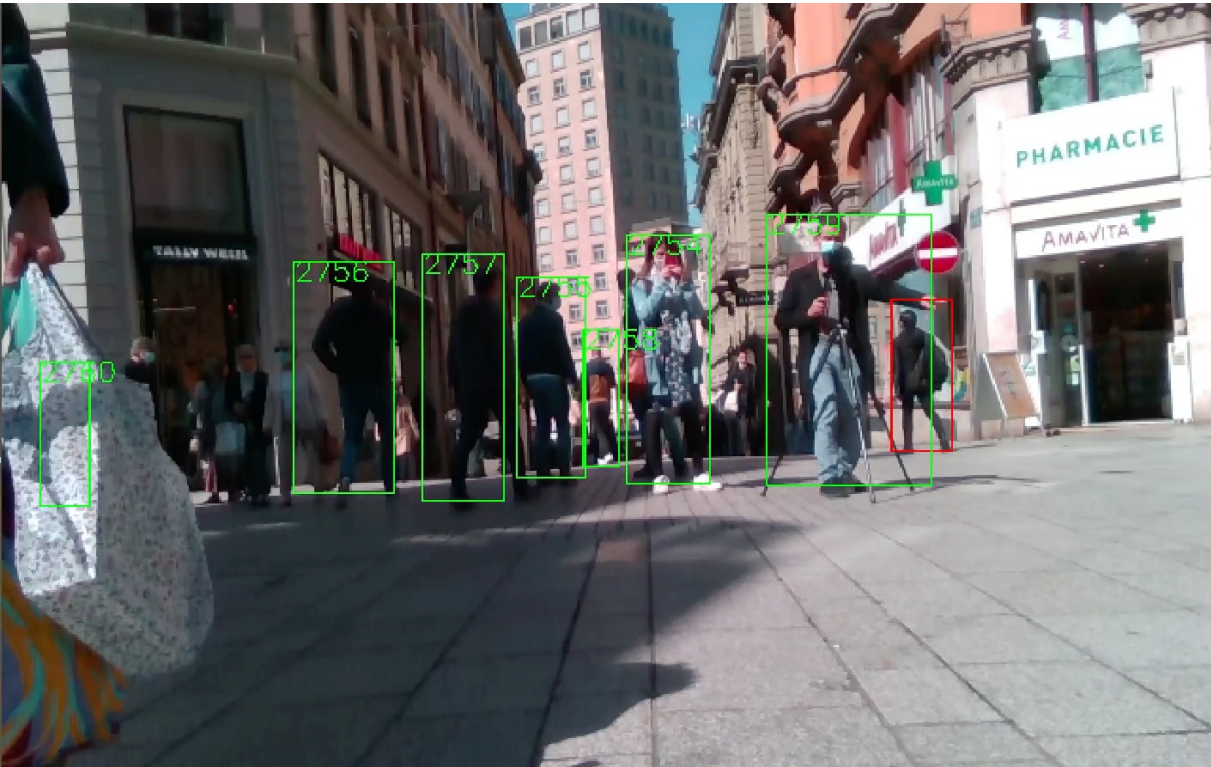
CONTROLLER COMPARISON IN MID DENSITY CROWDS

Metrics	Controller		
	MDS	RDS	Shared control
Avg. crowd density	0.12 ± 0.03	0.13 ± 0.03	0.12 ± 0.03
Max crowd density	0.45 ± 0.08	0.47 ± 0.12	0.51 ± 0.14
Min distance	1.19 ± 0.16	1.08 ± 0.18	1.20 ± 0.16
Time to goal	0.28 ± 0.09	0.32 ± 0.10	0.29 ± 0.07
Path length	1.41 ± 0.21	1.34 ± 0.20	1.52 ± 0.52
Jerk	0.13 ± 0.06	0.09 ± 0.03	0.15 ± 0.02
Contribution	0.58 ± 0.12	0.89 ± 0.09	0.49 ± 0.21
Avg. fluency	0.97 ± 0.02	0.99 ± 0.01	0.95 ± 0.02
Avg. agreement	0.71 ± 0.06	0.74 ± 0.04	0.89 ± 0.02
Virtual collision	3.50 ± 2.71	7.05 ± 7.92	4.25 ± 3.11
Actual collision	2/16	2/20	3/16

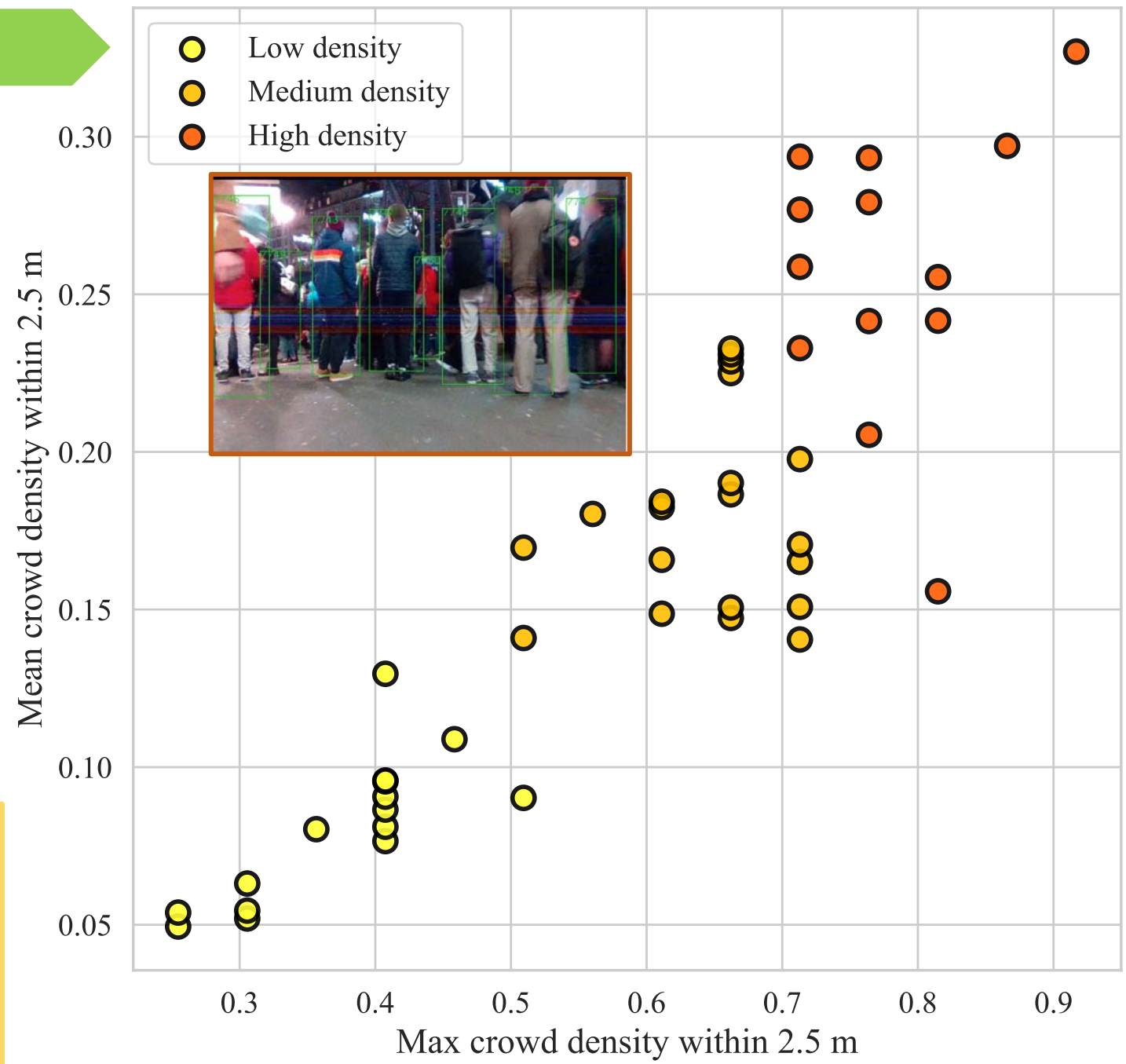
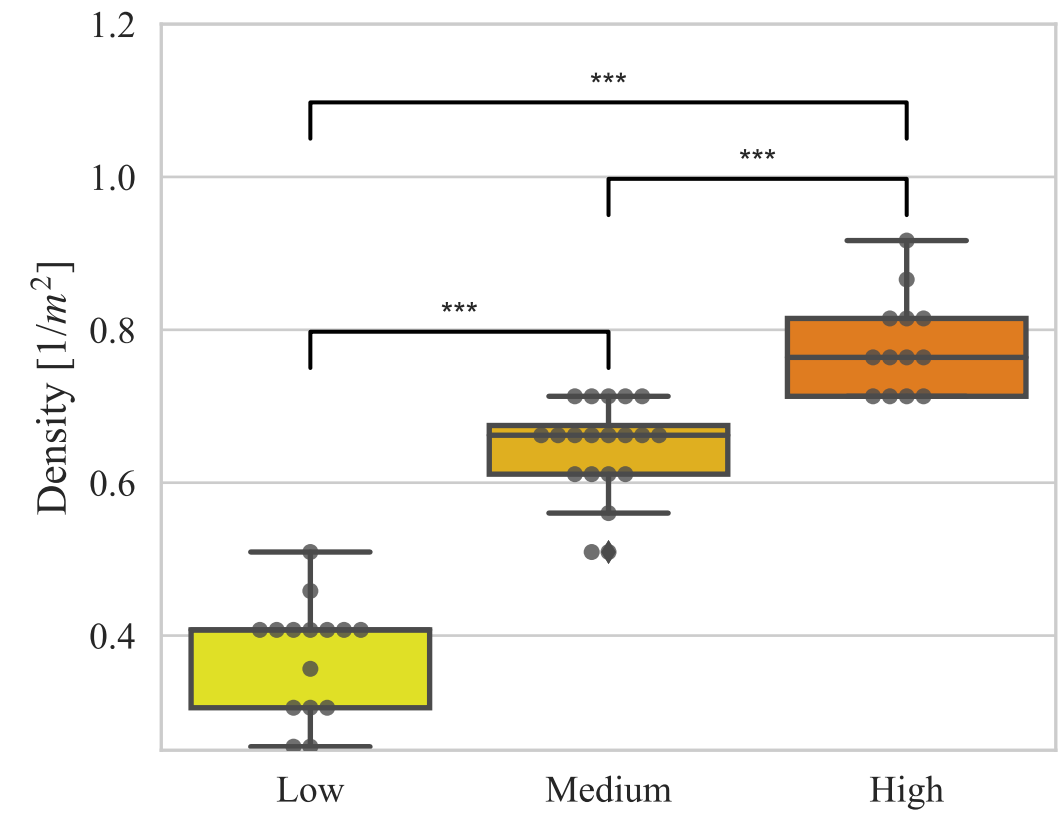
Measurable from egocentric data



Grzeskowiak, F., Gonon, D., Dugas, D., Paez-Granados, D., Chung, J., Nieto, J., Siegwart, R., Billard, A., Babel, M., & Pettré, J. (2021). Crowd against the machine: A simulation-based benchmark tool to evaluate and compare robot capabilities to navigate a human crowd. *IEEE International Conference on Robotics and Automation (ICRA-2021)*. <https://doi.org/10.1109/ICRA48506.2021.9561694>

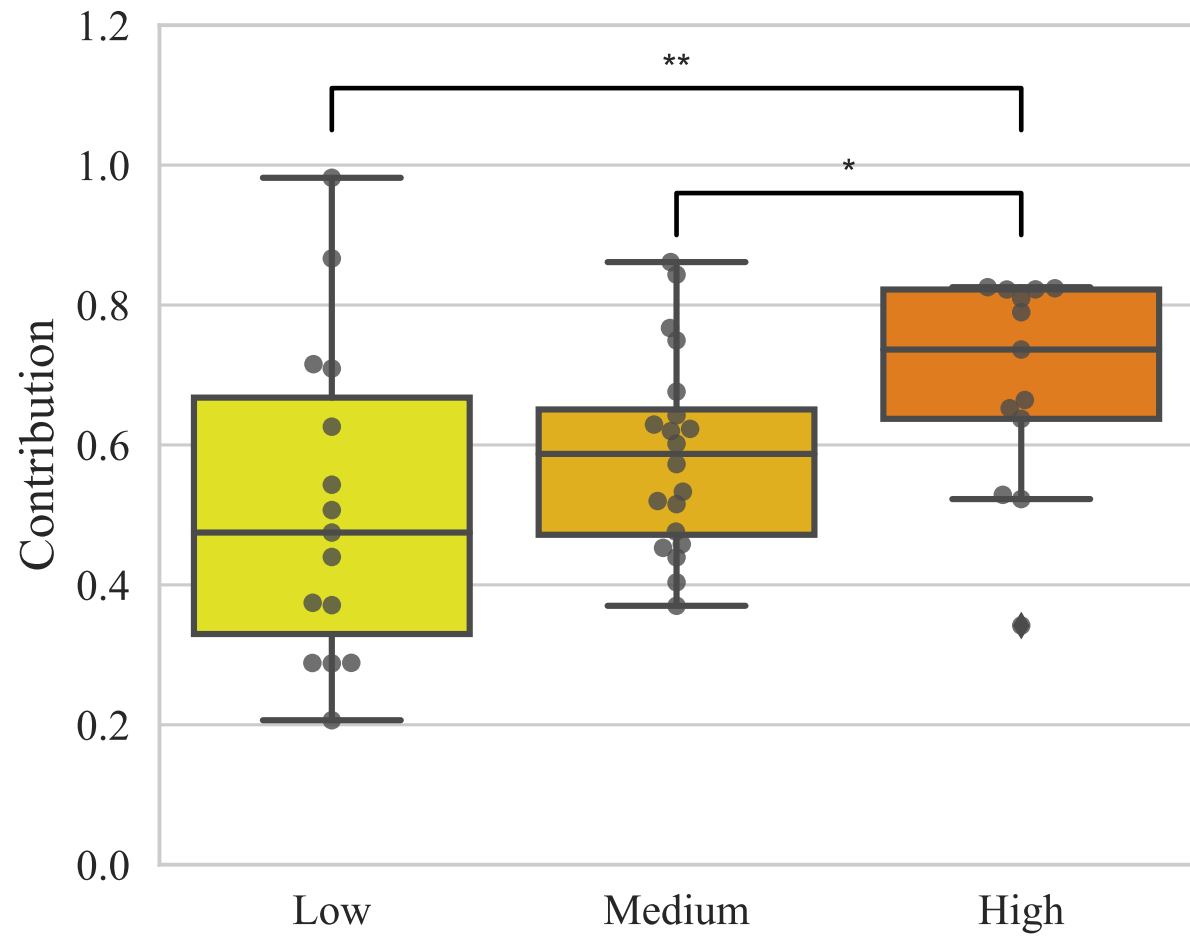


Crowd Type and Density

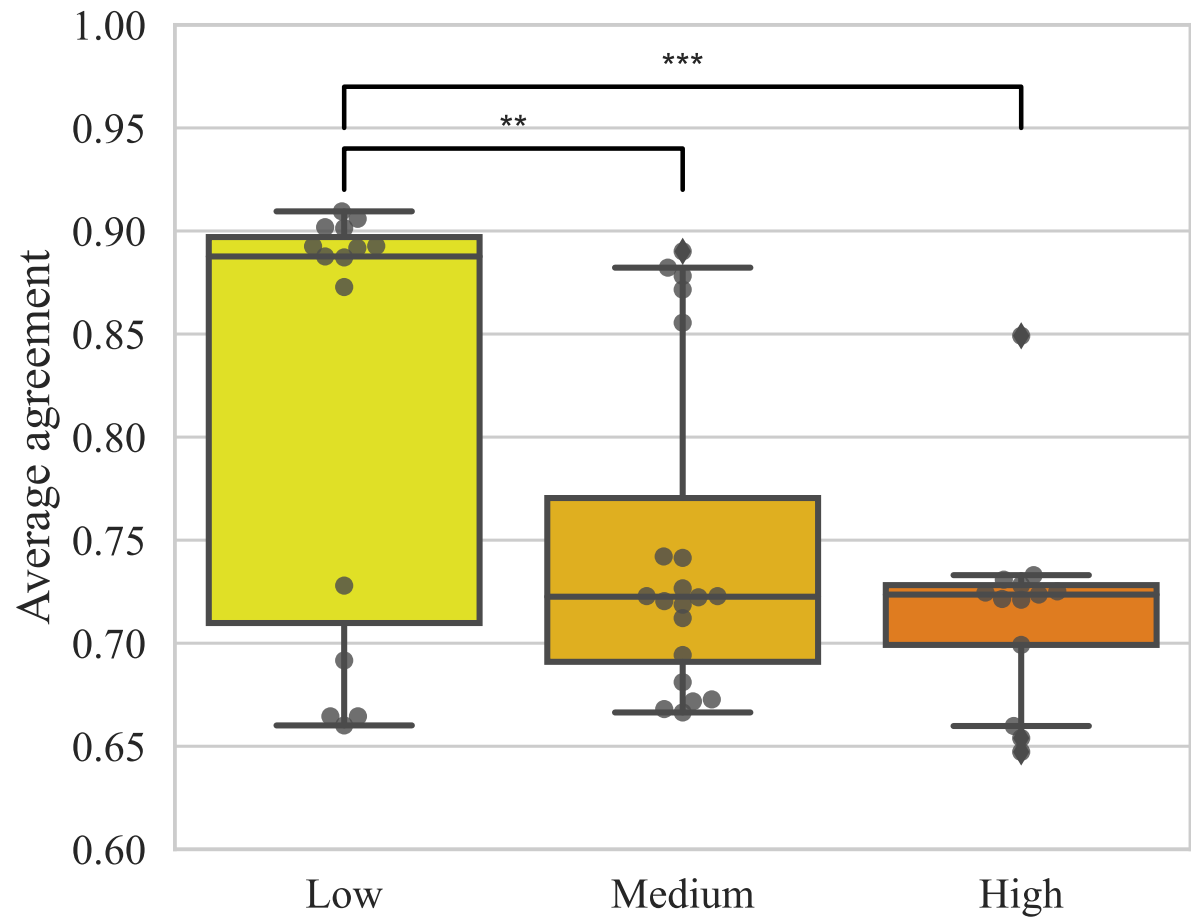


Crowd Type and Density

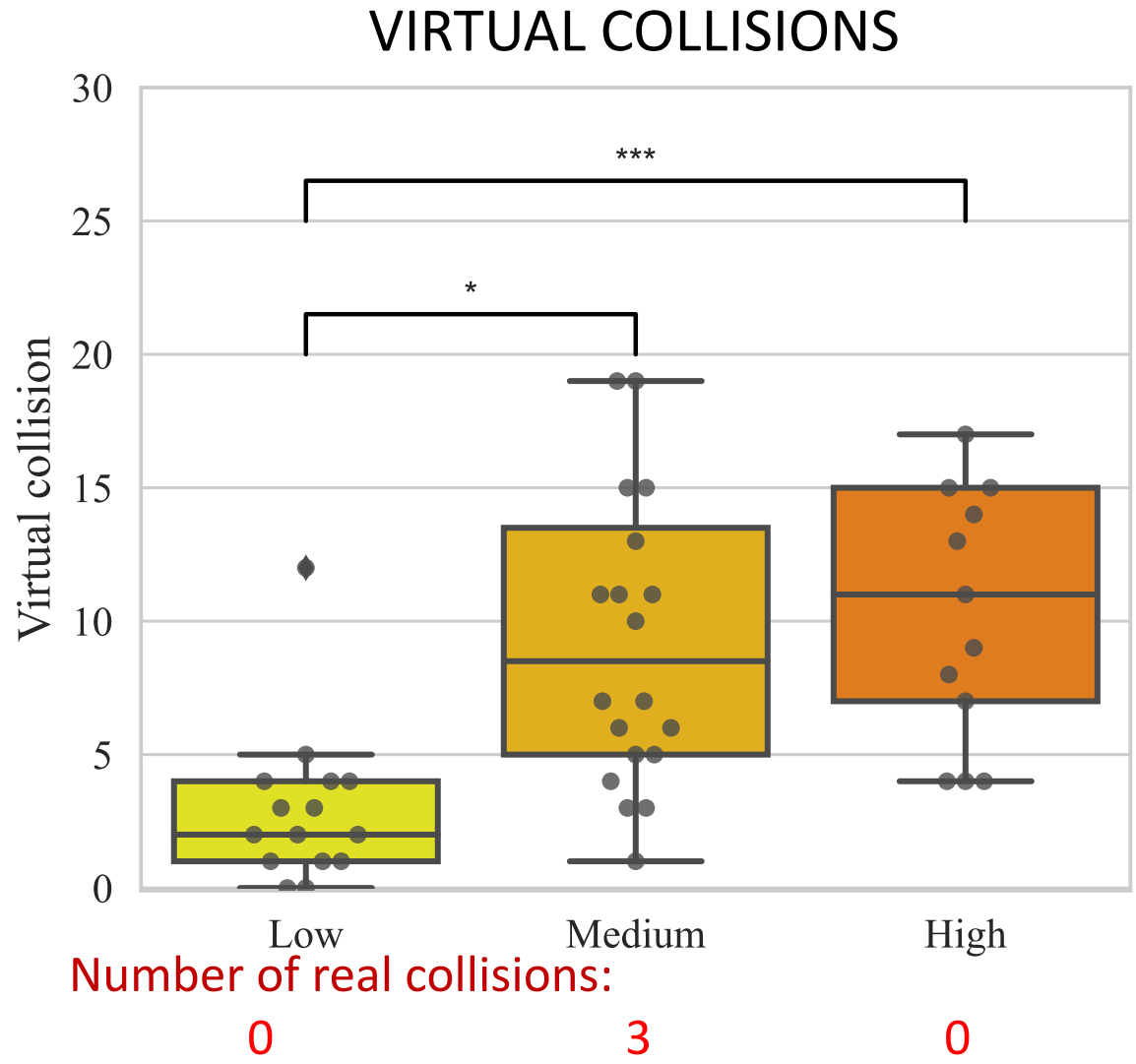
CONTROLLER CONTRIBUTION



USER AGREEMENT



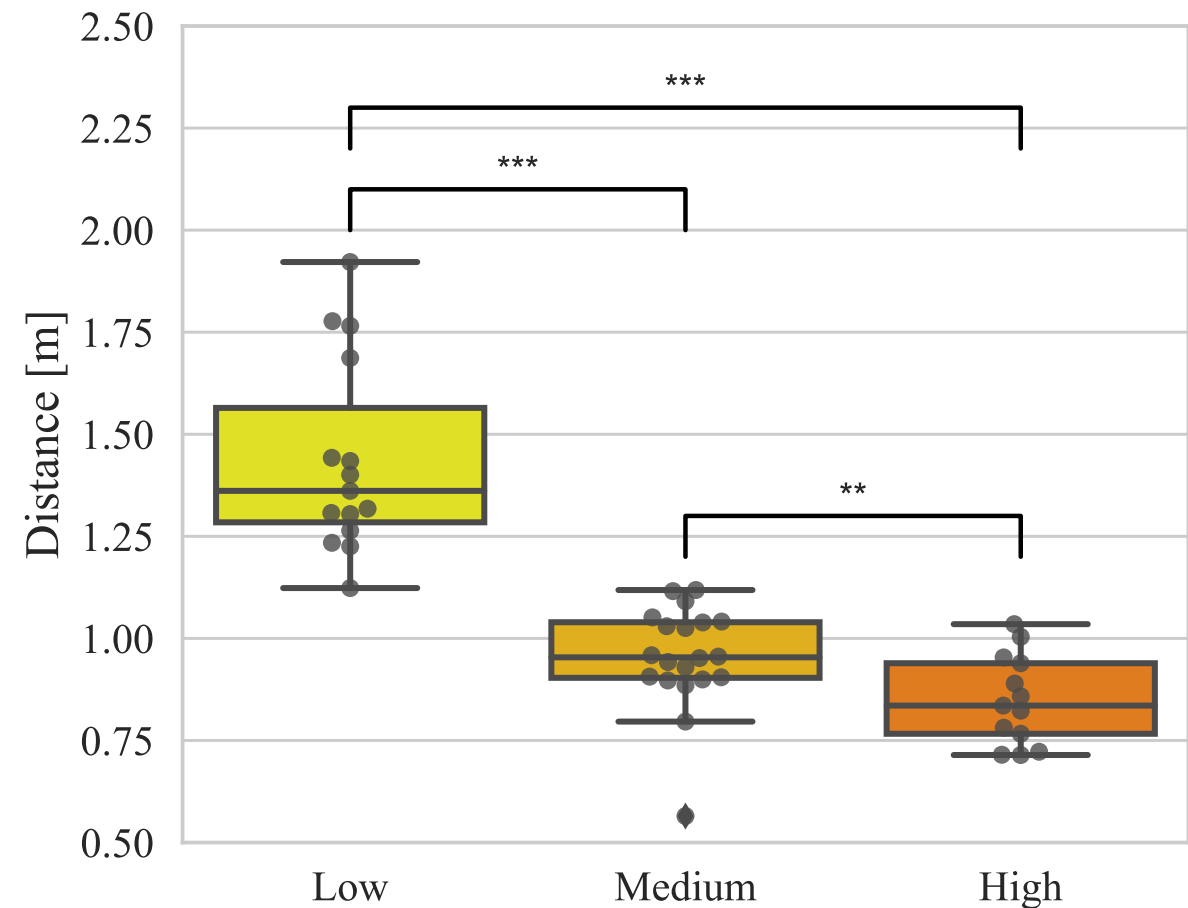
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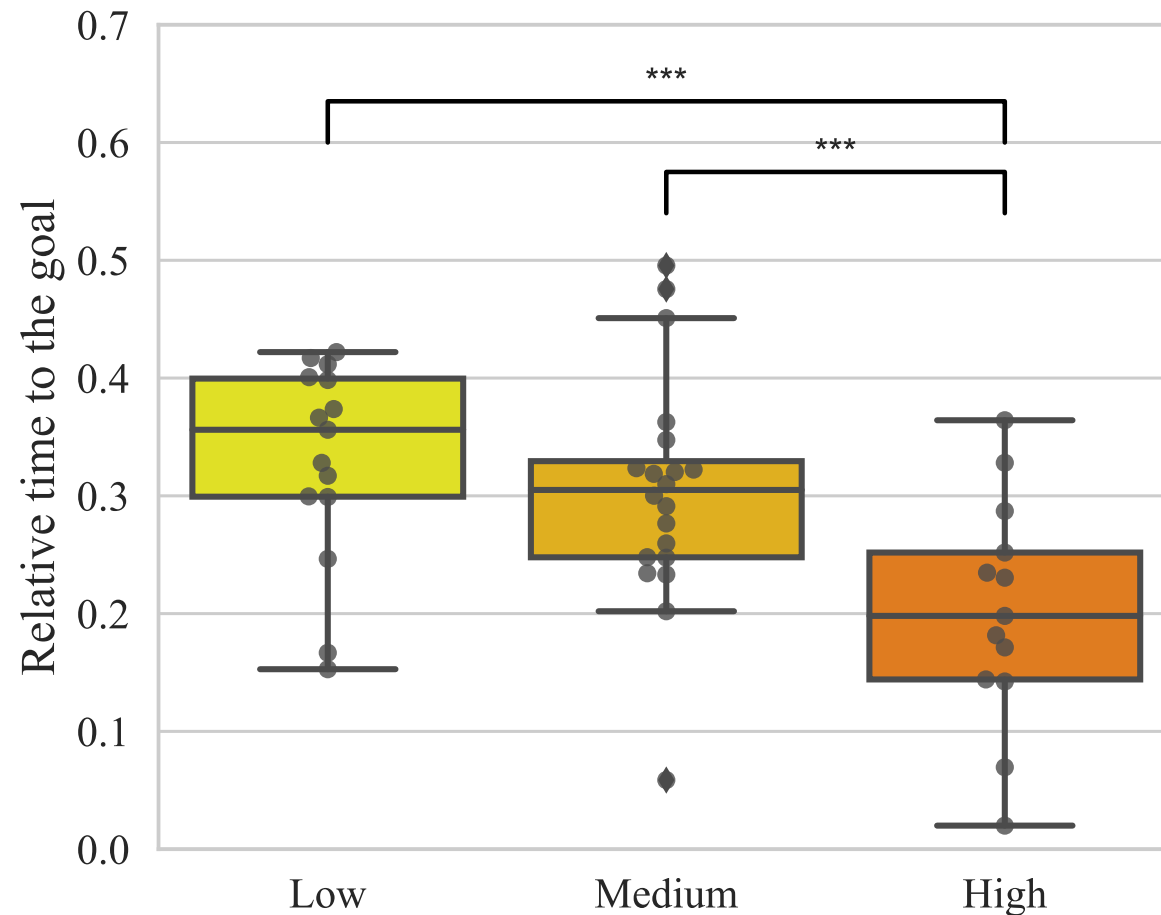
ANOVA results
* → p < 0.1
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*** → p < 0.01

Crowd Type and Density

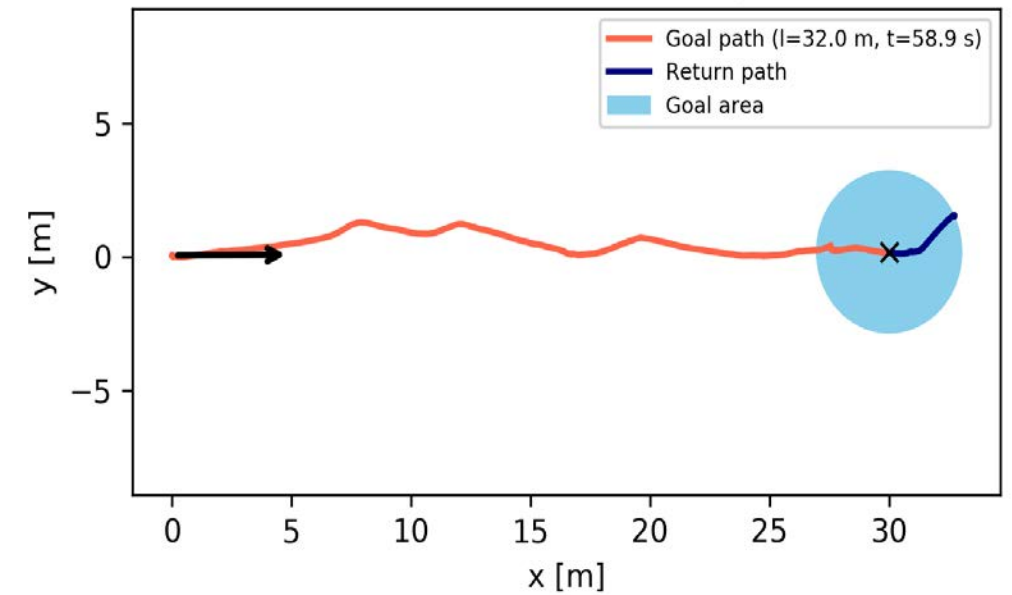
Min Distance to Pedestrians



TIME TO THE GOAL



Autonomy versus shared control



Paez-Granados D., He Y., Gonon D., Huber L., & Billard A., (2021), "3D point cloud and RGBD of pedestrians in robot crowd navigation: detection and tracking." IEEE Dataport, doi: <https://dx.doi.org/10.21227/ak77-d722>.

Recorded Data	
Recordings	250k frames of 2 x 3D point clouds
Tests	120 recordings
Time	~ 5 hours of data
Traveled distance	~ 12 km
Total raw data	~1.2 Tb

Smart robots that can interact with humans safely

How to measure autonomy?

Brain-Machine Interface

Not ready yet!

Robots understanding of their surrounding too limited

Social Navigation on crowds needs to understand their behavior

We need to evaluate robot's behavior for each applications and society

We need to agree on egocentric feasible metrics for real applications

ISO- 18646 - Robotics — Performance criteria and related test methods for service robots —
ISO/DIS 22737 - Intelligent transport systems — Low-speed automated driving (LSAD) systems for predefined routes



Metrics and Safety for Autonomous Robot Navigation

Dr. Diego Paez-Granados



ACKNOWLEDGEMENT

Funding

EU H2020 project "Crowdbot" (779942)

Toyota Mobility Unlimited Challenge (2019-21)

The experiments were approved with an ethical protocol by the human research ethical committee of EPFL

Approval No: HREC-032-2019 / 2021

LASA
Learning Algorithms and
Systems Laboratory

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Prof. Bastian Leibe, RWTH