

# Deep-PANTHER: Learning-based perception-aware trajectory planner in dynamic environments

J. Tordesillas Torres; J.P. How

## Abstract-

This letter presents Deep-PANTHER, a learning-based perception-aware trajectory planner for unmanned aerial vehicles (UAVs) in dynamic environments. Given the current state of the UAV, and the predicted trajectory and size of the obstacle, Deep-PANTHER generates multiple trajectories to avoid a dynamic obstacle while simultaneously maximizing its presence in the field of view (FOV) of the onboard camera. To obtain a computationally tractable real-time solution, imitation learning is leveraged to train a Deep-PANTHER policy using demonstrations provided by a multimodal optimization-based expert. Extensive simulations show replanning times that are two orders of magnitude faster than the optimization-based expert, while achieving a similar cost. By ensuring that each expert trajectory is assigned to one distinct student trajectory in the loss function, Deep-PANTHER can also capture the multimodality of the problem and achieve a mean squared error (MSE) loss with respect to the expert that is up to 18 times smaller than state-of-the-art (Relaxed) Winner-Takes-All approaches. Deep-PANTHER is also shown to generalize well to obstacle trajectories that differ from the ones used in training.

**Index Terms-** UAV, Imitation learning, perception-aware trajectory planning, optimization.

Due to copyright restriction we cannot distribute this content on the web. However, clicking on the next link, authors will be able to distribute to you the full version of the paper:

[Request full paper to the authors](#)

If your institution has an electronic subscription to IEEE Robotics and Automation Letters, you can download the paper from the journal website:

[Access to the Journal website](#)

## Citation:

*How, J.P.; Tordesillas Torres, J. "Deep-PANTHER: Learning-based perception-aware trajectory planner in dynamic environments", IEEE Robotics and Automation Letters, vol.8, no.3, pp.1399-1406, March, 2023.*