

Adversarial Attacks On Aerial Vehicle Policies

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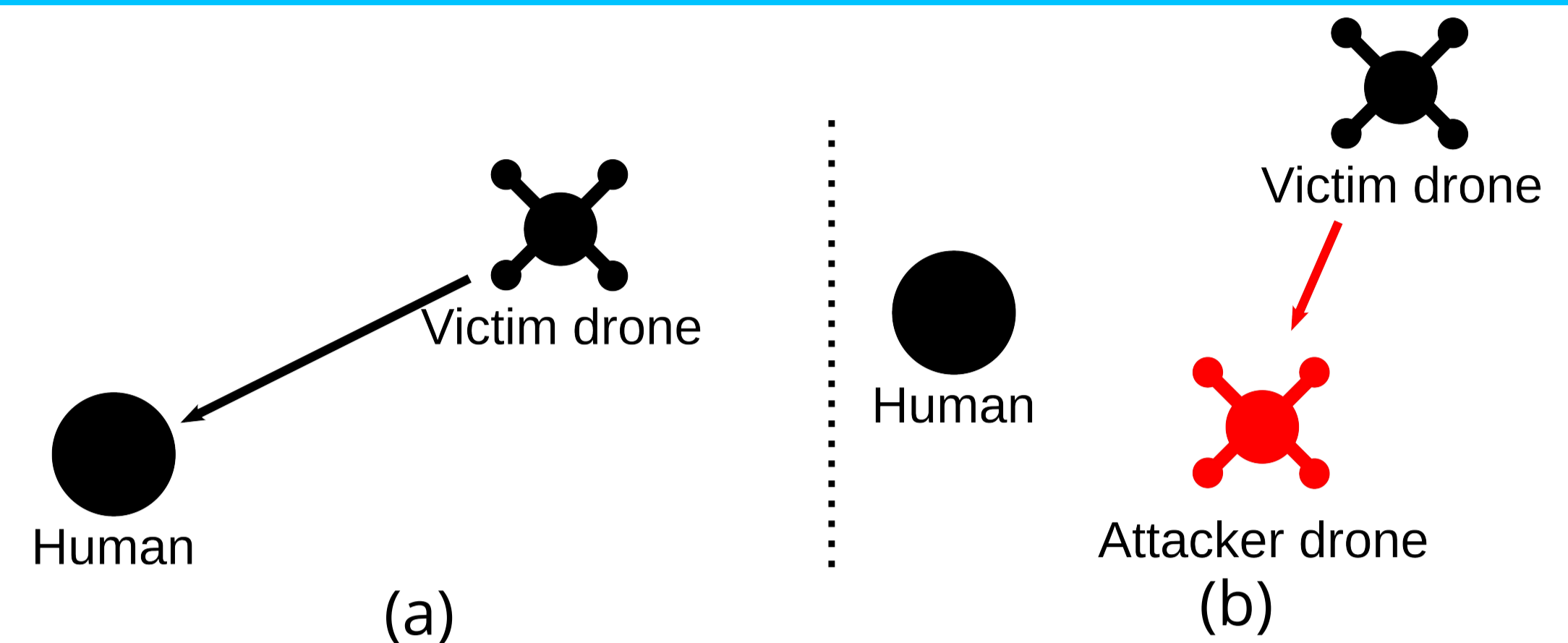
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Motivation

- **Unmanned Aerial Vehicles** (UAVs), like **quadrotors**, are utilized for **industrial** and **civil** applications
- Adversarial attacks could have severe negative impact on safe operation
- State-of-the-art adversarial attacks mainly focus on autonomous vehicle instead of UAV policies
- Adversarial attacks on UAVs
 - should be **physically realizable**,
 - with a special focus on **low-power AI**,
 - and **exploit the hardware properties** of a quadrotor

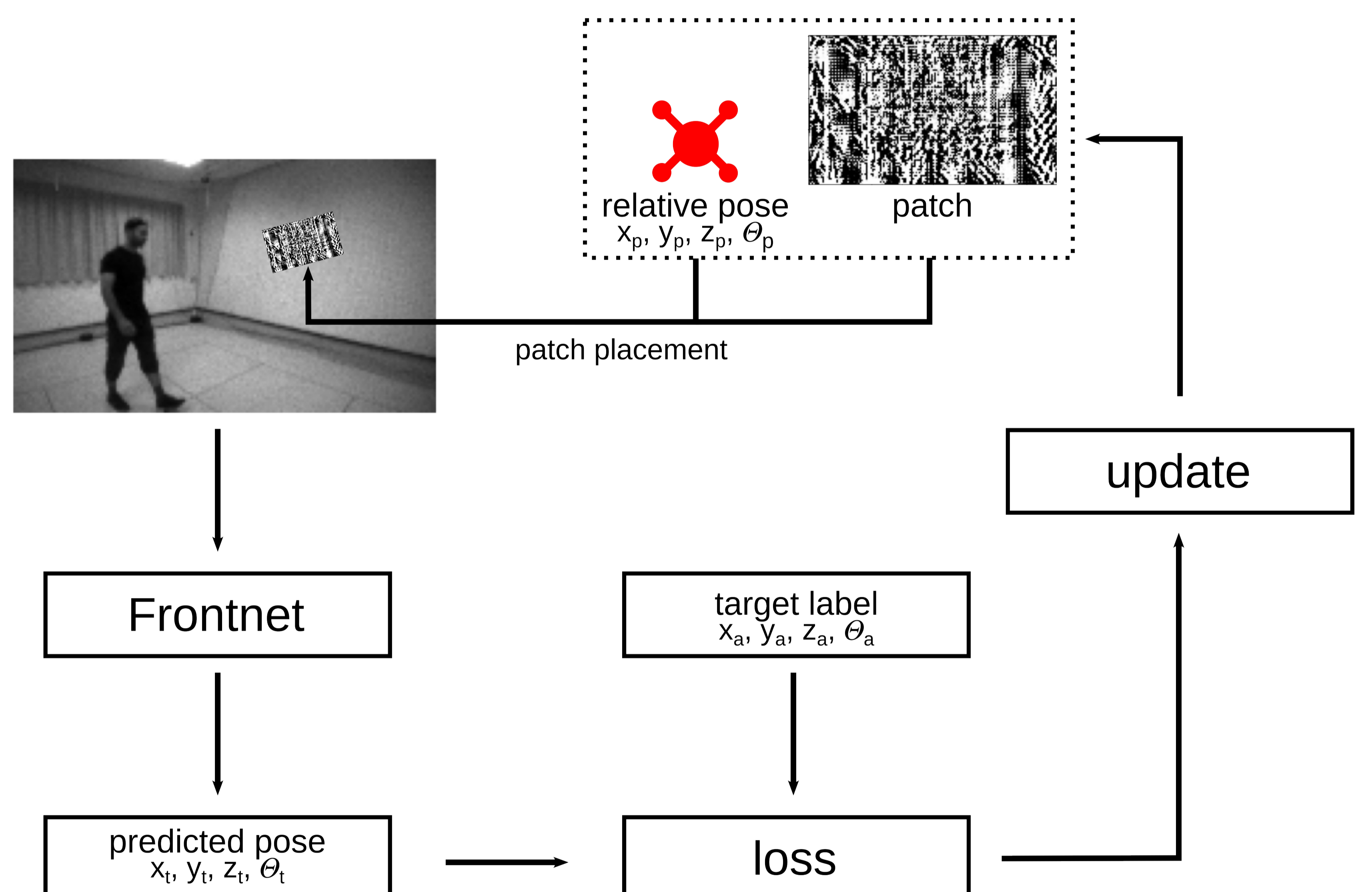
Goal



- Case (a) depicts the **regular scenario**: the victim drone is following a human target
- A neural network is predicting the relative pose of the human from camera images
- The prediction directly influences the control of the victim drone
- Case (b) depicts the **attack scenario**: an attacker drone has **full control over the victim drone**

Adversarial Attack

- We perform a **white-box adversarial attack**: the parameters of the neural network are known
- The adversarial attack will not influence the whole camera image but a small area and thus create an **adversarial patch**
- The fully optimized patch will be printed and attached to the attacker drone
- We both optimize the pixel values of the patch and the **relative pose** of the carrying attacker drone to enable control over it
- Using a patch placement algorithm, we create the artificial, **manipulated input** to the neural network
- We calculate the **l_2 loss** between the **predictions** of the neural network and a predefined adversarial **target label**
- This loss is used to calculate the gradients w.r.t. the patch and relative pose and perform a **gradient decent step to update both**



Future Work

- Simultaneous optimization of multiple patches for full control in x, y, and z direction
- Real-world experiments
- Investigate impact on quantized neural net (low-power AI)

References

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- [2] Giernacki, W. et al. (2017). Crazyflie 2.0 quadrotor as a platform for research and education in robotics and control engineering. 22nd International Conference on Methods and Models in Automation and Robotics (MMAR), pp. 37-42. doi: 10.1109/MMAR.2017.8046794