

Reproducibility Study on Adversarial Attacks Against Robust Transformer Trackers







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Adversarial Robustness

Adversarial perturbations deceive neural networks, leading to inaccurate outputs. Such adversarial attacks have been studied for various vision tasks ranging from:



Transformers have recently boosted object tracking performance, but their robustness under adversarial attacks remains unclear.

Research Goals & Contributions

- Goal 1. Assess effectiveness of attacks on different trackers.
- Goal 2. Compare transformer and non-transformer trackers.
- Goal 3. Evaluate robustness across different perturbation levels and output types (bounding box vs. binary mask).

A. Adversarial Attacks per Tracker Output

• TransT-SEG Tracker on VOT2022



Main Takeaway: The attacks applicable to transformer trackers have more impact on the accuracy of the object mask than the bounding boxes on VOT2022ST dataset.

B. Adversarial Attacks per Perturbation Level

• White-box attacks under perturbation level shift



RTAA Performance against TransT Tracker on UAV123

Main Takeaway: For RTAA attack, adding a higher perturbation level generates more perceptible noise for all frames, which damage more the overall tracking performance.

B. Adversarial Attacks per Perturbation Level

• White-box attacks under perturbation level shift



SPARK Performance against TransT Tracker on UAV123

Main Takeaway: Increasing the perturbation level on SPARK attack setting results in more superperturbed regions, i.e. regions with perceptible noise. C. Adversarial Attack per Upper-Bound

• Black-box attack under various upper bounds

IoU Performance against ROMTrack Tracker on UAV123



Main Takeaway: The outcome of the IoU attack is sensitive to its initialization. The evaluation process may take a long time due to unsuitable initialization point.

D. Transformer versus Non-transformer Trackers

- ROMTrack and MixFormer show higher robustness.
- SPARK and RTAA substantially degrade tracking performance; transformer models are generally more resistant than non-transformers.
- Despite transformer trackers (ROMTrack, TransT, and MixFormer) showcasing the top-3 performance, their evaluation scores more notably decreased after applying the IoU method.

Conclusion and Future Directions

- Conclusion: Transformer trackers exhibit superior robustness but require further exploration of targeted adversarial techniques.
- Future Work: Development of more sophisticated attacks to effectively challenge these models and enhance robustness testing.