

Christian Wilms¹, Rafael Heid¹², Mohammad Araf Sadeghi¹, Andreas Ribbrock², and Simone Frintrop¹

¹ University of Hamburg, Department of Informatics, Computer Vision Group ² zeroG GmbH, Germany



Which Airline is This? Airline Logo Detection in **Real-World Weather Conditions**

Motivation

Airline Logo Detection:

Localize and classify all airline logos on airplane tails in an image

General Logo Detection

- State-of-the-art systems mostly based on off the shelf object detectors like Faster R-CNN
- Datasets feature mainly clean images captured under controlled conditions

Open Questions

- Are off the shelf object detectors enough?
- · What about real-world weather like rain or fog?





Faster R-CNN [1]

Our system with proposed DA

Dataset for Airline Logo Detection

- Simple Image: Captured under clear weather conditions
- Difficult Image: Mostly captured under adverse weather conditions (e.g. fog, rain, ...)
- 2 Test Splits: One test split with only simple images, one with only difficult images

| Split | Images | Annotations | Classes | Simple Images | Difficult Images |
|---------|--------|-------------|---------|------------------|---------------------|
| Train | 4396 | 4517 | 41 | ✓ | |
| Val | 625 | 665 | 41 | ✓ | |
| Test1 | 1290 | 1397 | 41 | ✓ | |
| Test2 | 252 | 459 | 13 | | ✓ |
| Overall | 6563 | 7038 | 41 | | |





Simple Images



Difficult Images

Airline Logo Detection System



- State-of-the-art AttentionMask [2] for localizing logo proposals
- AttentionMask is adapted in terms of scales and output
- · Localization can be provided by the user as well (not used for the reported results)

Our system on simple images

Airline Logo Classifier

Classifier based on VGG

Optimized architecture

Data Augmentation

- 3 scenarios (fog, rain/dark clouds, raindrops)
- Learning free (no training data available)







Increase brightness

reduction

Local blurring

Results

Input Image

Green frames around images denote successful cases (correctly localized and classified), red frames denote failure cases (not correctly localized or

Airline logos found/not found by our system with DA on difficult images

















Evaluation

- mAP: Mean average precision
- DA: Our data augmentation strategy

| | Simple | Images | | Difficult Images | | | |
|------------------|--------|------------|--------|------------------|---------|------------|--|
| | | | w/o DA | | with DA | | |
| Method | mAP | mAP_{75} | mAP | mAP_{75} | mAP | mAP_{75} | |
| YOLOv3 [3] | 0.698 | 0.869 | 0.101 | 0.129 | 0.123 | 0.147 | |
| Faster R-CNN [1] | 0.659 | 0.826 | 0.118 | 0.102 | 0.128 | 0.117 | |
| Ours | 0.708 | 0.880 | 0.173 | 0.221 | 0.203 | 0.248 | |

- Tailored architecture works better than object detectors
- Adverse weather conditions have strong effect on the results

References

- [1] Ren, S.; He, K.; Girshick, R.; Sun, J.: Faster R-CNN: Towards real-time object detection with region proposal networks. In: NIPS (2015)
 [2] Wilms, C.; Frintrop S.: AttentionMask: Attentive, efficient object proposal generation focusing on small objects. In: ACCV (2018)
 [3] Redmon, J.; Farhadid, A.: YOLOV3: An incremental improvement. In: arXiv (2018)

Acknowledgment

e thank the administrators of planespotters.net for allowing us to use their database for this research.