

interim - Drone For Your Safety

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Project Review

Motivation

The critical scarcity and high risk involved in collecting real-world drone footage of humans trapped in active wildfires.

Specifications

A YOLOv8 object detection model trained on a dataset, generated by a Python script that injects synthetic fire and atmospheric effects into real imagery with automatic labeling.

Novelty

This project introduces a novel 2D data augmentation pipeline that synthesizes extreme fire and smoke scenarios, enabling the training of Search and Rescue drones to detect victims in low-visibility environments without hazardous data collection or complex 3D engines.

Changes from proposal

- we decided to use a data set from ROBOFLOW instead of generating with CARLA
- We decided to focus on generating specifically fire and smoke.



Previous Work

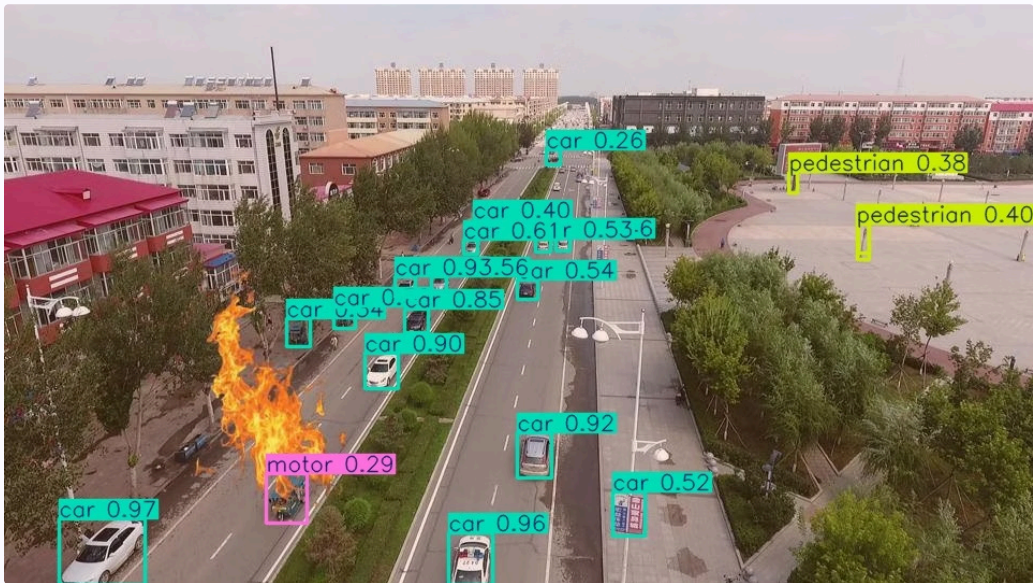
title/year	task	methods	data	results	relation to our project
<i>Understanding the Impact of Image Quality and Distance of Objects to Object Detection Performance (2023)</i>	Analyzing how image resolution, compression, and object distance affect detection accuracy and costs.	Tested spatial resizing and JPEG compression levels: evaluated optimal resolutions for constrained bandwidth/storage.	A custom dataset created by combining the TJU and Eurocity datasets with varying resolutions.	Proved that higher spatial resolution significantly enables a greater detection range for distant objects.	Provides scientific proof that reducing quality (like our 2D/ compression pipeline) is a valid trade-off for drone storage.
<i>Robustness Analysis of Object Detection Models under Weather Degradation (2025)</i>	Analyze how image quality degradation (noise, blur, reduced resolution, weather effects) impacts object detection performance.	Evaluation of object detection models (including YOLO) under controlled image degradations such as blur, noise, and fog, combined with data augmentation experiments.	Real-world adverse weather datasets and synthetically degraded images simulating realistic visual conditions.	The study shows that image degradation significantly reduces detection accuracy, while training with degraded and augmented images improves model robustness and generalization.	Provides scientific evidence that training on degraded images (blur, noise, reduced resolution) improves object detection robustness under real-world adverse conditions.

Dataset

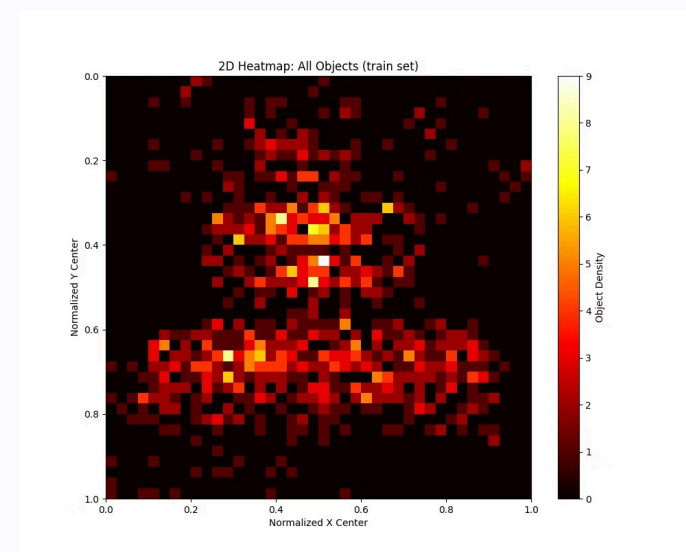
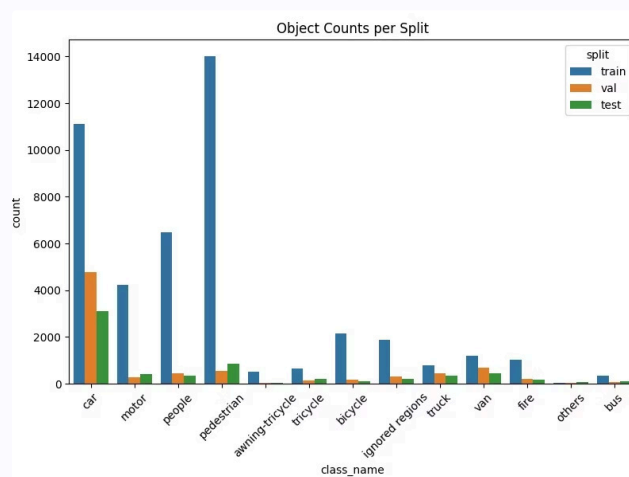
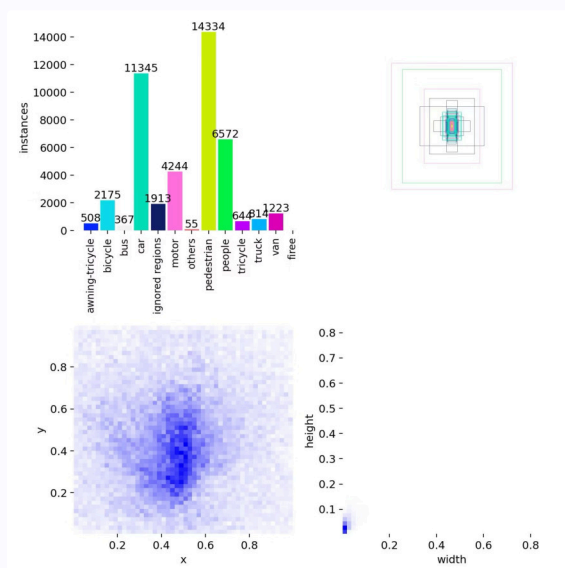
before



after



Baseline solution and results



Plan

Apply stratified splitting to ensure consistent class distribution across sets and prevent model evaluation bias using scikit-learn

To decrease the quality of the fire images

Generating the rest of the dataset and train it

Make sure the fire is spread across the entire width of the image

Generating 10% of the train dataset with ControlNet in order to enhance the dataset's photorealism

Evaluation of the final dataset

Thank you