

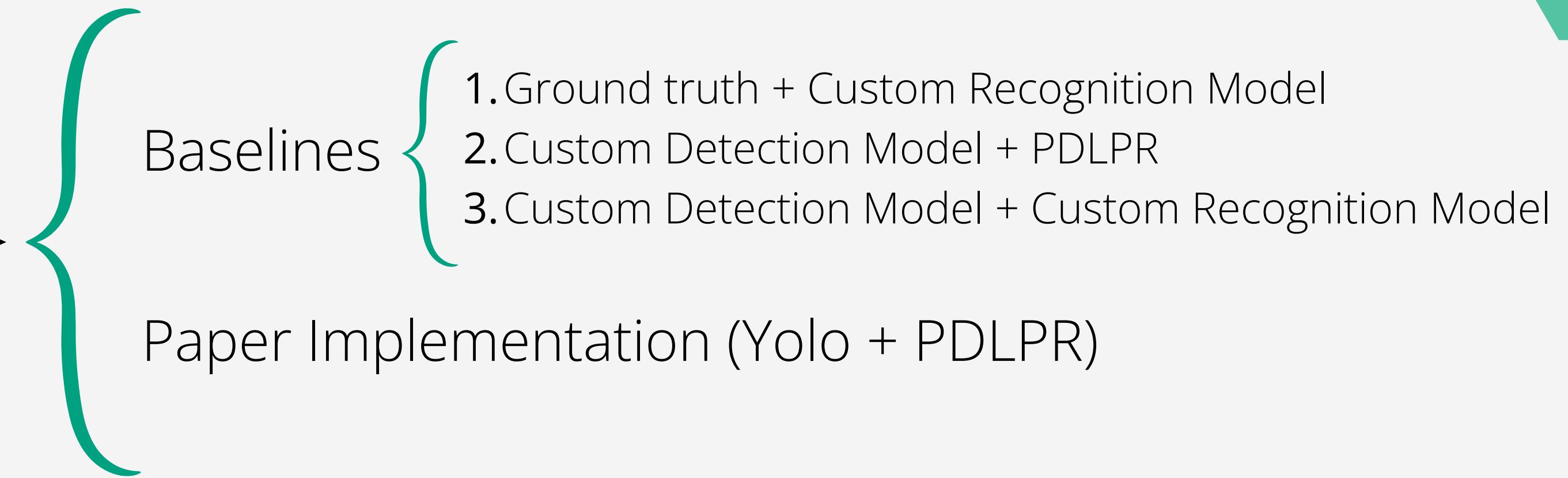
Group

- Filippo Casini
- Giovanni Zara

Computer Vision Project

Outline

2 Tasks →

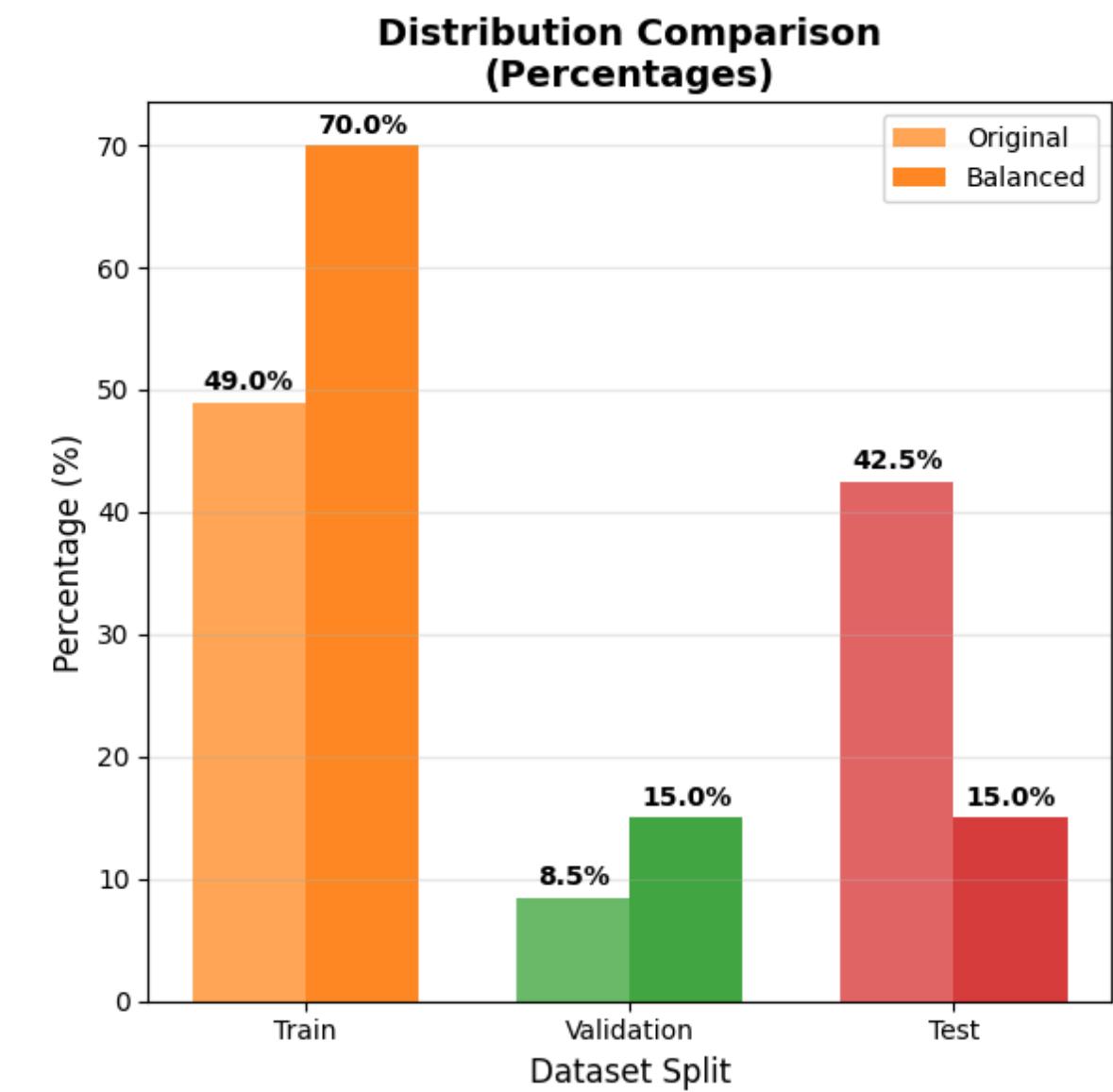
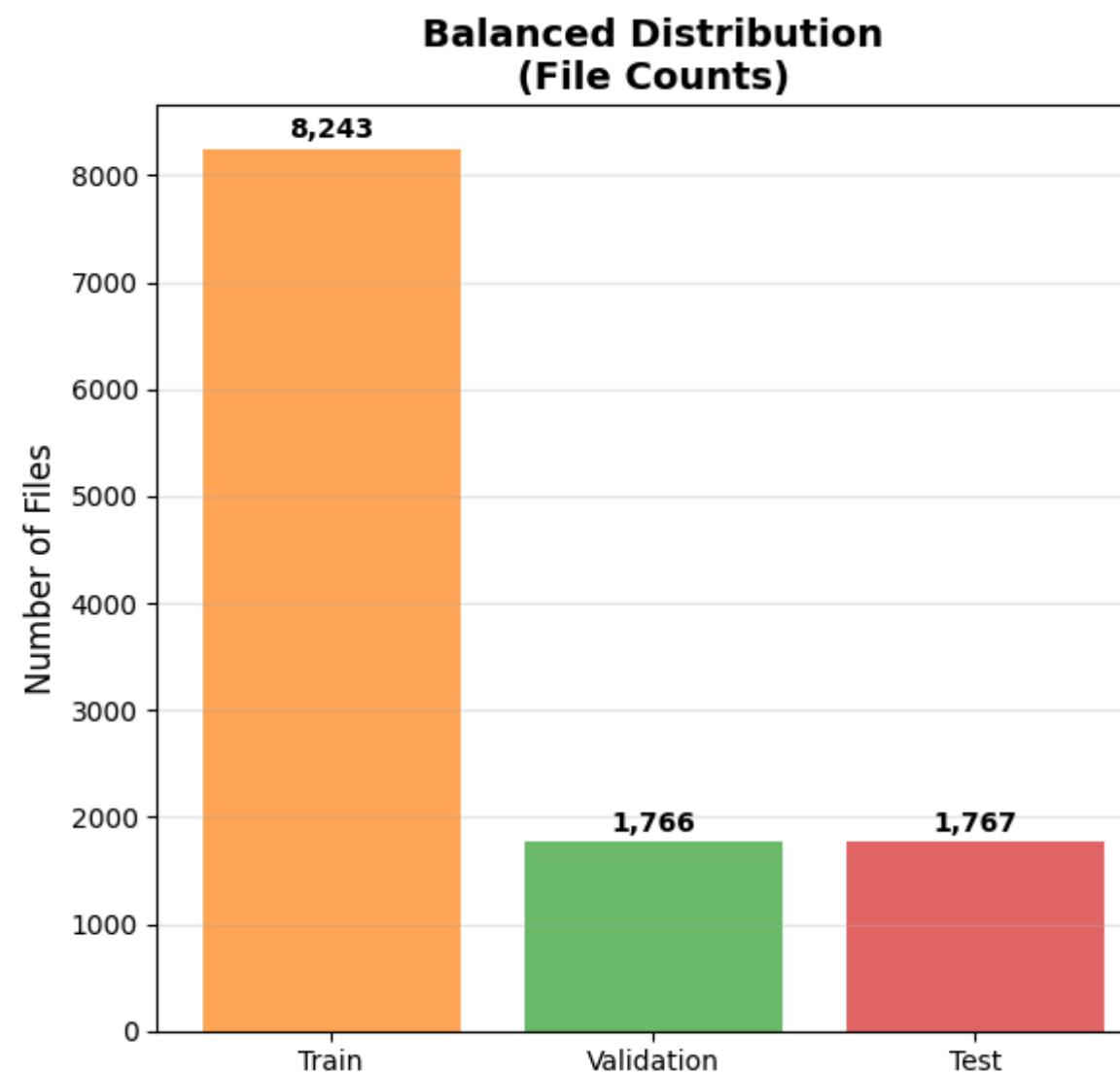
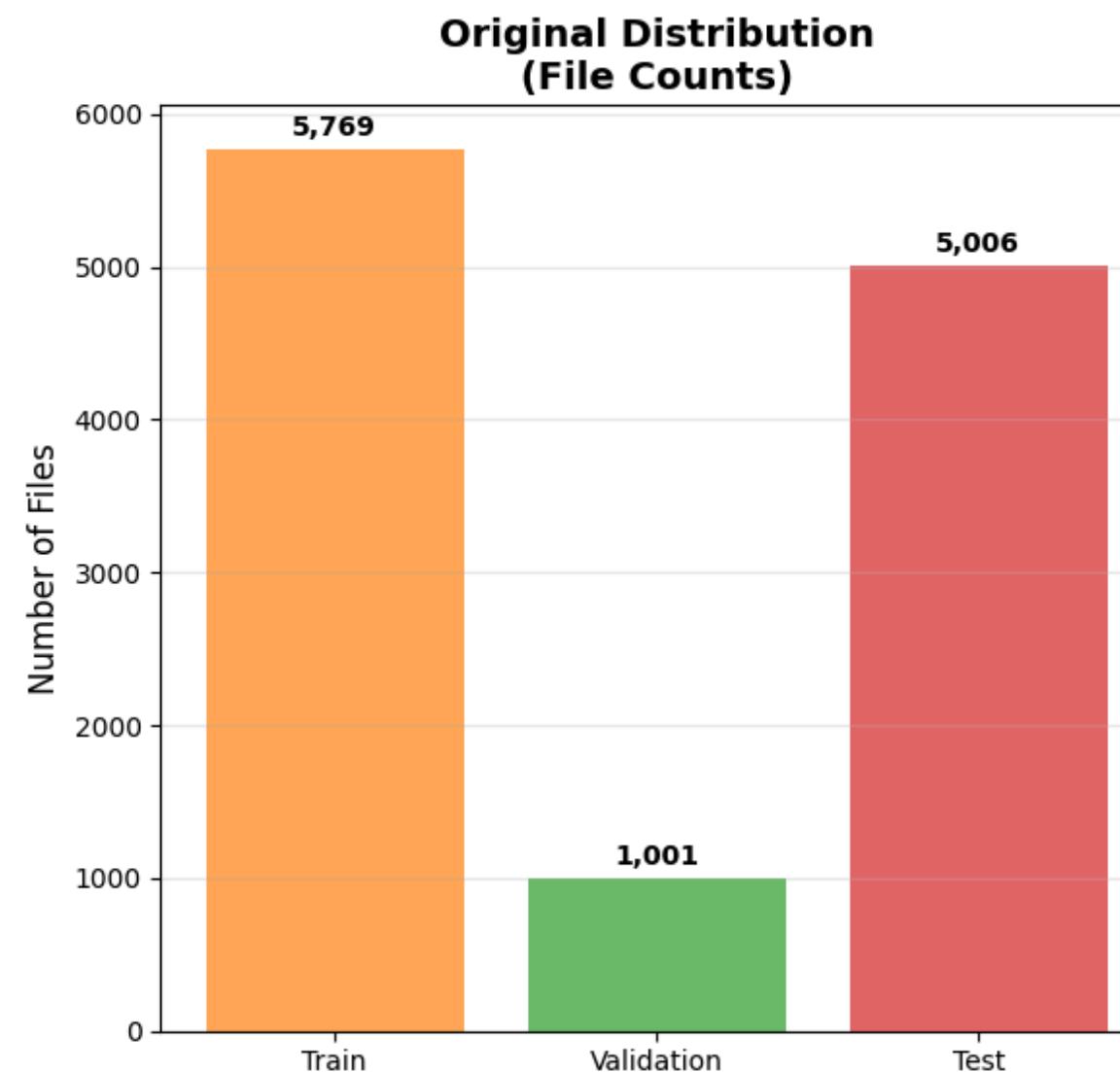


Baselines:

- 1. State of the art + Custom Model
- 2. Custom Model + State of the art
- 3. Full custom Pipeline

Data

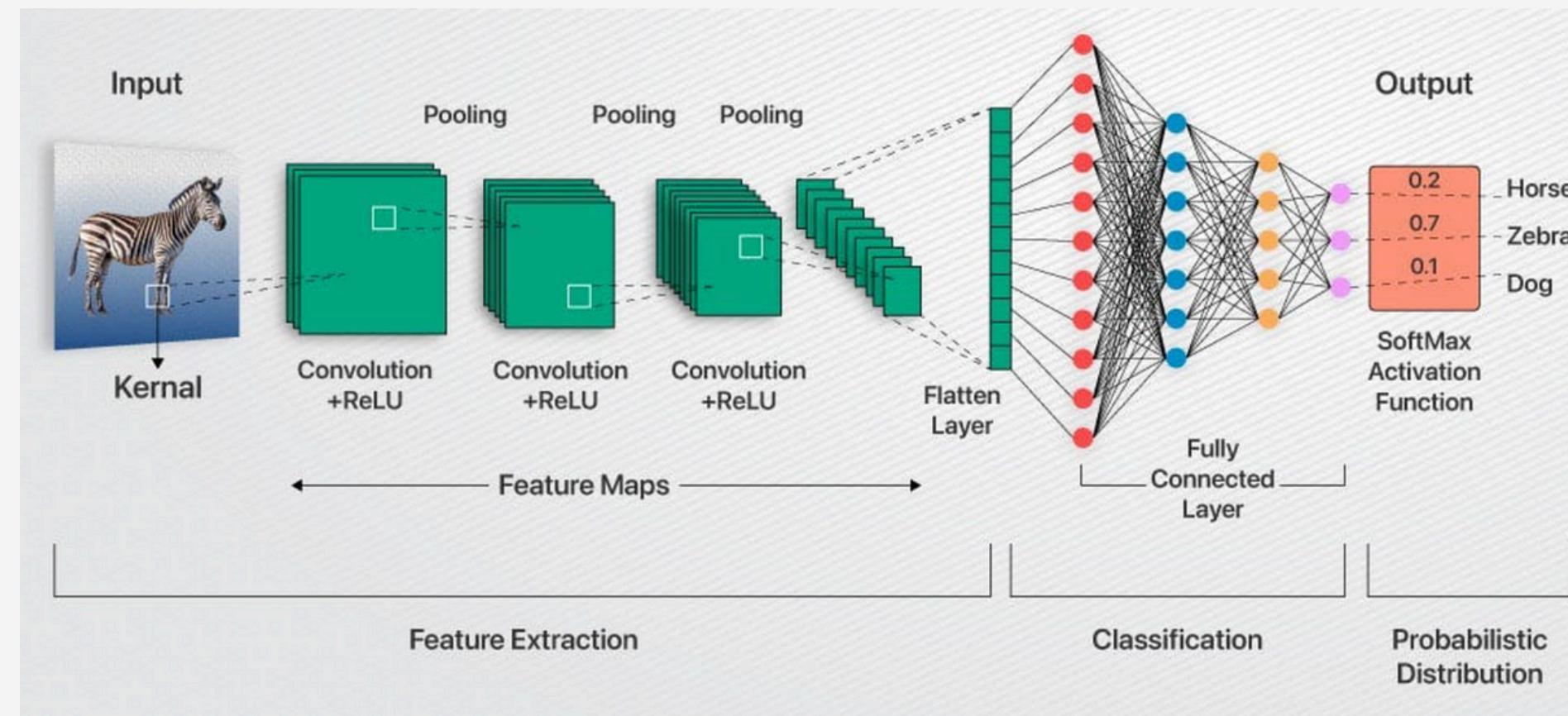
We used the CCPD_green dataset, containing about 11k samples rebalanced in the following way:



Custom Detection Model

Fine tuning of `fasterrcnn_resnet50_fpn` with:

- 500 proposed regions
- 2 classes (1 for the car plate and 1 for the background)
- Intersection over Union to evaluate the results
- 1 training epoch



Average Intersection over Union
on Test Set = 0.87

Custom Detection Model

Example of the Model

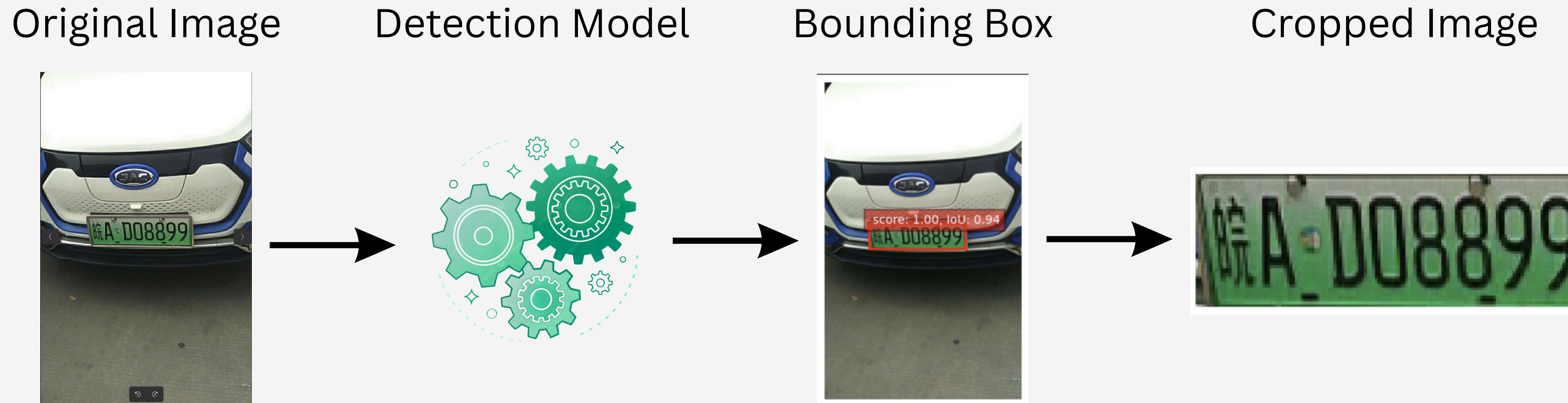
Ground Truth Bounding Box



Predicted Bounding Box

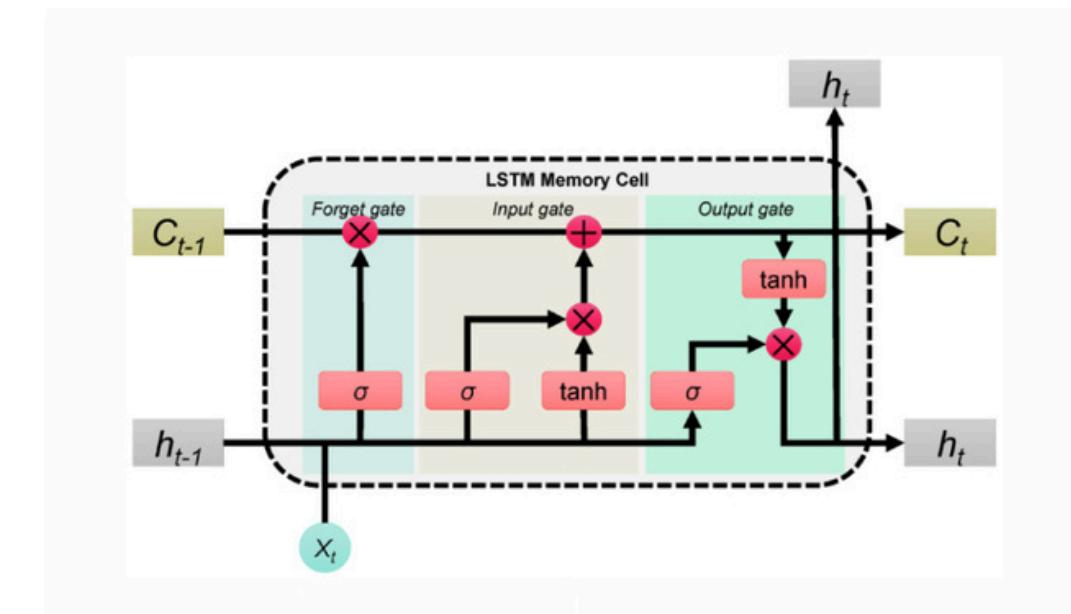
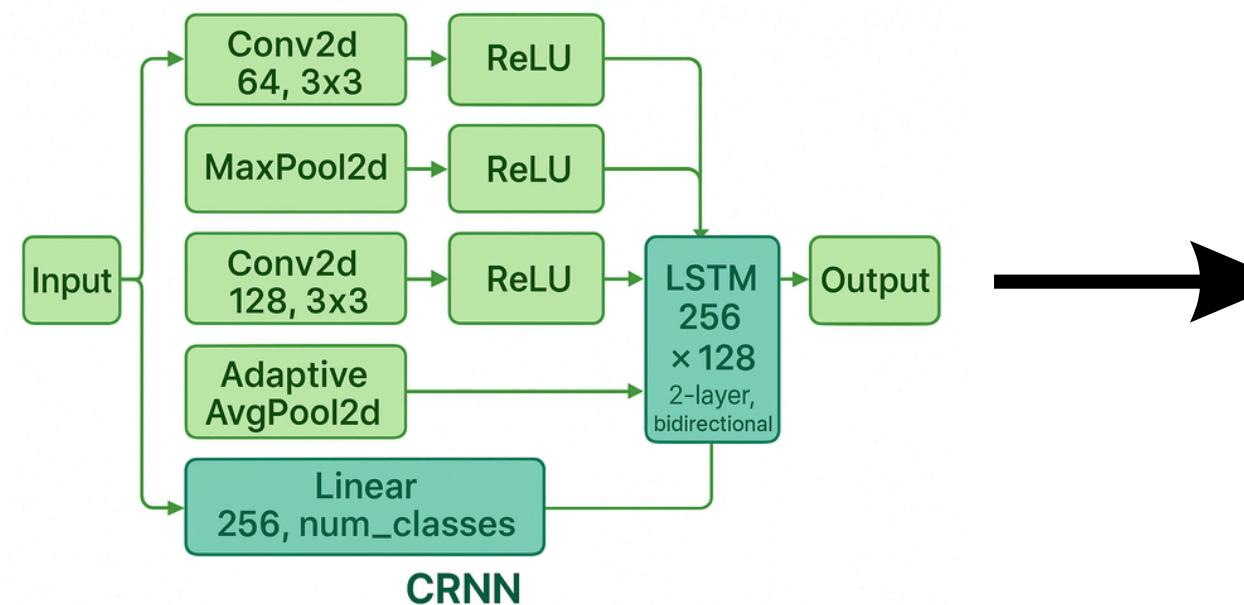


Custom Detection Model Pipeline



Custom Recognition Model

Architecture and Pipeline



LSTM architecture

Training Parameters: 1.177.925

Cropped Image



Recognition Model

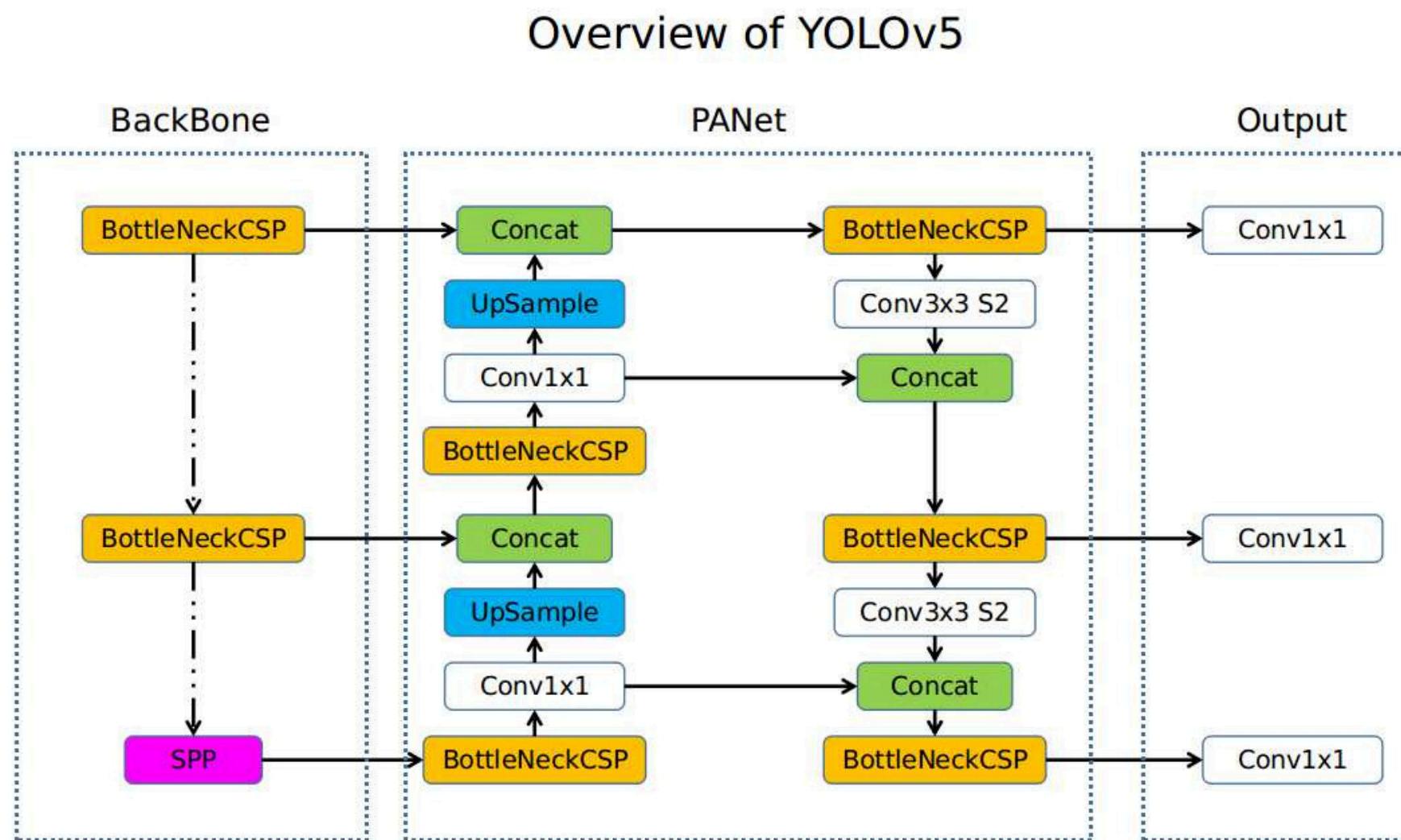


Predicted Text

皖AD08899

Yolo Finetuning

We finetuned Ultralytics YoloV5 to distinguish between car plate and background, applying a light data pre-processing.



Pretrained Checkpoints

Model	AP ^{val}	AP ^{test}	AP ₅₀	Speed _{GPU}	FPS _{GPU}	params	FLOPs
YOLOv5-s (ckpt)	35.5	35.5	55.0	2.1ms	476	7.1M	12.6B
YOLOv5-m (ckpt)	42.7	42.7	62.4	3.2ms	312	22.0M	39.0B
YOLOv5-l (ckpt)	45.7	45.9	65.1	4.1ms	243	50.3M	89.0B
YOLOv5-x (ckpt)	47.2	47.3	66.6	6.5ms	153	95.9M	170.3B
YOLOv3-SPP (ckpt)	45.6	45.5	65.2	4.8ms	208	63.0M	118.0B

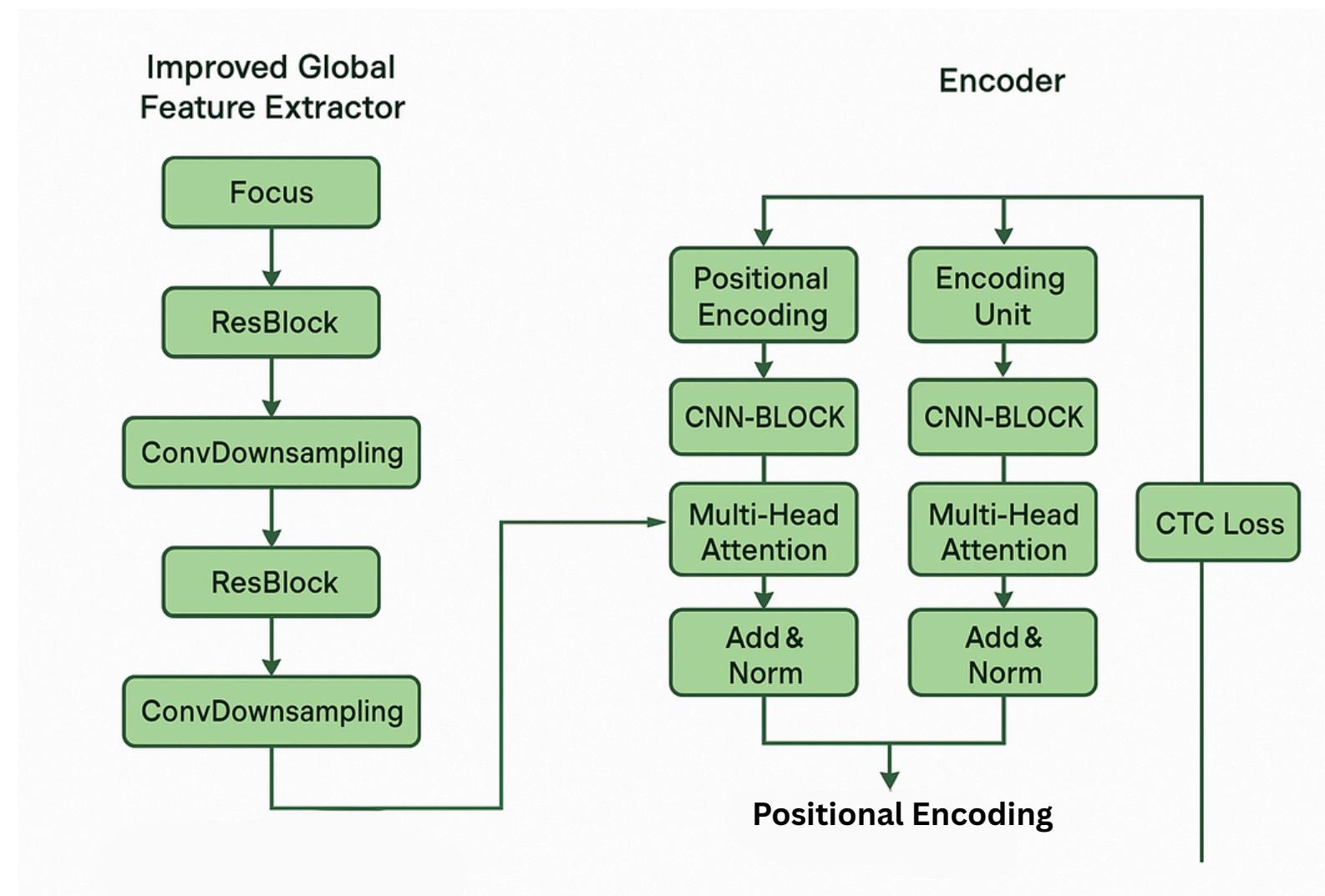


Yolo Metrics

Yolo FineTuned	Metrics
Number of Parameters	7.012.822
GFLOPs	15.8
Maximum Average Precision (IoU>0.5)	0.995
Maximum Average Precision (0.9<IoU<0.95)	0.902
Instances per Image	1

PDLPR Implementation

We reproduced the architecture described in the paper simplifying some over complex passages while trying to maintain an accurate text recognition pipeline.



First Baseline

Ground truth + Custom Recognition Model

For this baseline we simply cropped the images using the ground truth and applied the Custom Recognition Model.

The goal of this Baseline is to show how important is the recognition part with respect to the detection one.

Cropped image with GT



Recognition Model



Predicted Text

皖AD35199

Second Baseline

Custom detection Model + PDLPR

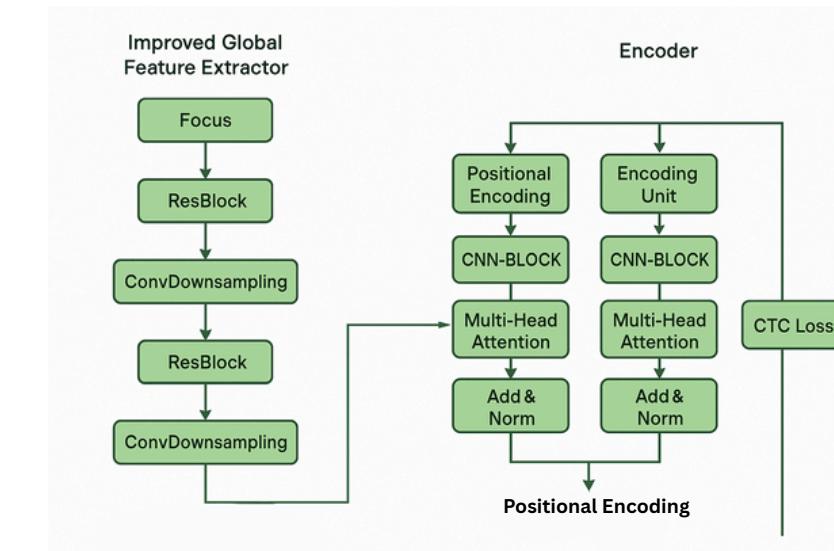
For this baseline we simply cropped the images using the detection model and applied PDLPR, the state of the art recognition model introduced in the paper.

The goal of this Baseline is to show how important is the detection part with respect to the recognition one.

Cropped Image with
detection Model



PDLPR



Predicted Text

皖AD08899

Third Baseline

Custom Detection Model + Custom Recognition Model

For this baseline we simply cropped the images using the detection model and applied the Custom Recognition Model.

The goal of this Baseline is to test how well our models were performing with respect to the paper implementation.

Cropped Image with
detection Model



Recognition Model



Predicted Text

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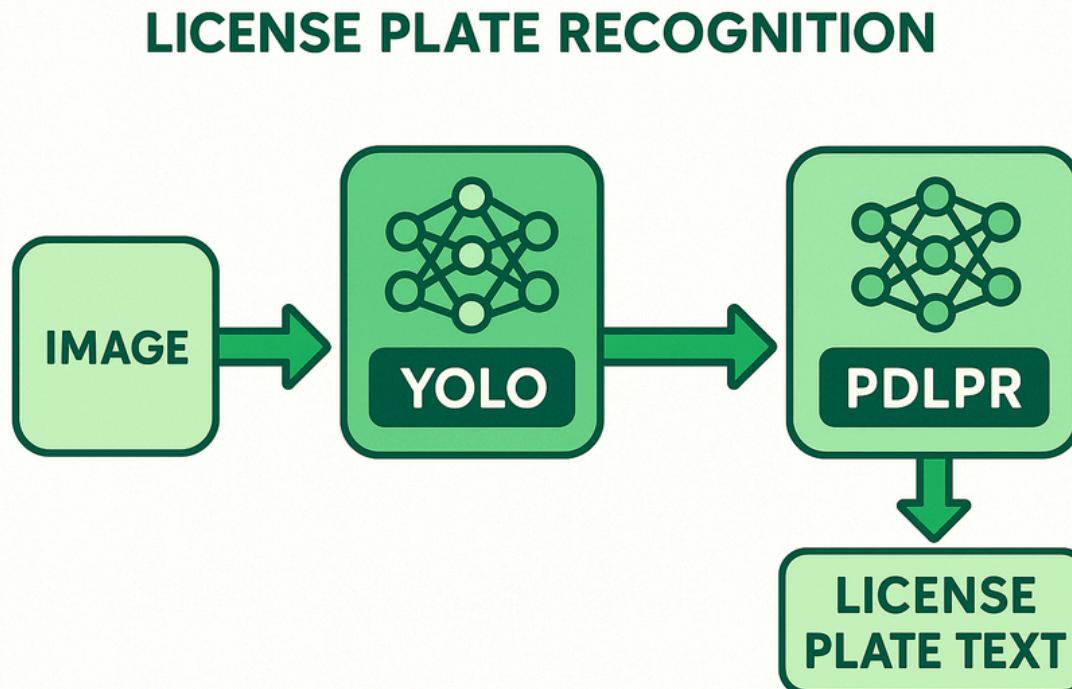
Paper Implementation

As said earlier, the detection part has been carried over by using a lightweight Yolo version, while the recognition part has been simplified by using the CTC Loss instead of a complex Auto Regressive Decoder.

We have been able to do so due to a limited-domain dataset and a less flexible approach with respect to the one described in the paper.

Online Implementation

Since this model should be deployed in real time applications we also created an online version that takes the original image and extracts the text in a direct way.



Results

Models	Plate Accuracy on Test Set	Character Accuracy on Test Set
First Baseline	0.8970	0.9680
Second Baseline	0.8273	0.9670
Third Baseline	0.8975	0.9669
Paper Implementation	0.9184	0.9778

More Metrics

Models	Province Accuracy	Alphabet Accuracy	Length Error Rate
First Baseline	0.9904	0.9904	0.0424
Second Baseline	0.9853	0.9836	0.0102
Third Baseline	0.9987	0.9881	0.0470
Paper Implementation	0.9911	0.9916	0.0006

Paper Vs Our Implementation

Metrics	Paper
Plate Accuracy on Test Set	0.9940
Speed (FPS)	159.8

Metrics	Our Paper Implementation
Plate Accuracy on Test Set	0.9184
Speed (FPS)	110

Thanks For Attention