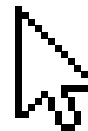




# License Plate OCR

## Computer Vision Project



Made by: Tomaily Tatiana, Gorelskii Roman, Sergazin Iskander  
Supervisor: Kopylov Ivan



# 01 Introduction

Description. Application. Problem definition. Technology stack.

# 02 Tasks Division

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## 08 Deployment & Demonstration

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## 09 Conclusions & Further Perspectives

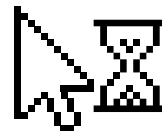
Key results. Unsolved problems. Ways to improve the model.

## 10 References



01

# Introduction



## Description

**Manual controlling of vehicles** which are allowed to enter the area is often **slow, error-prone, and inconvenient**. Our project **solves this problem by developing** an **automatic license plate recognition system** that detects a car's license plate and checks if it exists in a **pre-approved database**.

## Application

Potentially, it can **be integrated with automated gates**, allowing only authorized vehicles to pass through them. Potential users include **residential complexes, corporate offices, and event organizers seeking efficient access control**.



## Problem definition

The main goal of this project is to **automatically recognize vehicle license plate numbers from camera footage** or images and **check whether the detected number exists in a predefined access list.**

## Technology stack

- Python
- OpenCV
- Telegram Bot Api
- OCR method implementation



02

# Project stages Tasks Division



	Roman	Iskander	Tatiana
Review of existing solutions and data collection			
Bot and Database implementation			
Model deployment, training and metrics improvement			
Matching plates with database, returning a response			
Prototype development			





# 03

## Review of Existing Solutions



## Approaches already used

- **Traditional methods** using OpenCV with **contour** detection or **edge** detection
- Deep learning-based **object detection models like YOLO**
- Deep learning-based CRNN (**Convolutional Recurrent Neural Networks**) models
- **Optical character recognition** (OCR) – include Tesseract OCR (open-source engine), EasyOCR

## Our method

**Lightweight and fast license plate OCR pipeline** inspired by the Fast Plate OCR approach

- **Efficiency:** optimized for real-time processing
- **Simplicity:** easy to deploy and maintain
- **Accuracy in Diverse Conditions:** better generalization
- **Scalability:** allows future enhancements



# 04 Data Preparation



## Description of the dataset and Visualization

**Source:** we used an open-source dataset from the Fast Plate OCR GitHub repository.

This dataset contains images of vehicles with visible license plates.



**Quantity:** 2874 for training and 764 for validation

**Format:** images in .jpg format, Argentina plates

## Data preprocessing

The augmentation technique is used to get better predictions



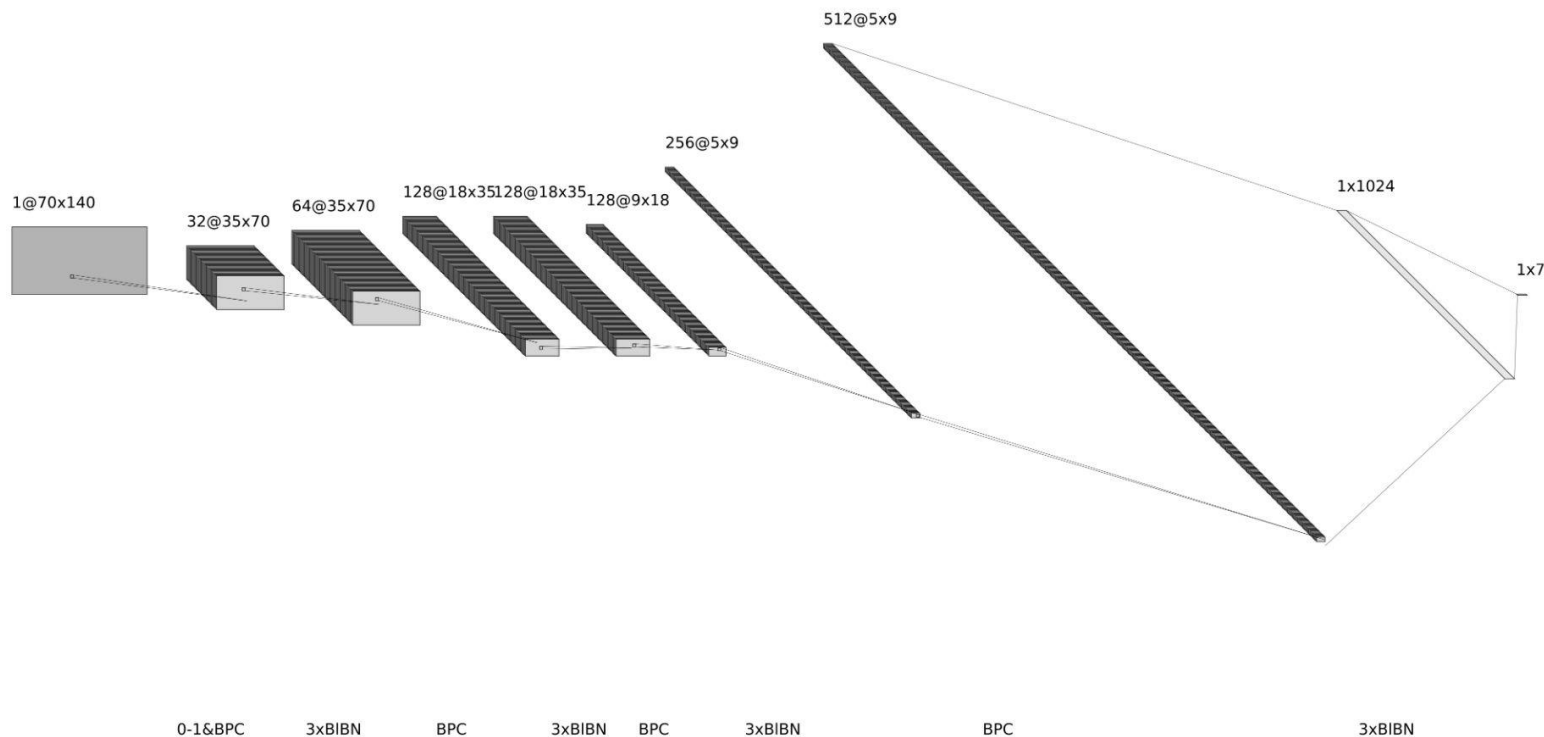
Data before and after augmentation



# 05 Model Development

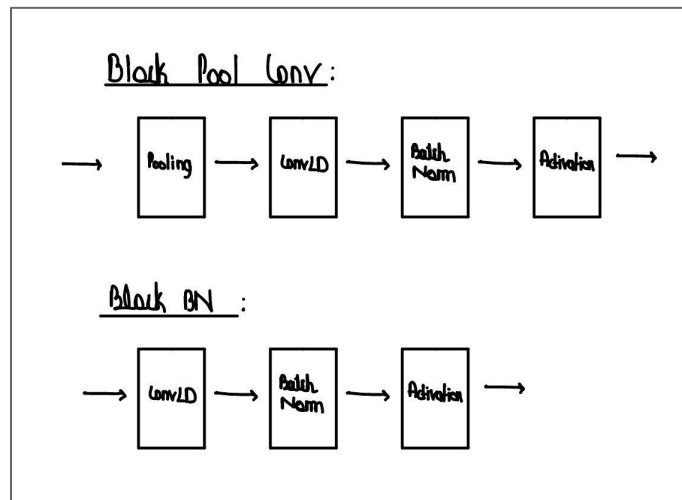
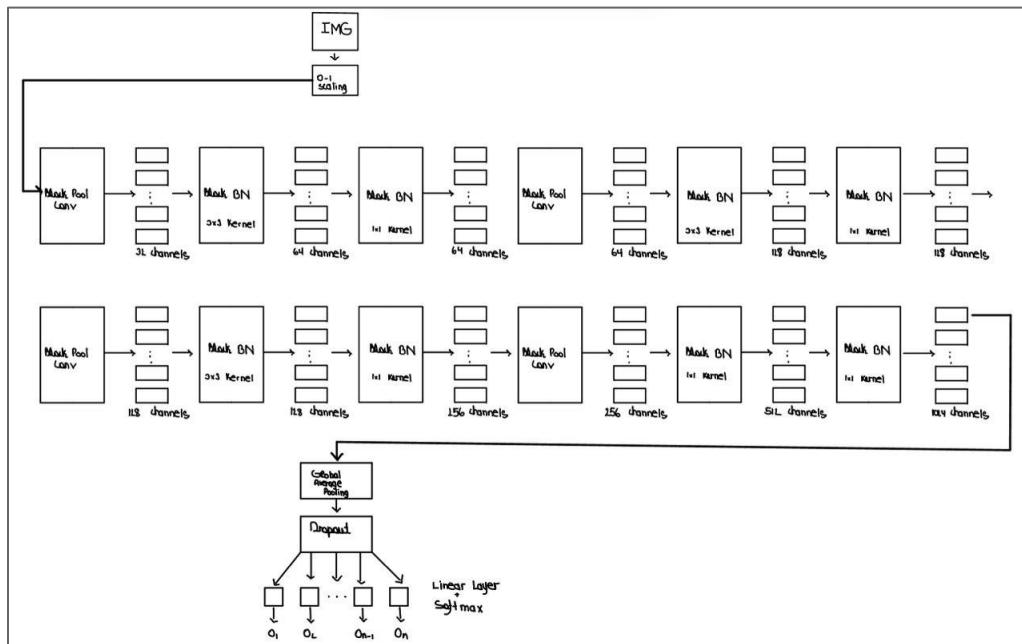


# Neural Network Architecture





# Neural Network Architecture (2)





# Why the approach was chosen?

## Fast Plate OCR

- **Lightweight** model (Throughput 400 frames/s)
- Can be **trained from zero** (8hrs 2500 img)
- **Pre-trained** models (argentinian/european/world) plates







# Strategy & Hyperparameters

## early\_stopping

Stop training when  
val\_plate\_acc does  
not improve over  
100 epochs

## epochs

750

## reduce\_lr

0.5 reduction in LR  
when no improvement in  
val\_plate\_acc

## Config File

7 characters  
predicted. Input image  
70×140

## batch\_size

128

## ReLU & MaxPooling


Activation f'n &  
Pooling Layer (can  
be chosen)



# 06

## Implementation and Training

# Code & Environment

- 
- Python 3.10
  - requirements.txt
  - GitHub

## Github repositories:

LicensePlateDetection (Our project) -  
[https://github.com/RomanGorelsky/CV Project Plates](https://github.com/RomanGorelsky/CV_Project_Plates)

FastPlateOCR -  
<https://github.com/ankandrew/fast-plate-ocr>



# Training Process

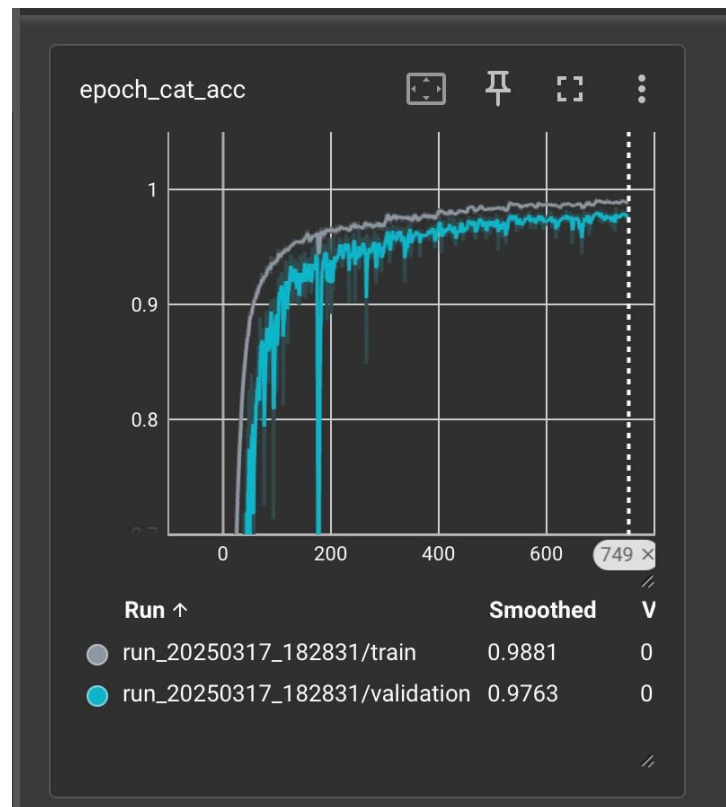
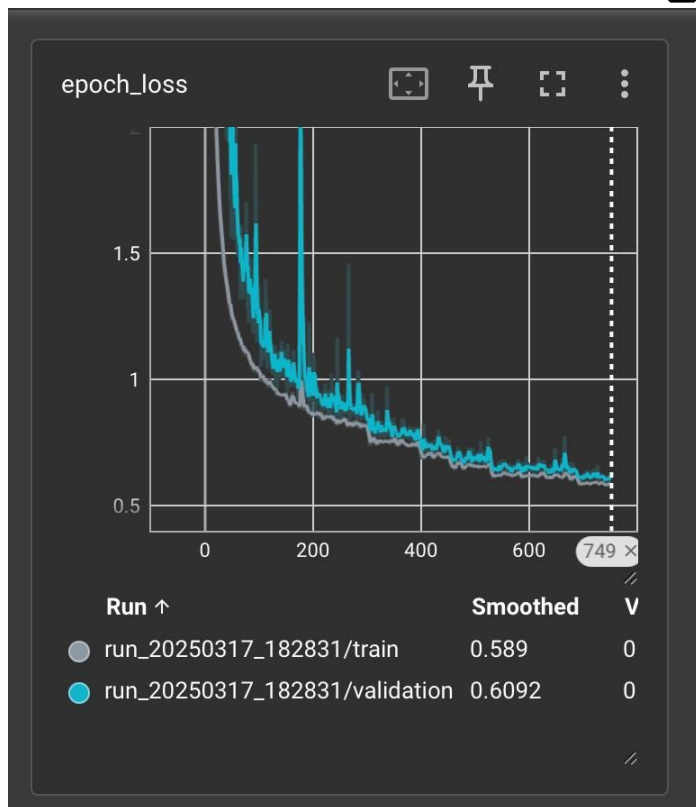
## Training

- **Loss:** Mean categoricalCrossEntropy over each license plate digit
- **Optimizer:** Adam

$$\mathcal{L} = - \sum_{i=1}^C y_{\text{true},i} \cdot \log(y_{\text{pred},i})$$



# Training Process (2)





# 07 Model Evaluation

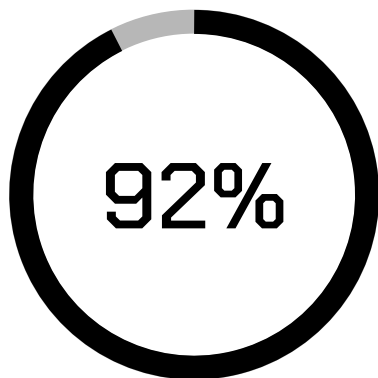


# Metrics

	Explanation
<b>Accuracy</b>	Plate accuracy over whole dataset
<b>Plate Precision</b>	Average character precision within incorrect plates
<b>Top 3-k</b>	Whether correct character in top-3 model predictions



# Testing on data



Accuracy

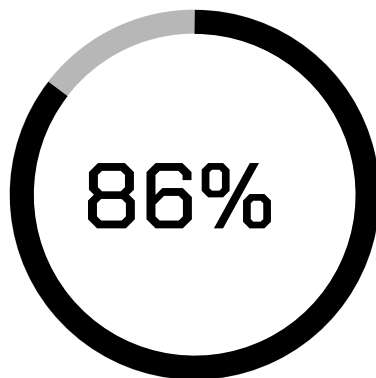
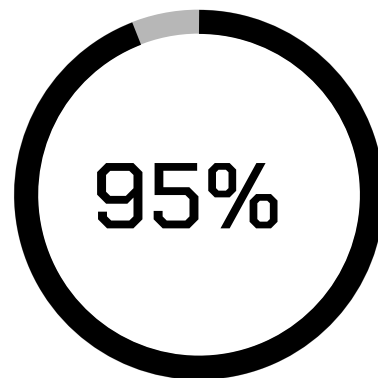


Plate  
precision

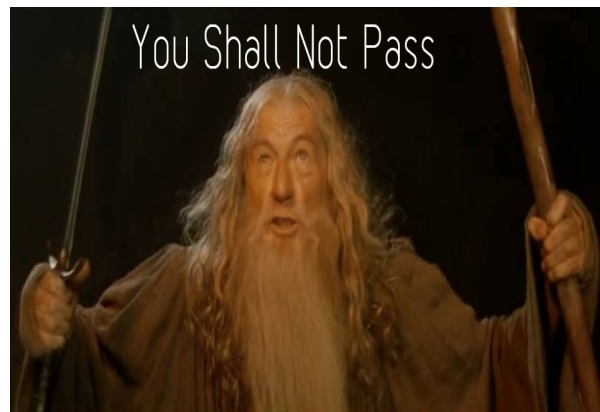


Top 3-k

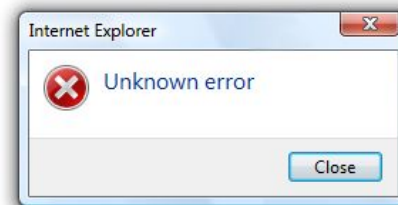




# Error interpretation



Model



- **Confuses letters** (Q/O, U/V)
- **Learns distribution** of data (LLLDDLL)
- **Poor performance** on plates from a **different** distribution (**Russian** plate dataset)



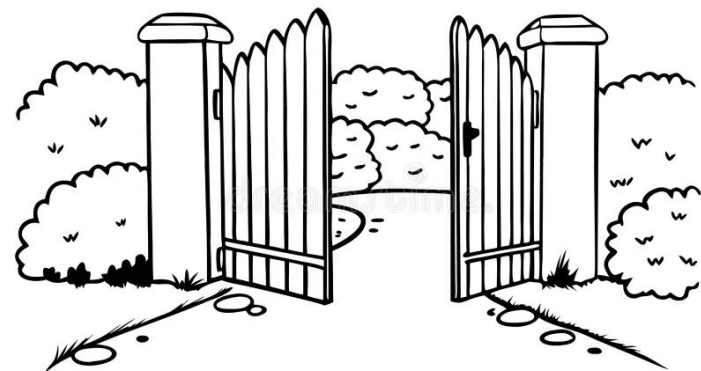
08

# Deployment and Demonstration



## Model usage in production

The model offers the possibility to implement the automatic **opening of gates and barriers** when the car arrives. The realisation consists in the **possession of the camera** located near the gate and **the database** in which license plate collection is stored. Such systems can be installed at various locations for personal or commercial use.





# API development

In order for a driver to have the possibility to add their plate for granted access the API with **telegram bot** was developed:

- Library **"pyTelegramBotAPI"** was used
- The bot was made **asynchronous** allowing multiple users to utilise it at once
- With the prompt **"Plate: ..."** the license plate is added to the database
- **Duplicates** are checked and not added
- All the plate numbers are stored in **uppercase**



The plate number is added into the database!



# Interactive

# prototype



09

# Conclusions & Further Perspectives



## Key results

- Project dataset
- Fast **Plate OCR** methodology **training**
- **Telegram** bot deployment
- **Interactive prototype** implementation

## Unsolved problems

- **No model versatility**
- Simulation and visualisation of **gate opening process**

## Ways to improve the model

- Train on license plates **from different countries**
- Grant access for **specific time length**
- **Extraction** of plates in the **wild**
- Implement the check of **similar plates combinations**
- Implement API ability to **delete specific plates**
- Integrate model into API to **add plates from photo**



# 10 References





Fast Plate OCR:

<https://github.com/ankandrew/fast-plate-ocr>

Requirements:

[https://drive.google.com/file/d/1p0sDDkpPJjHkerbEo5BDa0c3PVdW\\_UaJ/view?usp=sharing](https://drive.google.com/file/d/1p0sDDkpPJjHkerbEo5BDa0c3PVdW_UaJ/view?usp=sharing)

Project's GitHub:

[https://github.com/RomanGorelsky/CV\\_Project\\_Plates.git](https://github.com/RomanGorelsky/CV_Project_Plates.git)



# Thanks!

Do you have any questions?