

The University of Texas at Arlington

**Applied Statistics and Data Science** 

## ASDS-5305-001-DEEP LEARNING/NEURAL NETWORKS

## **Project Title:**

**Automatic Number Plate Recognition of Cars (ANPRC)** 

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#### 1. Problem Statement

The project aims to create an automated system for recognizing car number plates from images or videos. The system should function in diverse environments, handling challenges such as:

- Different lighting conditions
- Various plate formats
- Noisy, occluded, or motion-blurred images

#### 2. Dataset Overview

The dataset includes 453 images in JPEG and XML formats. The images consist of cars with visible license plates.

#### 3. Data Preprocessing

The preprocessing involved cleaning and augmenting data to ensure better model performance. This included annotating images to highlight license plates for training the model.

Total annotations: 225 Training annotations: 180 Validation annotations: 45

As shown in above figure we have 225 total annotations, that we have divided in 180 for training and 45 for validation.

	filepath	height	width	xmin	xmax	ymin	ymax
0	/content/images/N38.jpeg	768	1024	397	645	590	641
1	/content/images/N193.jpeg	615	802	274	502	2	59
2	/content/images/N112.jpeg	344	500	179	347	256	290
3	/content/images/N218.jpeg	332	500	176	301	212	238
4	/content/images/N73.jpeg	190	265	124	209	102	130

The figure above shows the pixel address of detected licence plate co-ordinates in the image.

## **Before and After annotation:**

Below we have shown the original image on the left and the annotated licence plate on the right.





#### 4. Model Building

The **Inception ResNetV2** model was used to detect and recognize text on license plates.

## **Key Features of Inception ResNetV2:**

#### 1. Inception Modules

 Efficiently processes features at multiple scales by combining convolution operations with different kernel sizes in parallel.

## 2. Residual Connections

- o Inspired by ResNet, these connections help avoid the vanishing gradient problem, making it easier to train very deep networks.
- They allow information to bypass certain layers, improving the flow of gradients during backpropagation.

## 3. High Accuracy

 Inception ResNetV2 offers high accuracy for detecting patterns in complex images, such as those with noise or occlusions.

## 4. Pre-trained Weights

 The model can leverage pre-trained weights from large datasets like ImageNet, reducing training time and improving performance on smaller datasets.

# **Testing Model:**



Image tensor shape: torch.Size([3, 729, 901])

## 5. Testing Results

The model was tested on new images, showing the ability to identify license plate text with varying confidence levels. Example outputs:

• Detected Text: KL 51 (Confidence: 0.87)

• Detected Text: K (Confidence: 0.59)

• Detected Text: 4999 (Confidence: 0.42)

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## 6. Conclusions

The model effectively identifies license plate text, although confidence levels vary depending on the text components and image conditions.

#### **References:**

#### **Dataset:**

https://www.kaggle.com/datasets/aslanahmedov/number-plate-detection

## **Record Link:**

https://mavsuta-my.sharepoint.com/:v:/g/personal/jkb5002\_mavs\_uta\_edu/ EfRV7l8HE\_1GpVG5RRx3RicB6tnEKKOvcLt7OGCHcOdAlQ?e=7e5h6s