



PRACTITIONER COMMUNITY

June 17th, 2021



Welcome



Agenda

We have prepared a very interesting agenda for today's session



AI News & Updates



Business Case Overview



AI Model Building



Q&A Time

Go to **www.menti.com** and use the code **3939 7971**

News and Updates



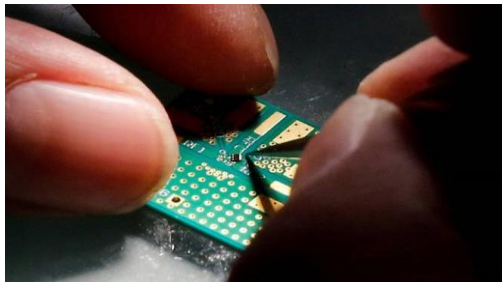
Facebook's new AI tool can replicate, replace and edit handwritten text using a single word in an image.

[Read more...](#)



By developers, for developers: Metabob's AI tool will forever change how developers approach debugging.

[Read more...](#)



Google is using AI to design chipsets in just six hours.

[Read more...](#)



News and Updates



Artificial Intelligence predicts how patients with viral infection will fare.

[Read more...](#)



Azure Maps Creator is now generally available in Microsoft Azure Service.

[Read more...](#)



News and Updates



An update by

Jair Ribeiro

Senior Business Analyst – Artificial Intelligence



Business Problem Statement

IRec – Image Recognition Emission Class

Requirement is to automate the mundane **manual audit of the trucks entering the factory.**

Why Audit?

- To capture the license plate numbers
- Check the Emission class of the truck
- Making sure trucks meet minimum emission class requirement

Pilot Site: **Tuve**

On a high-level following is the problem statement

Truck Emission Class detection

- Auto detect the Emission class of all incoming trucks into plant
- Create a report with logistics company details
- Share report on Power BI board for further actions



PL Sustainable transport system

Identified areas and actions to achieve by 2025



World class sustainable
transport system

“ To lead by example, we should have a world class sustainable transport system by 2025 ”

REDUCE TRANSPORT DEMAND

- Localization and **nearshoring** supply base
- Reduce **unnecessary shipments**
- Reduce **air shipment** acceptance
- 3-D printing for low running parts

OPTIMIZE TRANSPORATION

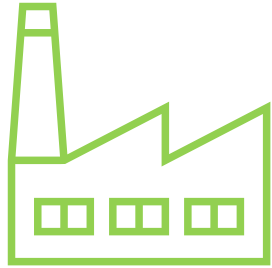
- Proactive transport network optimization
- Focus on load consolidation and filling degree in the supply chain
- Loadsharing with other shippers

IMPROVE FLEET EFFICIENCY

- **Emission class requirements**
- Utilize right mode of transports – increase usage of rail and sea
- Slow steaming ships and alternative fuel, High capacity vehicles
- LNG/CNG and early adaptation of electric trucks

*Reduce CO2
emissions
from freight
transports
per produced
unit by 30%
by 2025**

Emission Class – AS IS Process

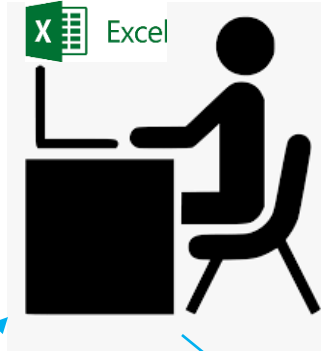


Trucks parked at Tuve
Factory-LX gate

Volvo Group IT

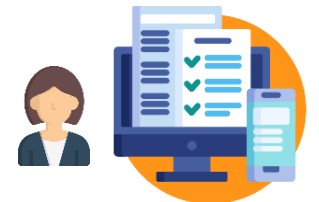


Manual Capture by
requesting the truck
Driver



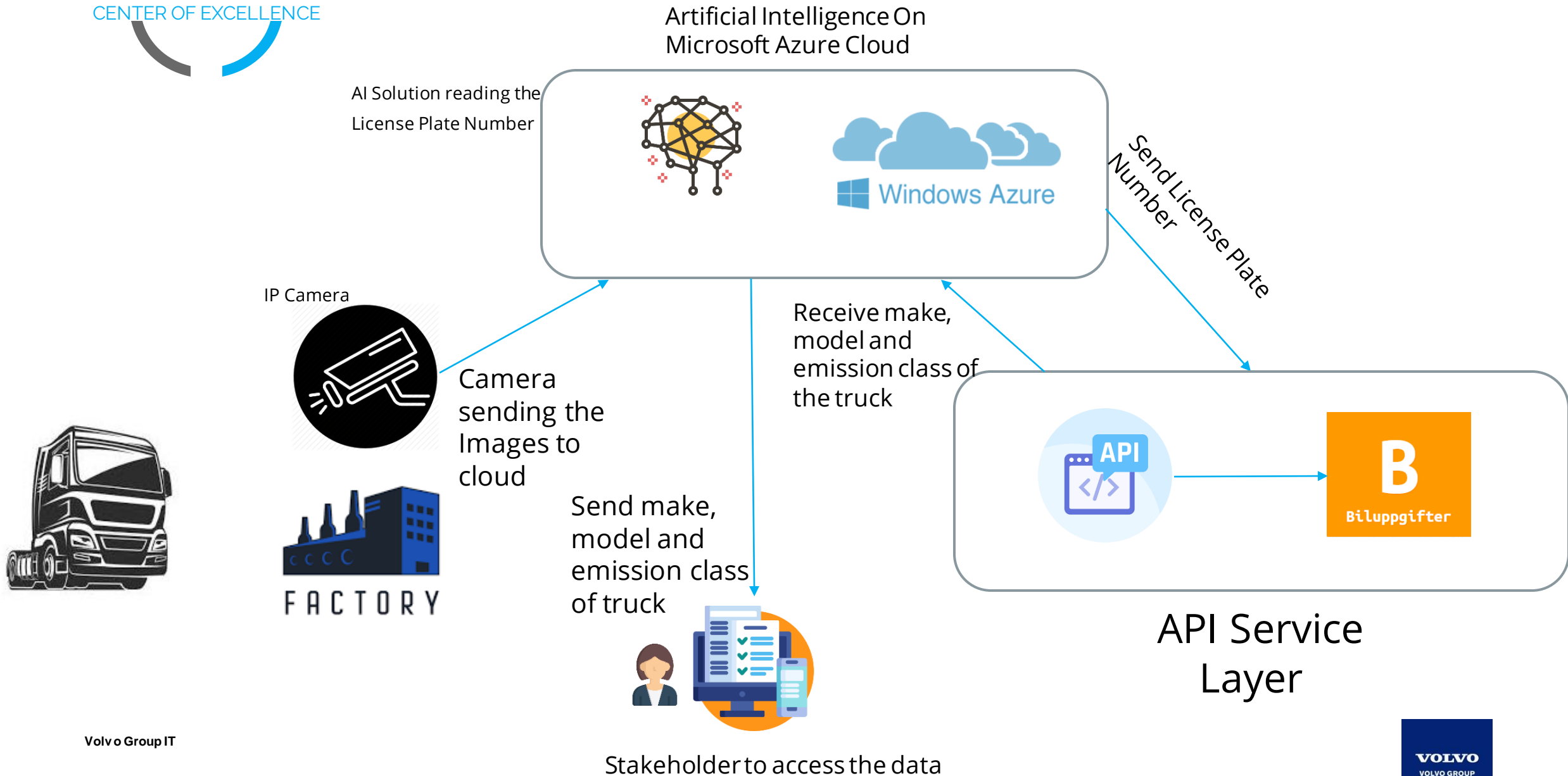
Manual entry
into Excel at LX
gate

Frequency:
Goods receipt capture:
Every day; 2 shifts
Sustainability report: 1
full day once per year

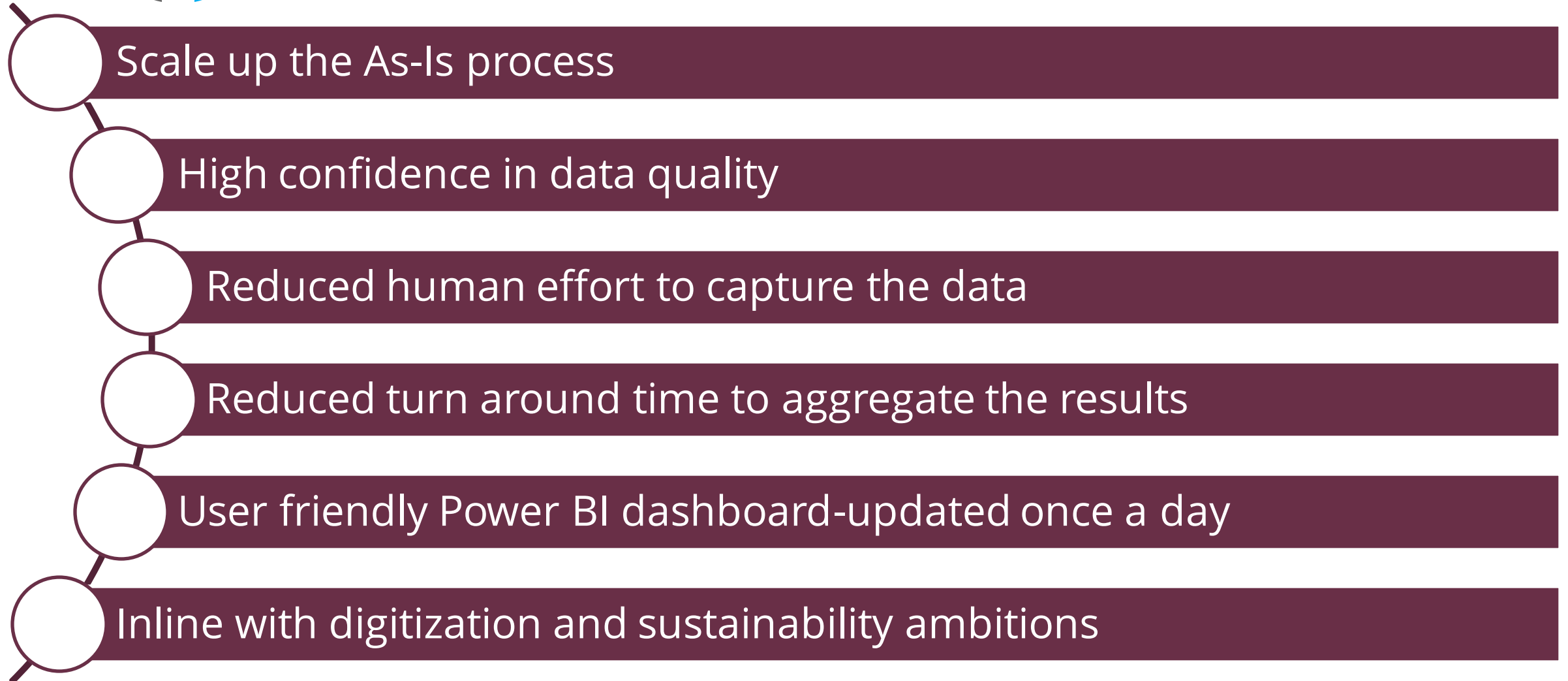


Stakeholder have
limited access to the
data

AI – Computer Vision Solution



BUSINESS VALUE





High Level Technical Flow

Volvo Facility Network

- Outdoor All weather - Casing IP66/IP67
- Resolution 1920x1080
- **Frame rate** Up to **25/30 FPS** (50/60 HZ) in all resolutions
- Power over Ethernet **POE**
- IR illumination Optimized IR (**Day & Night**)
- Operating conditions -40 °C to 60 °C (-40 °F to 140 °F)
- **Event Triggers Features** - Cross Line Detection, Video Motion Detection
- Lux - Color: 0.07 lux, at 50 IRE F1.2; **B/W** : **0.01 lux**, at 50 IRE F1.2
- Video compression - H.264
- Connectors Shielded RJ45

IP POE Camera: AXIS P1448-LE
LPR-Front at Tuve: LX Gate



Captures the images only
when the truck arrives at the
Gate

RJ45
Cable
Images are transmitted



Network
Switch

RJ45
Cable



Stand Alone Windows 10
Machine
Saving images for last three
days

Camera, Switch & Stand-
alone machine exist in the
same facility network area

Volvo Group IT

Microsoft Azure

Azure

Microsoft Azure
Blob Storage



Batch accounts



Tables

Tabular data storage



Azure Cosmos DB

Fetch the Truck details from
Biluppgifter

Copy batch
scheduled pushing
data to Azure



Power BI

Biluppgifter

B

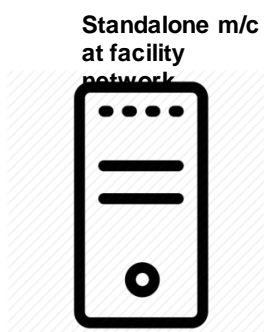
Biluppgifter

VOLVO
VOLVO GROUP

Solution pipeline - Azure



Volvo Facility Network

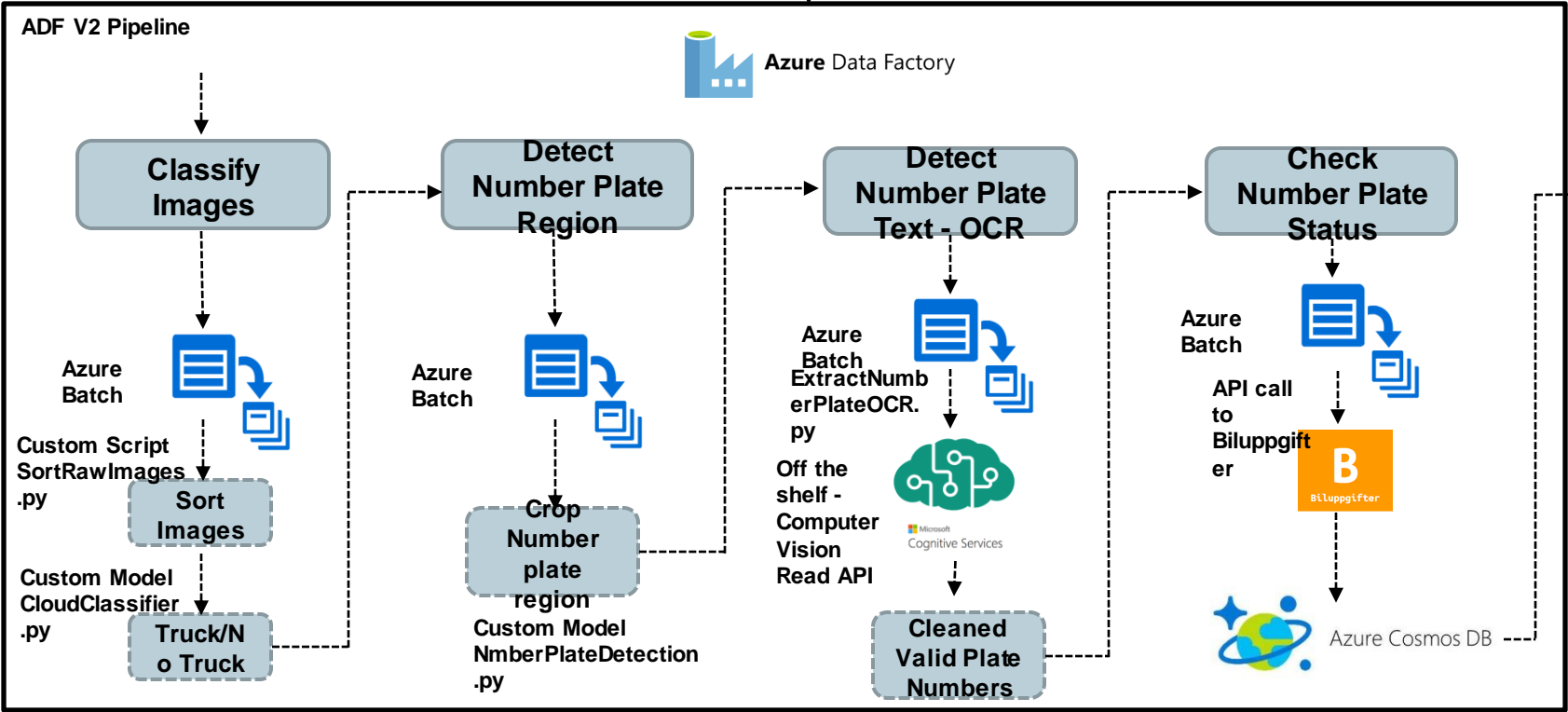


Standalone m/c at facility network

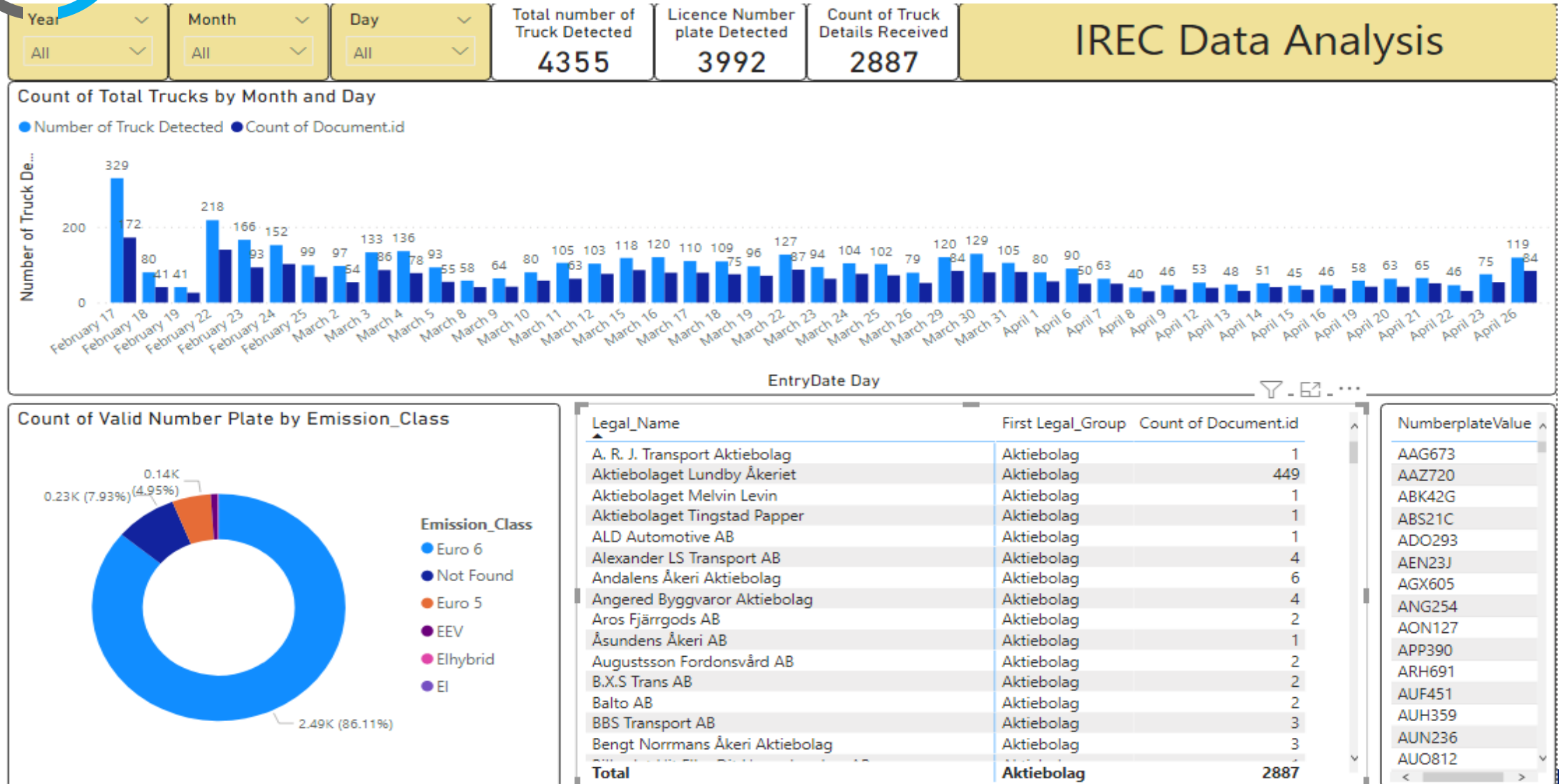
AZCOPY script moving files to Azure

Scheduled job running on standalone machine at factory

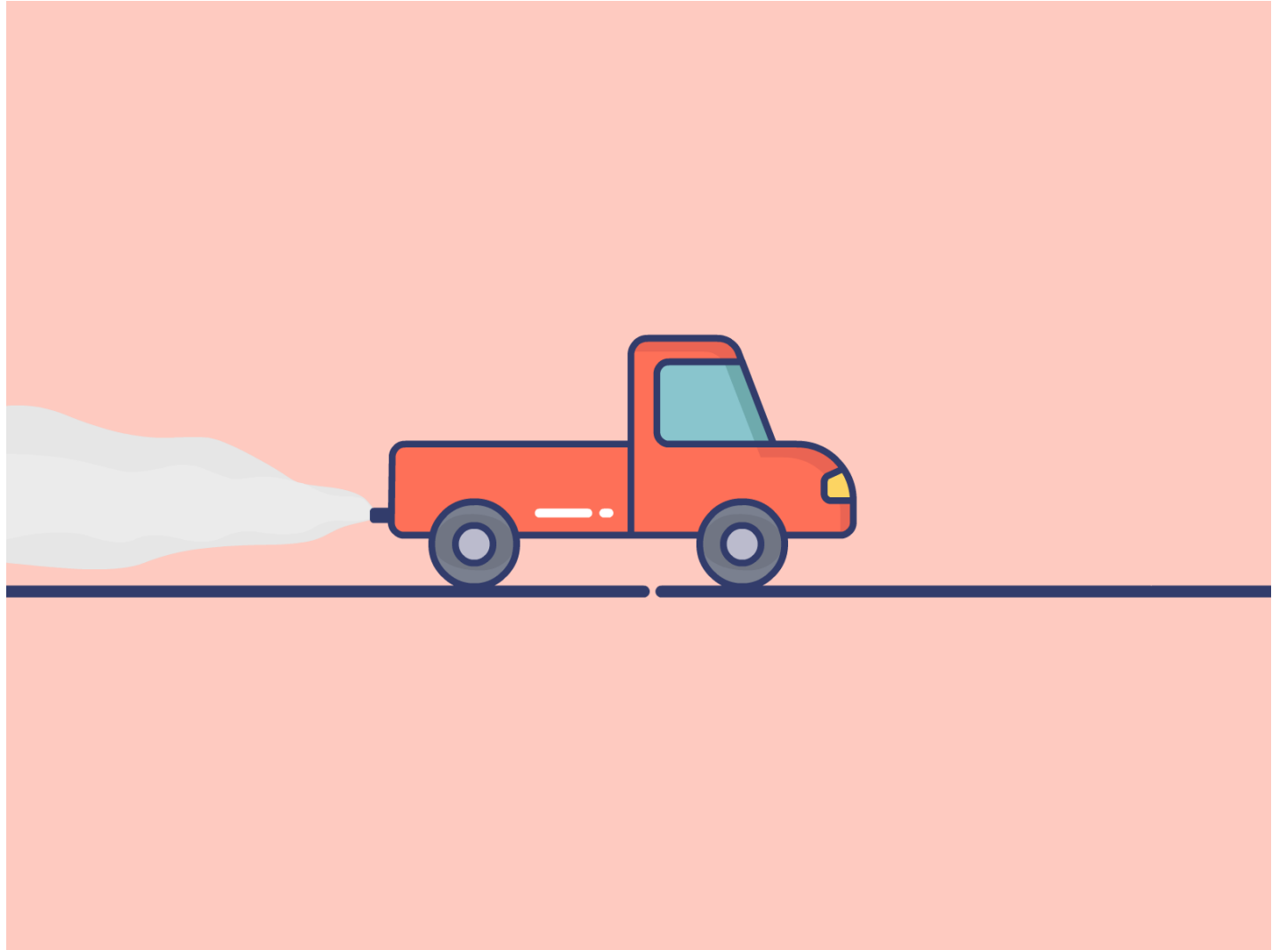
Microsoft Azure Environment Pipeline & Services



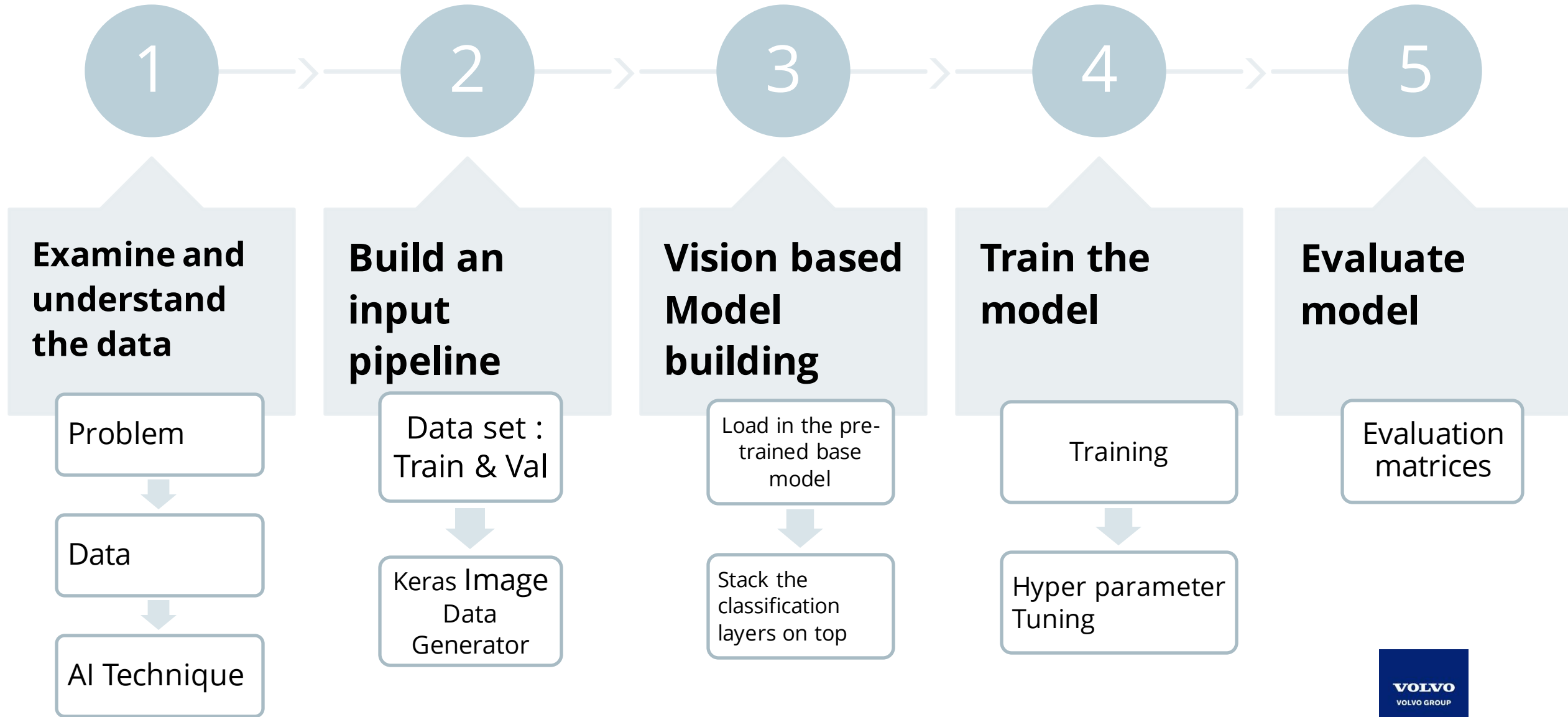
Solution output



Building the Truck/No Truck Model Using Transfer Learning

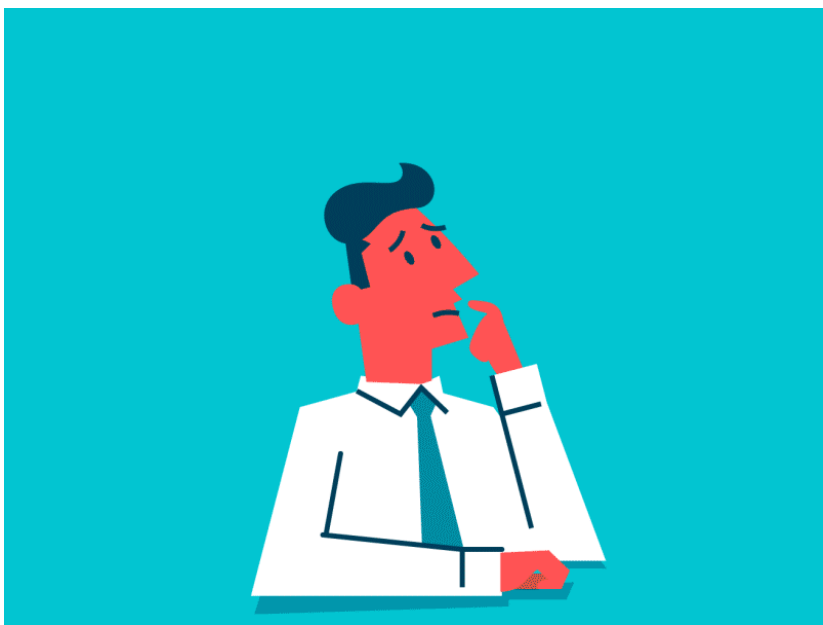


Machine learning workflow



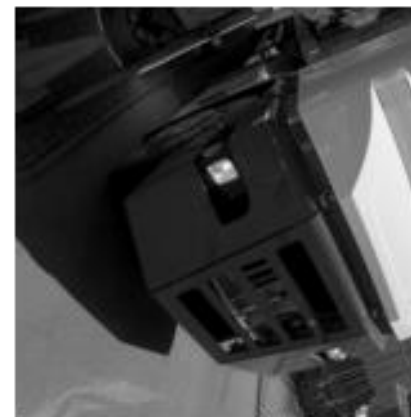
1.Examine and understand the data

- Understand the problem your solving
- Understand the Data
- Identify the right AI technique to solve



2. Build an input pipeline, in this case using Keras Image Data Generator

- The training Data set has **5622** files belonging to 2 classes.
- The Validation data set has **1228** files belonging to 2 classes.
- When you don't have a large image dataset, it's a good practice to **artificially introduce sample diversity** by applying random, yet realistic, transformations to the training images.



3. Building Vision Model

Create the base model from the pre-trained convnets

- We have built the base model from the Xception model developed at Google.

This is pre-trained on the ImageNet dataset, a large dataset consisting of 1.4M images and 1000 classes.

- ImageNet is a research training dataset with a wide variety of categories like Person, car, jackfruit etc.
- Transfer learning : [click](#)

```
model.summary()
```

```
Model: "model"
```

Layer (type)	Output Shape	Param #
=====		
input_2 (InputLayer)	[(None, 720, 720, 3)]	0

sequential (Sequential)	(None, 720, 720, 3)	0

tf.math.truediv (TFOpLambda)	(None, 720, 720, 3)	0

tf.math.subtract (TFOpLambda)	(None, 720, 720, 3)	0

xception (Functional)	(None, 23, 23, 2048)	20861480

global_average_pooling2d (Gl	(None, 2048)	0

dropout (Dropout)	(None, 2048)	0

dense (Dense)	(None, 1)	2049
=====		
Total params: 20,863,529		
Trainable params: 9,480,393		
Non-trainable params: 11,383,136		

4. Train & Hyper Parameter Tuning

- In the feature extraction experiment, you were only training a few layers on top of an Xception base model.
- The weights of the pre-trained network were not updated during training.
- One way to increase performance even further is to train (or "fine-tune") the weights of the top layers of the pre-trained model alongside the training of the classifier you added.
- The training process will force the weights to be tuned from generic feature maps to features associated specifically with the dataset.

After fine tuning the model nearly reaches 96% accuracy on the validation set.



5. Evaluation and prediction

- Finally, you can verify the performance of the model on new data using test set.

```
loss, accuracy = model.evaluate(test_dataset)
print('Test accuracy :', accuracy)
```

```
15/15 [=====] - 110s 7s/step - loss: 0.0860 - accuracy: 0.9583
Test accuracy : 0.9583333134651184
```



License Plate Detection Model using YOLO



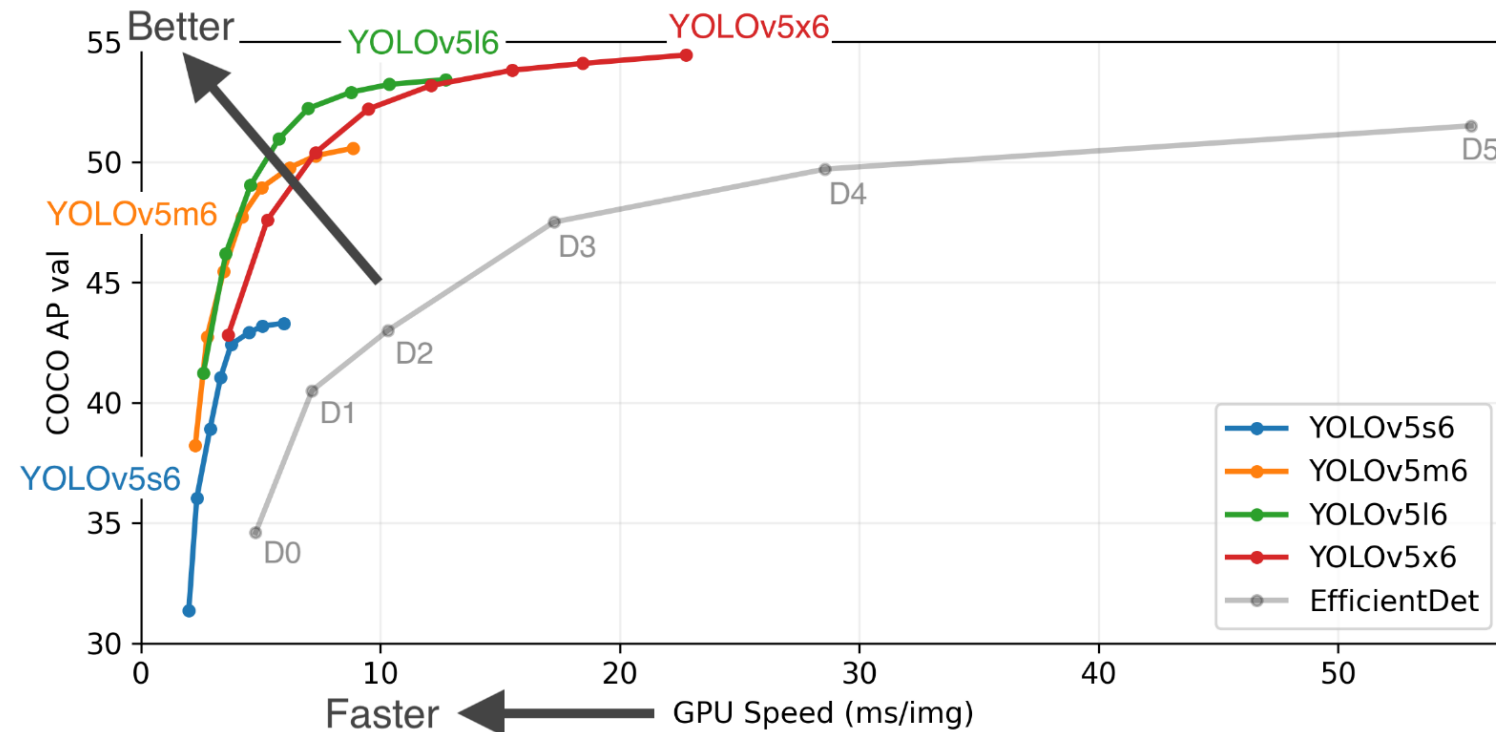
Computer Vision - Object Detection

When it comes to deep learning-based object detection, there are three primary object detectors you'll encounter:

- R-CNN and their variants, including the original R-CNN, Fast R-CNN, and Faster R-CNN
- Single Shot Detector (SSDs)
- YOLO

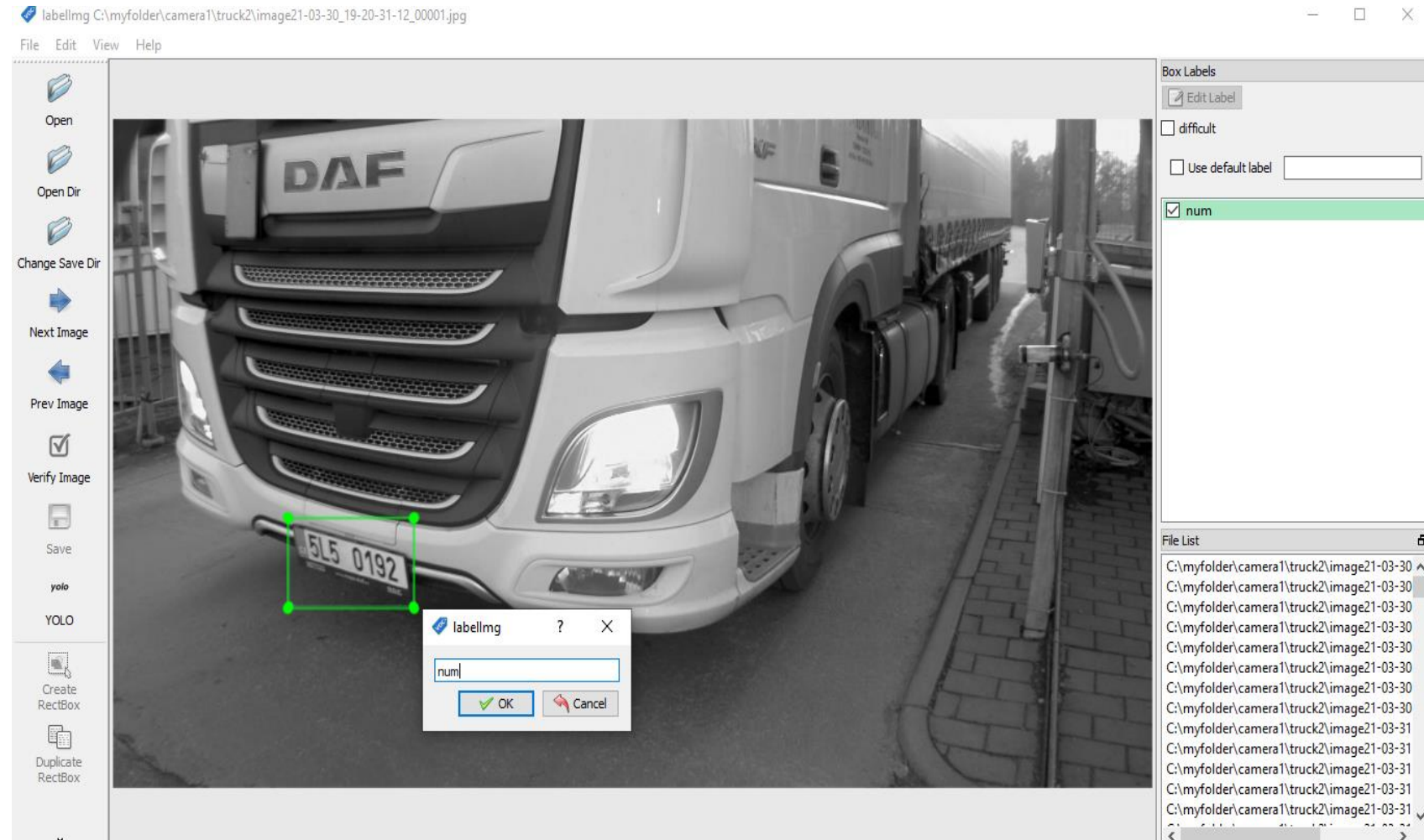
YOLO

- The biggest advantage of using YOLO is its superb speed – it's incredibly fast and can process 45 frames per second.



Data preparation and Image labelling

- Resizing, preprocessing the images, if necessary.
- Image labelling
 - After using a tool like [CVAT](#), [makesense.ai](#) or [Label box](#) to label your images
 - export your labels to **YOLO format**, with one *.txt file per image (if no objects in image, no *.txt file is required).
The *.txt file specifications are:



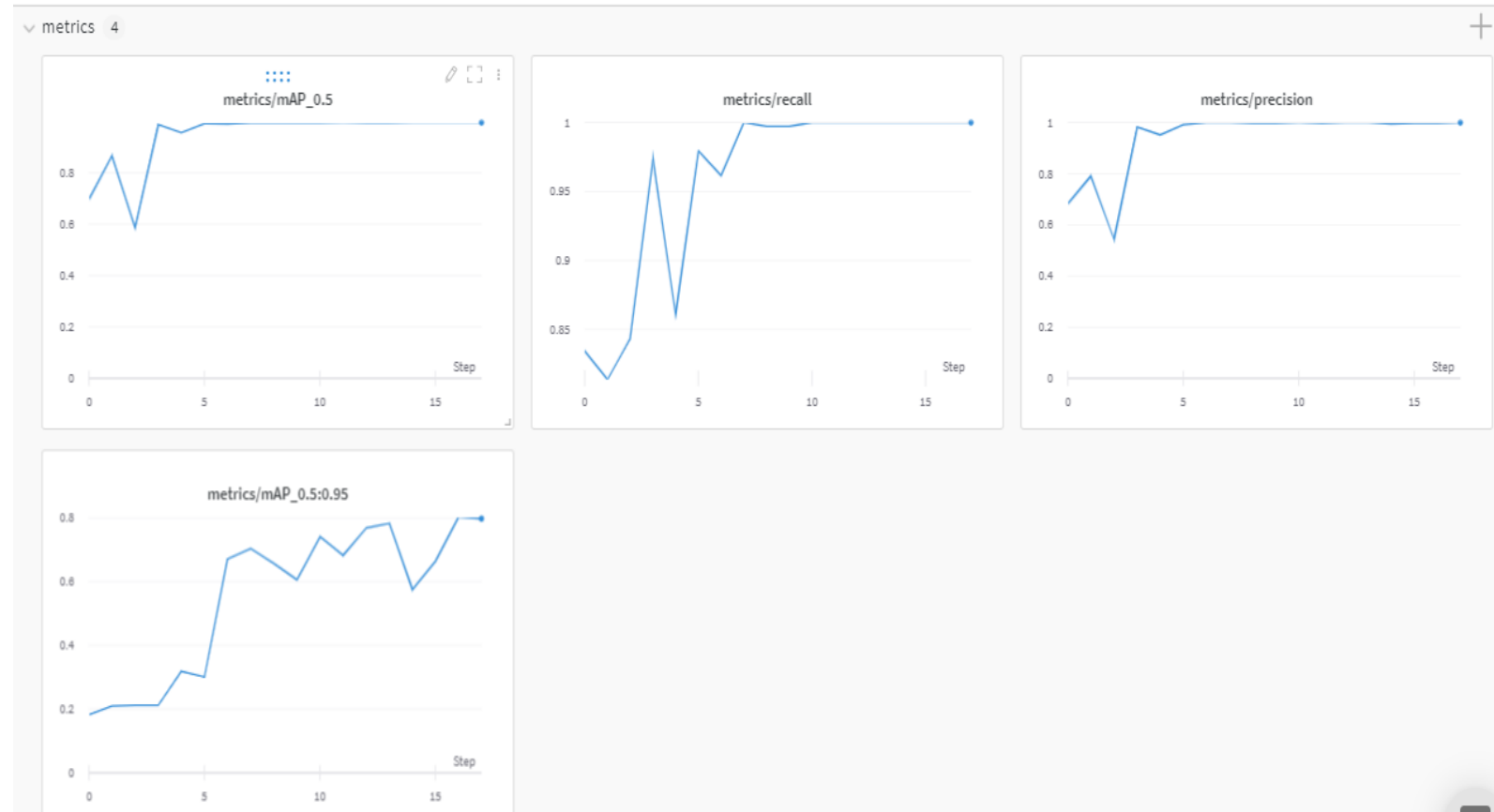
Model Training and Evaluation Metrics:

- The model was able to Achieve **99.67%** accuracy with the threshold of **0.50 IOU**, that is with at least 50% or above overlapping with ground truth of labelled images.
- Similarly, **80.28%** of accuracy with the threshold of **0.95 IOU**, that is with at least 95% or above overlapping with ground truth of labelled images.

```
# Train YOLOv5s on COCO128 for 5 epochs$
```

```
python train.py --img 640 --batch 16 --epochs 25 --data data.yaml --weights yolov5s.pt
```

Mean average precision (mAP), recall, precision where the evaluation matrix considered for the case.

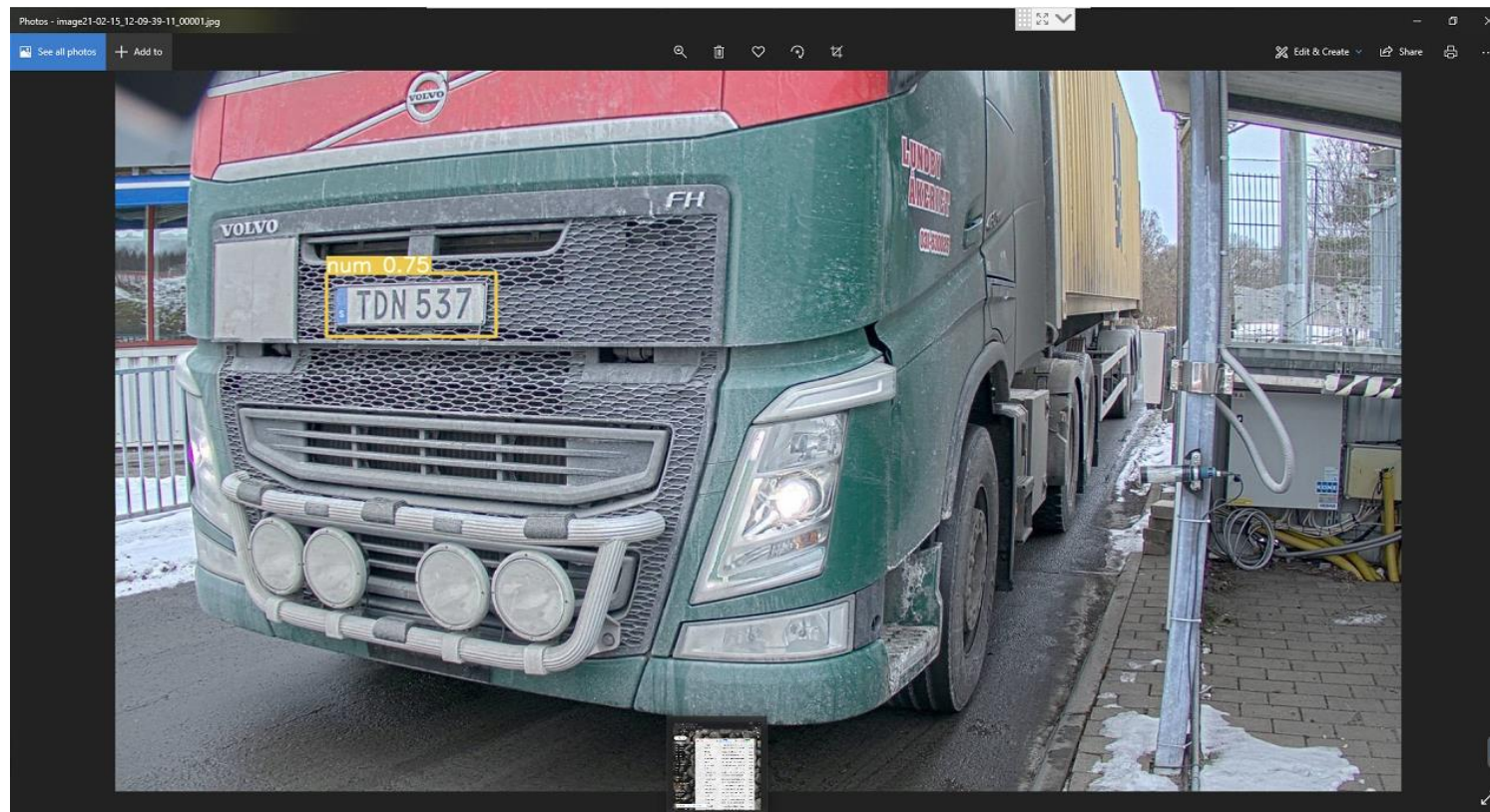


Output

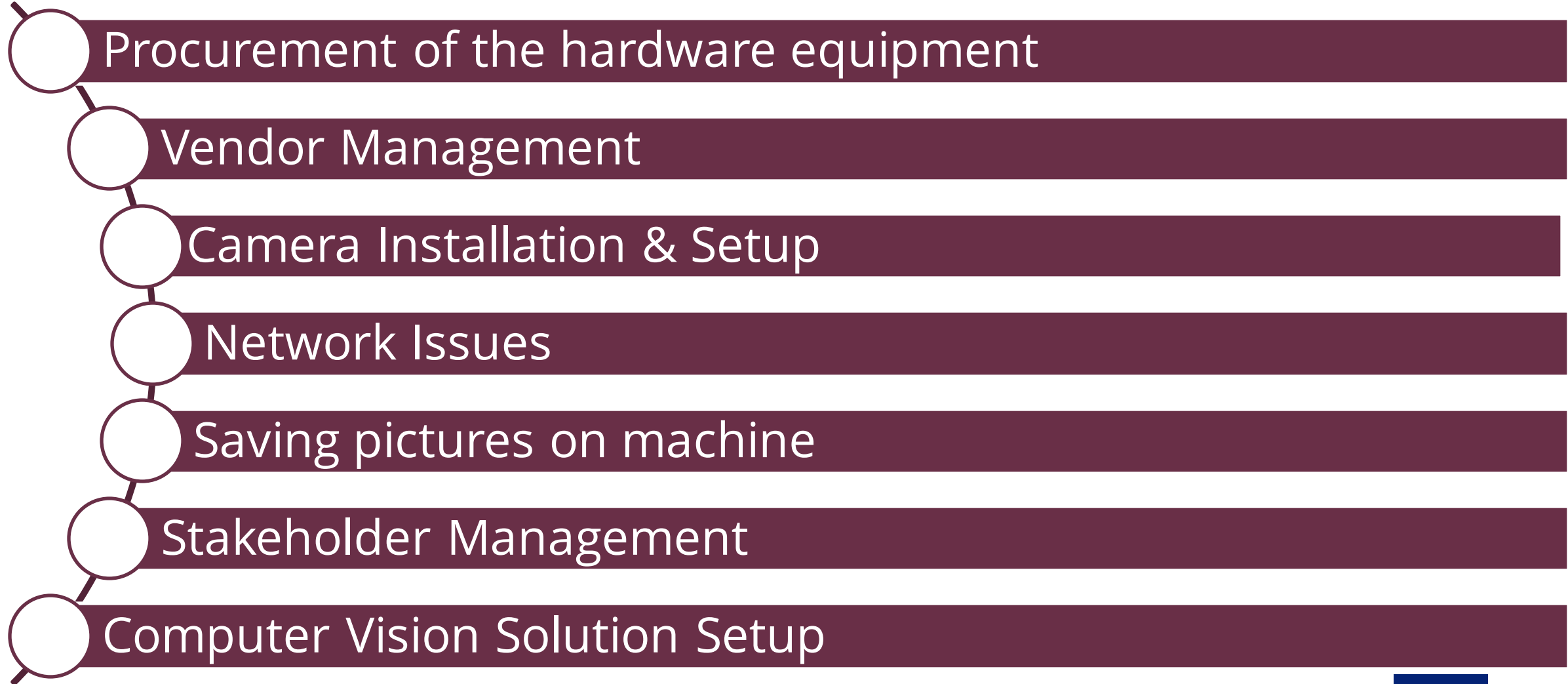
- Bounding box with confidence value is displayed on sample test image
- Other Different input Sources

```
0 # webcam
file.jpg # image
file.mp4 # video
path/ # directory
path/*.jpg # glob
'https://youtu.be/NUsoVIDFqZg' # YouTube video
'rtsp://example.com/media.mp4' # RTSP, RTMP, HTTP
stream
```

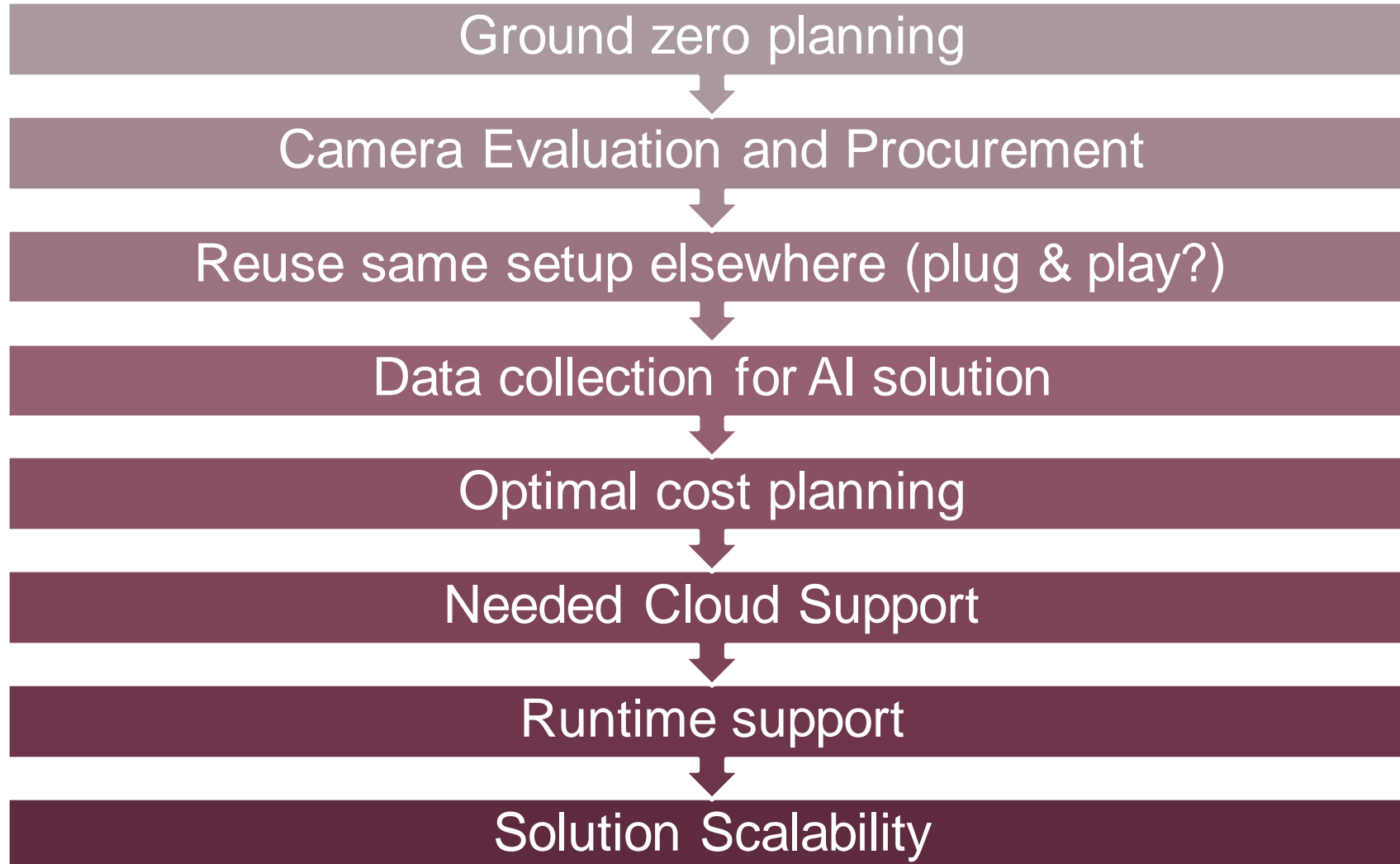
- A custom Function build to crop bounding box.
- This cropped image will be further used by Read API to extract text from the image.



LESSONS LEARNT



Recommendations



QUESTION & ANSWER TIME



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Useful Links

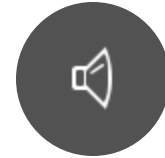
Here you have the links for the content we have shared during the presentation. Click on the icon to access the site.



AI Enterprise Guidelines



[AI & ML Community Portal](#)



Emerging Technologies and Integration



[AI & ML Yammer Group](#)

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