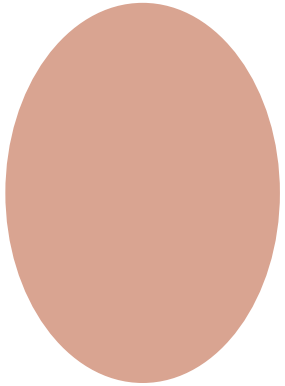
The background of the slide is a photograph of a dark-colored car driving on a wet asphalt track. The car is moving away from the viewer, and its rear is slightly blurred, suggesting motion. The wet surface of the track reflects the car and the surrounding environment. In the background, there are green trees and a clear sky. A white rectangular box is overlaid on the right side of the image, containing the title and subtitle text.

AI-based augmentation of camera data

KI-basierte Augmentierung von Kameradaten

Team Members



KEVAL



AMAN SURESH RUPARELIYA



VIKAS ANIL TIWARI



JOSHI VARDHAN LELLA



JONAS RAPP



SABRINA RIEGER



LENNART MASSARCZYK



FELIX NEUSINGER



ASHOKA GOLITHADKA
UDAYASHANKARA

Clear



Target

The goal of this project is to implement spray on good weather images with the help of artificial intelligence

Wet





Definition of spray

In the context of roads, the term "spray" usually refers to the swirling water created by traffic when it rains. Particularly in heavy rain and at high speeds, the water can splash on the roadway, creating spray that can obstruct drivers' vision and make the roadway slippery.

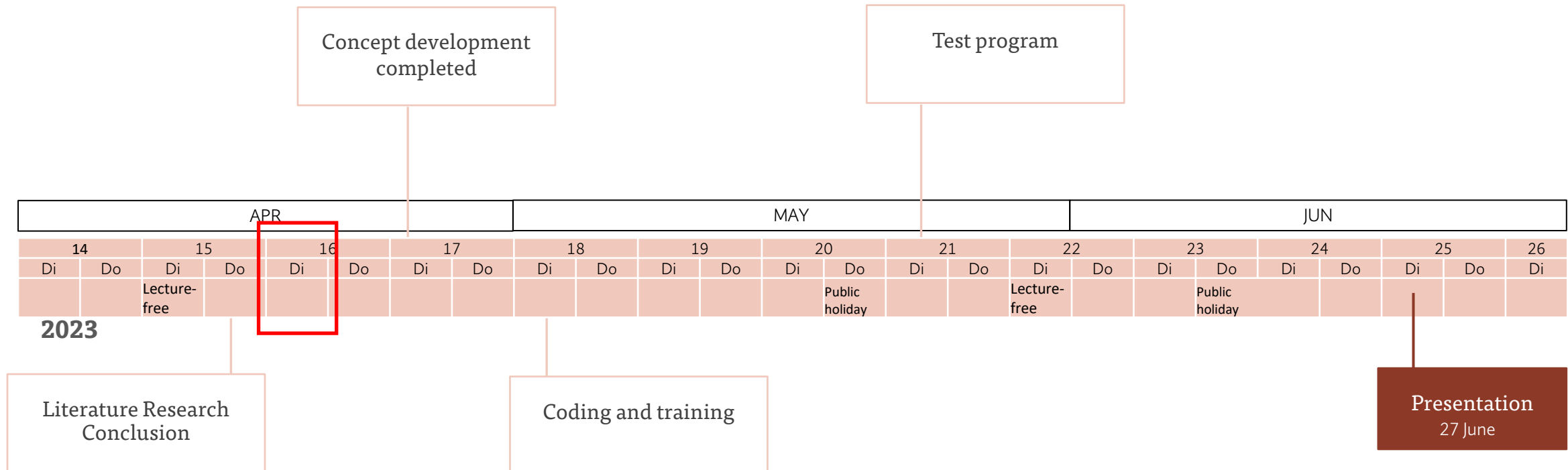
General information

Meeting with project leader and presentation of results	Tuesday 11:35 - 13:05 Room C203
Group discussion and division of work packages	Thursday 18:30 - 20:00 Discord / in person (Room tbd)

Tools

Organization	Miro	https://miro.com/app/board/uXjVMYoGMsl=/
Data exchange	GitHub	https://github.com/thi-spray-augmentation/ss23
Programming	Python	
Online Meeting	Discord	https://discord.com/channels/1087746349476360262/1087746349476360265
Input data	Moodle	https://moodle.thi.de/course/view.php?id=8930

Milestone plan



Agenda 18.04.

1. Present approaches
2. Questions
3. Define / decide concept
4. Define Next Meeting

Approaches

Deep
Convoluti-
onal GAN

Lennart & Felix

Neural Style
Transfer
(modified VGG-
19) &
Day2Dusk
(GAN -> UNIT)

Jonas & Sabrina

Image
processing of
Segmentation
masks

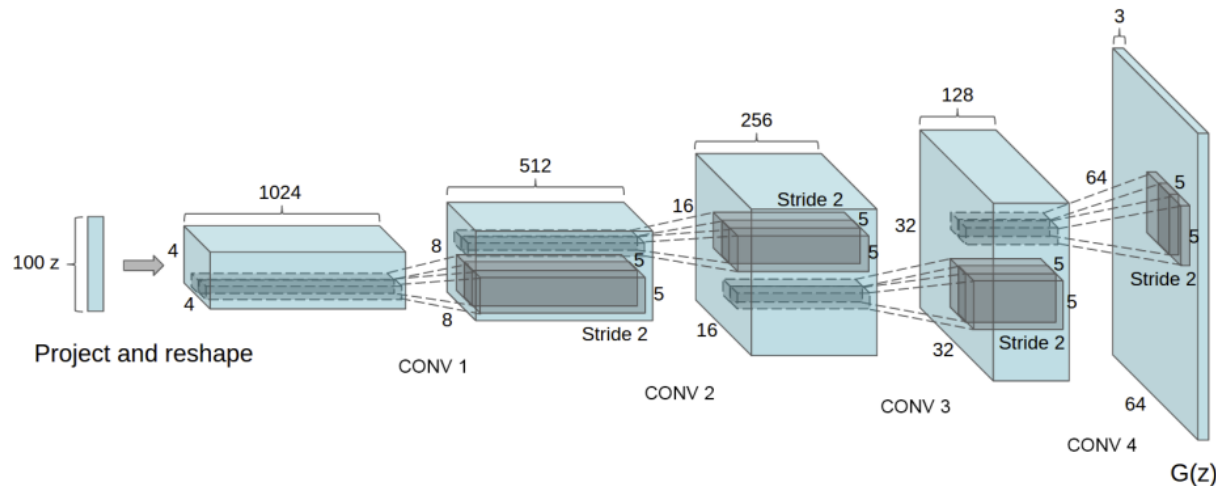
Aman & Keval

Python &
Opencv to
introduce
Spray

Ashoka G U

1. Deep Convolutional GAN

A DCGAN is a direct extension of the GAN described above, except that it explicitly uses convolutional and convolutional-transpose layers in the discriminator and generator, respectively



Problems

- Figuring out how to feed good weather pics to learned network
- A lot of different approaches with different depth of documentation on the use
- Difficult to fully understand

Advantages

- Is capable of learning nearly every weather condition
- Not really complicated to set up a simple network
- Realistic looking pictures

How much training input is needed?

- Min. 300 bad weather pictures

Own assessment

- We just scratched the surface with this approach, but dcGAN has a lot of potential

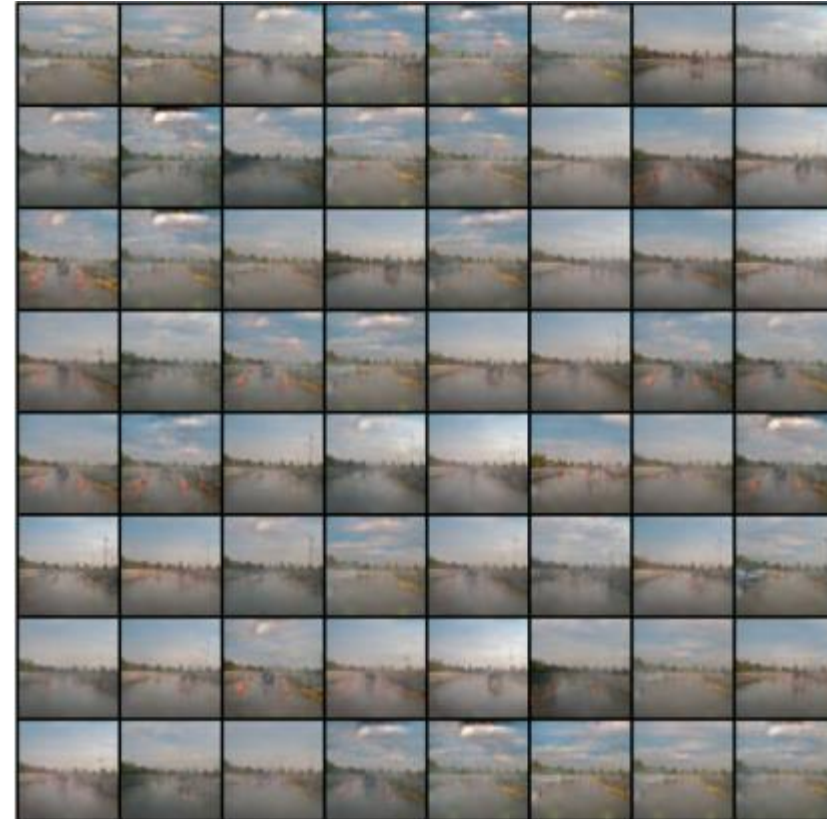
-> good to use further

1. Deep Convolutional GAN- Results

Before



After



2. Neural Style Transfer (modified VGG-19) & Day2Dusk (GAN -> UNIT)

Testing of existing DNN for image augmentation and manipulation given by MathWorks. Try to adapt these towards our needs and get a deeper insight into structure and basic settings regarding training of DNN.



Problems

- Preprocessing of data (expected input size of DNN) was partly not correct
- Extremely long training using whole dataset for training

Advantages

- Is capable of learning nearly every weather condition
- Not really complicated to set up a simple network
- Realistic looking pictures

How much training input is needed?

Could not be identified yet, but using our dataset + basic augmentation should deliver a hopefully okay result already

Own assessment

- If applying correctly should be able to generate the desired results

2. Neural Style Transfer (modified VGG-19) & Day2Dusk (GAN -> UNIT) - Results

Before



After

Transfer Image After Iteration
20



2. Neural Style Transfer (modified VGG-19) & Day2Dusk (GAN -> UNIT) - Results

Before



After

Transfer Image After Iteration
20

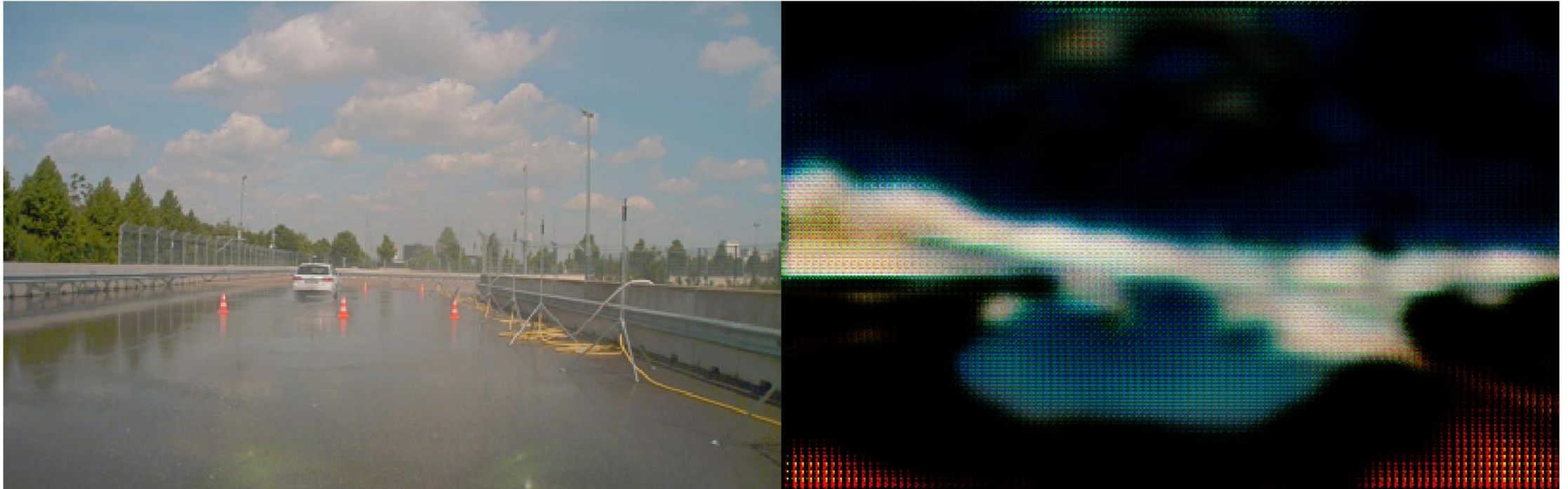


2. Neural Style Transfer (modified VGG-19) & Day2Dusk (GAN -> UNIT) - Results

Before:

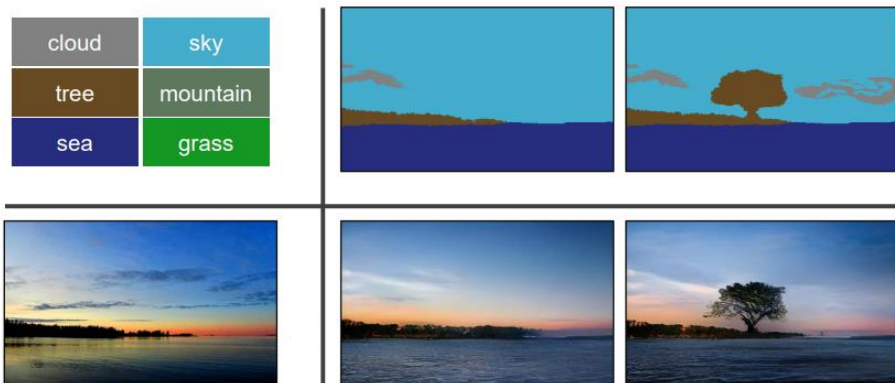
After:

Test Wet Image 1 with Translated Clear Image



3. Image processing of Segmentation masks

This approach involves creating segmentation model which creates masks of noise(spray) from the wet image dataset. Then masks of cars from the dry image dataset. Then we apply various image blending techniques to get the desired results.



Problems

- We need to generate masks for all images of both the datasets.
- The 'Segment Anything Model (SAM)' which was released by Meta could only create masks for cars.
- The quality of the output is unpredictable at the moment.

How much training input is needed?

- The images in our dataset are enough.
- But binary masks are required for all images.

Advantages

- This approach is comparatively less complicated to apply.
- The masks can also be used for supervised learning in GAN.

Own assessment

- I am not sure about the quality of the result this approach can produce working alone. It can be paired with other approaches to boost the output.

3. Segmentation masks - Results

Dry



Wet



4. Python & Opencv to introduce Spray

OpenCV (Open Source Computer Vision) is a free, open-source computer vision and machine learning library that allows developers to perform a wide range of image and video processing tasks.

OpenCV is used in a variety of applications, such as object detection, image processing, facial recognition, and robotics. It provides a large set of pre-built functions and algorithms for image and video analysis, including basic image processing functions like filtering and edge detection, as well as advanced algorithms like feature detection and machine learning

Problems

- Detecting the exact location
- Spray pattern and Intensities
- Adding noise and blur depending on the weather condition
- Need to do trial and error

Advantages

- Not complex
- Specific
- No need of CNN

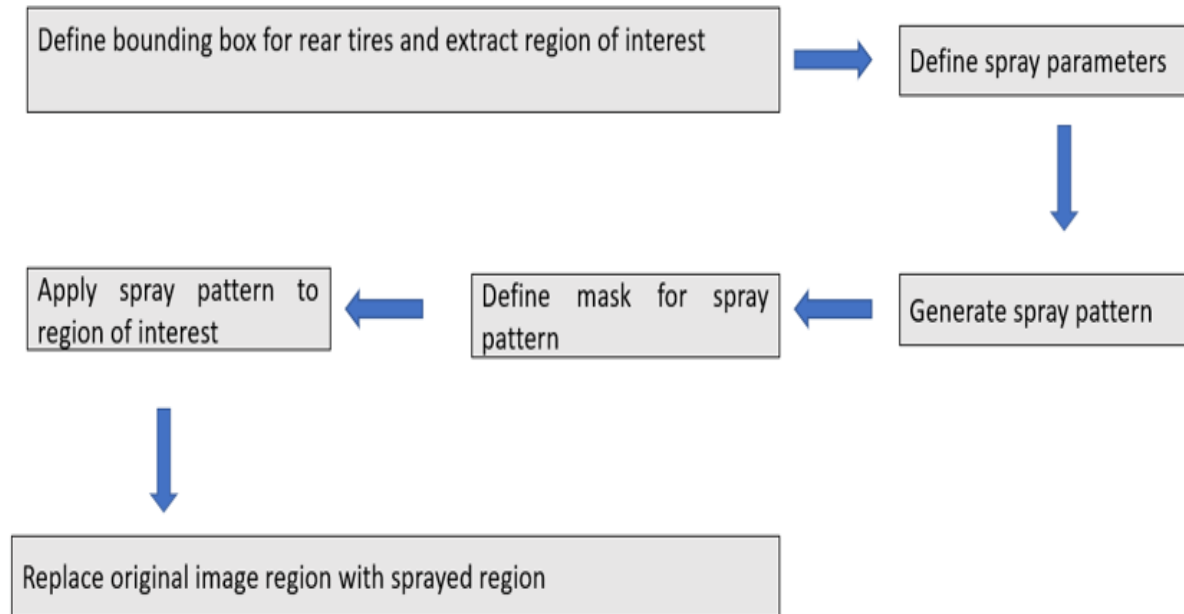
Own assessment

How much training input is needed?

- Not able to judge as of now, need to try many patterns and intensities

- To try with different spray patterns and intensities
- To change the brightness and check (Conditions)
- Can we use YOLO to detect water spray and water content present in the road and then introduce it to a dry image?

4. Python & Opencv to introduce Spray - Steps



1. Read the image
2. Convert the image to grayscale
3. Apply Gaussian blur to the grayscale image
4. Apply Canny edge detection to the blurred image
5. Find contours in the edge detected image
6. Create a mask of zeros with same dimensions as original image
7. Draw contours on the mask
8. Apply bitwise operation on original image and mask to get only the part of original image where water spray is present.

4. Python & Opencv to introduce Spray - Steps

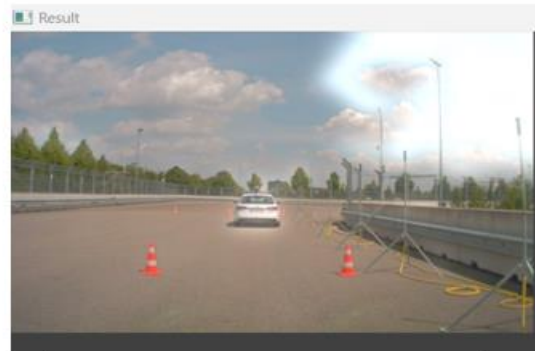
Trail 1



Dry road without spray



Wet road with spray



Output

4. Python & Opencv to introduce Spray - Steps

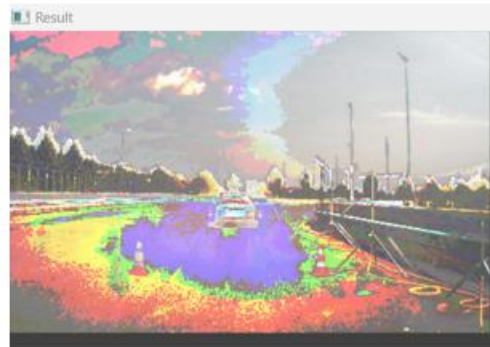
Trail 2



Dry road without spray



Wet road with spray



Output

4. Python & Opencv to introduce Spray - Steps

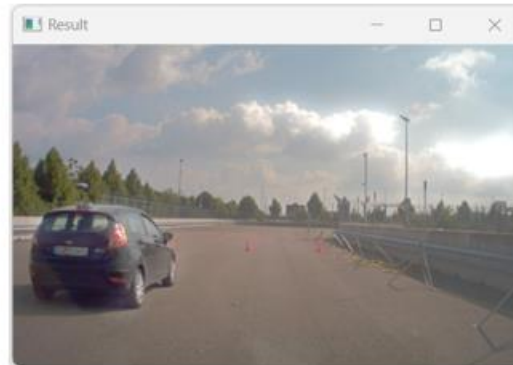
Trail 3



Dry road without spray



Wet road with spray



Output

Questions

1. Yolo – algorithm (need the tires)
2. Definition of the target?
3. Model spray? – intensive?
4. What is our data used for?
5. Day / Night also needed?