

# Renault Group



## Construction of a Surrogate Model: Multivariate Time Series Prediction with a Hybrid Model

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Driving Simulation & Virtual Reality Conference & Exhibition

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# **Summary**

## **01 Introduction**

BACKGROUND, ISSUES AND OBJECTIVES

## **02 Data description**

INPUT PARAMETERS AND OUTPUT TIME SERIES

## **03 Surrogate model construction**

CLASSICAL METHODS, AGGREGATION AND *HYBRIDATION*

## **04 Conclusion**

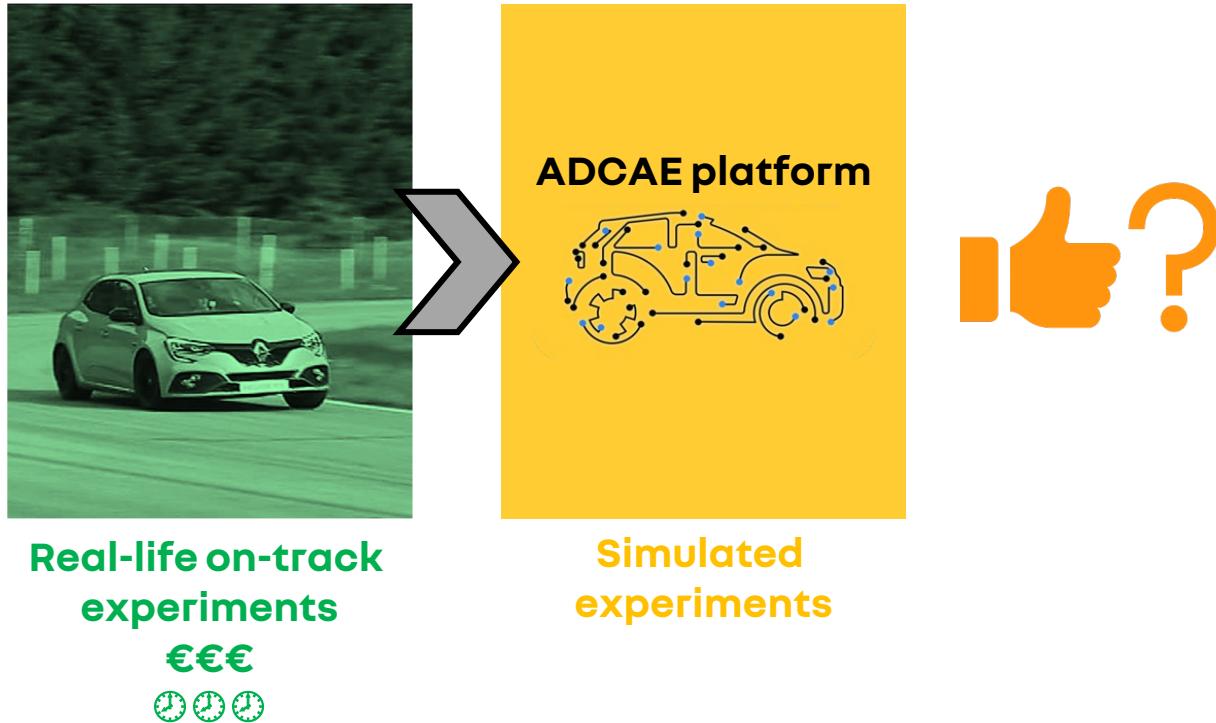
NUMERICAL RESULTS AND COMPUTATION TIMES

# 01

## Introduction

BACKGROUND, ISSUES AND OBJECTIVES





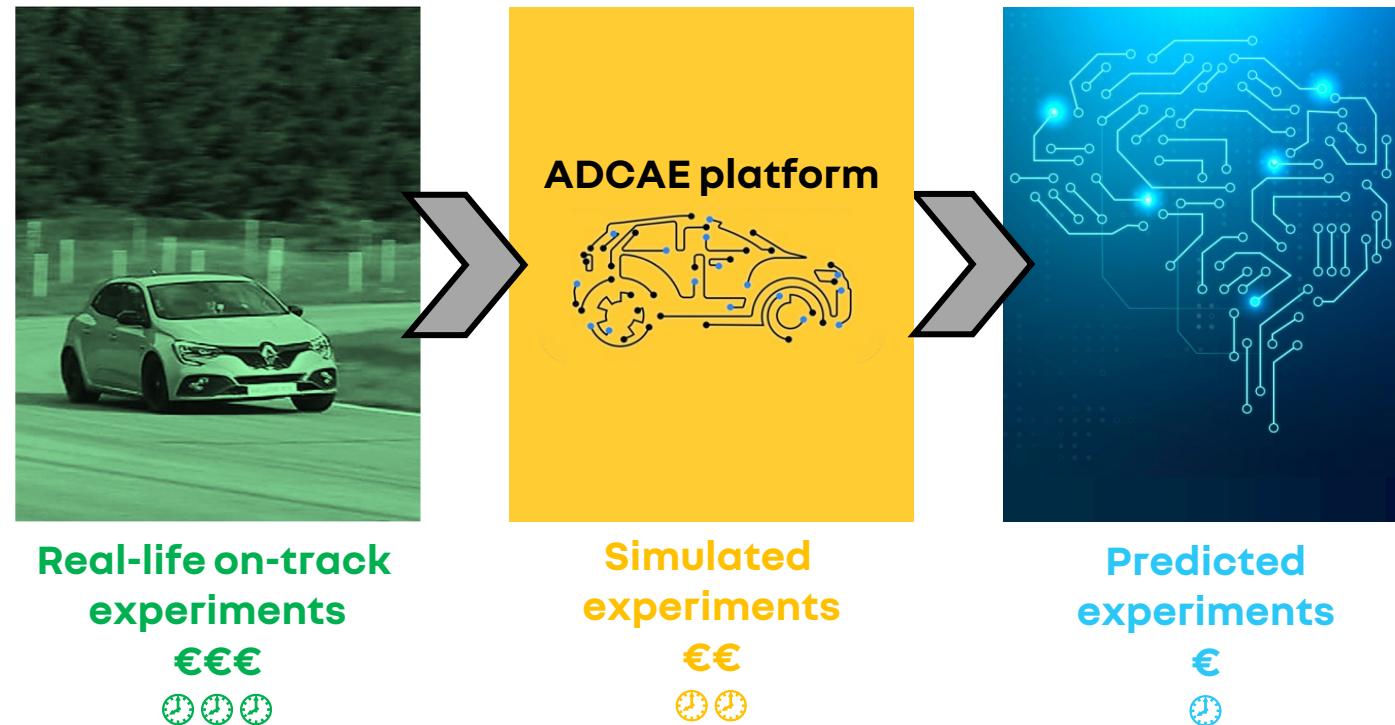
## CONTEXT: Validation and certification of autonomous vehicles

- Real-life on-track experiments are costly and time-consuming
- Use the simulator through simulated experiments to integrate them into the vehicle certification and homologation process

*Are digital simulations sufficiently correlated with reality to be used legally?*

## GLOBAL OBJECTIVE: Simulator calibration

- Develop a methodology for gauging the quality of simulations and adapting their use, in order to prove that it is possible to supplement or even replace track tests with simulator tests



## PROBLEM:

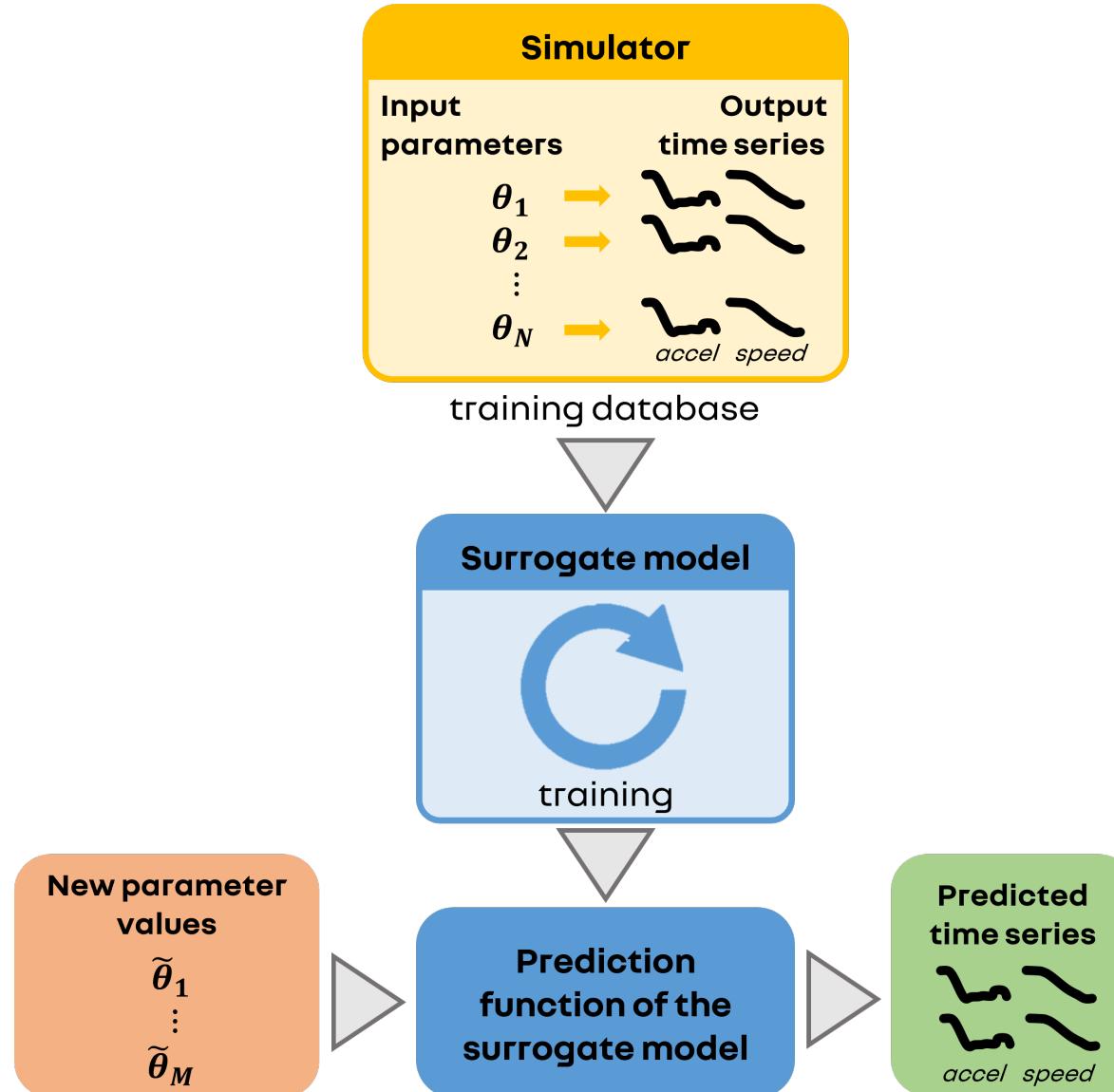
The simulator takes a long time to launch (several minutes for each launch), which is restrictive in certain contexts:

- Relaunch the simulator repeatedly in an iterative process
- Massive simulation (~1M)

## OBJECTIVE:

- Build a surrogate model based on Machine Learning methods that will mimic the simulator
- A model that is as accurate as possible while keeping computation times to a minimum

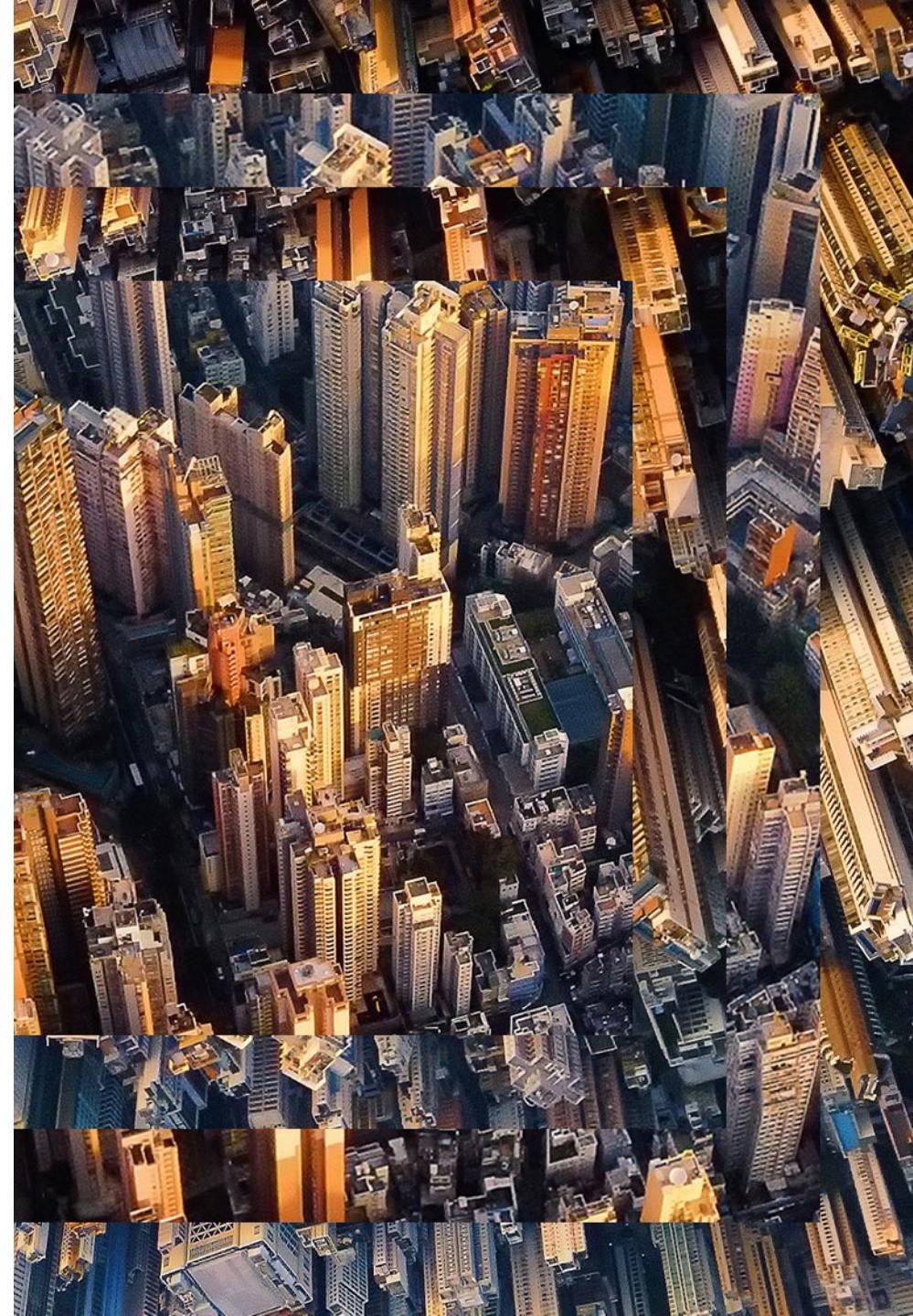
# 01 - Introduction: background, issues and objectives



# 02

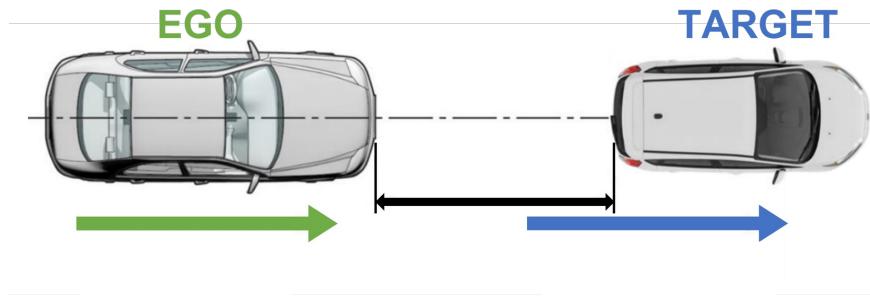
## Data description

INPUT PARAMETERS AND OUTPUT TIME SERIES



## 02 - Data description: input parameters and output time series

### CONSIDERED SCENARIO:



Two vehicles in motion: testing the AEB of the second vehicle (**EGO**) following sudden braking by the first (**TARGET**).

### DATA FORMAT:

- Input: 7-number vector
- Output: 2684-number vector
- 1642 experiments in total

### CONSIDERED LOSS:

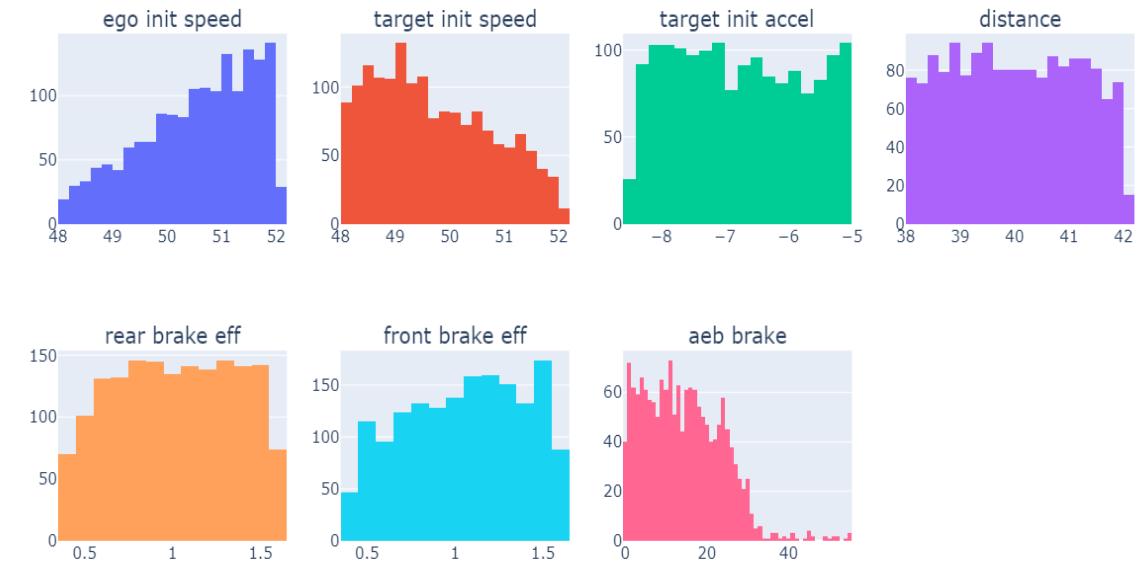
$$\text{RMSE}(\text{true value, prediction}) = \sqrt{\frac{1}{n} \sum_{i=1}^n (\text{TR}_i - P_i)^2}$$

**TRAIN**  
1442

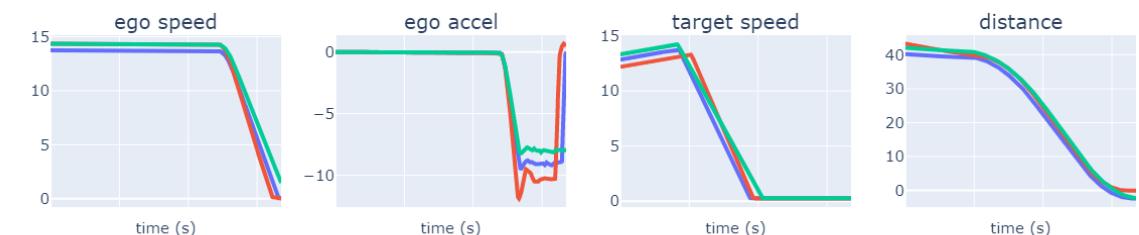
**VALIDATION**  
100

**TEST**  
100

### INPUT PARAMETERS : 7 values



### OUTPUT PARAMETERS : 4 time series, 671-time steps



# 03

## Surrogate model construction

CLASSICAL METHODS, AGGREGATION AND HYBRIDATION



## RMSE ( $\times 10^{-2}$ ) between true and predicted values

- All proposed methods :

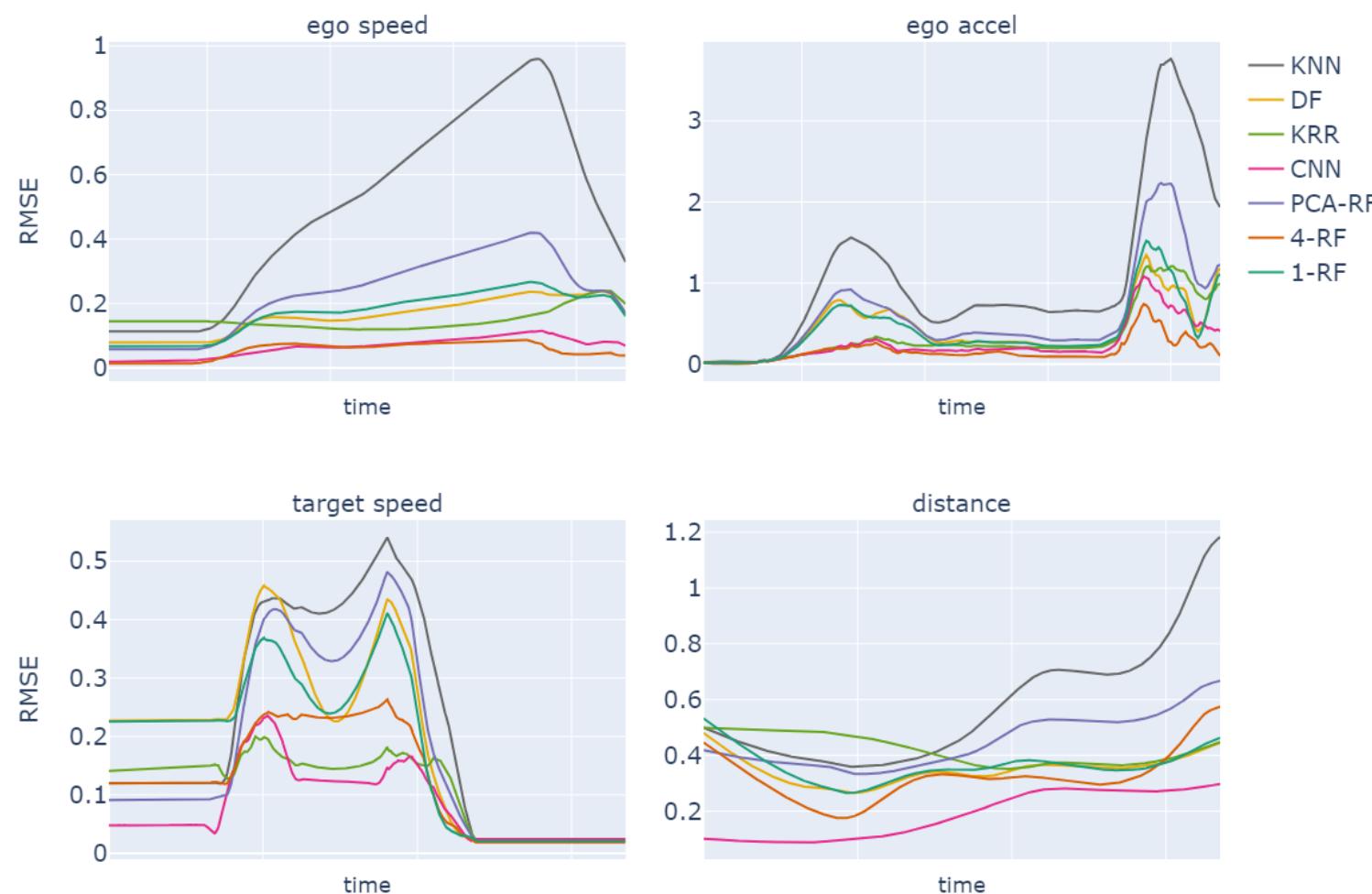
		k-NN	KRR	CNN	DF	1-RF	4-RF	PCA-RF	
RMSE ( $\times 10^{-2}$ )	train	9.23	0.04	1.22	4.26	1.24	0.71	1.95	TRAIN
	validation	30.15	7.21	2.31	7.05	7.27	3.69	12.35	VALIDATION
Training time		0.05 sec	0.22 sec	1h	13 min	42 sec	53 sec	7.42 sec	
Prediction time		0.01 sec	0.02 sec	1.74 sec	0.90 sec	0.09 sec	1.00 sec	1.24 sec	

- More detailed results :

	Ego speed (m/s)	Ego accel (m/s <sup>2</sup> )	Target speed (m/s)	Distance (m)	VALIDATION
<b>k-NN</b>	11.00	64.77	8.02	36.83	
<b>KRR</b>	2.20	7.14	1.63	17.86	
<b>CNN</b>	0.18	3.85	<b>1.06</b>	<b>4.16</b>	
<b>DF</b>	1.32	8.88	5.80	12.19	
<b>1-RF</b>	1.36	9.54	4.86	13.31	
<b>4-RF</b>	<b>0.12</b>	<b>1.47</b>	2.34	10.84	
<b>PCA-RF</b>	2.33	20.53	5.58	20.96	

k-NN :  $k$  nearest neighbors ; KRR : Kernel Ridge Regression ; CNN : Convolutional Neural Networks ; DF : Deep Forest ; RF : Random Forests ; PCA-RF : Random Forests with PCA

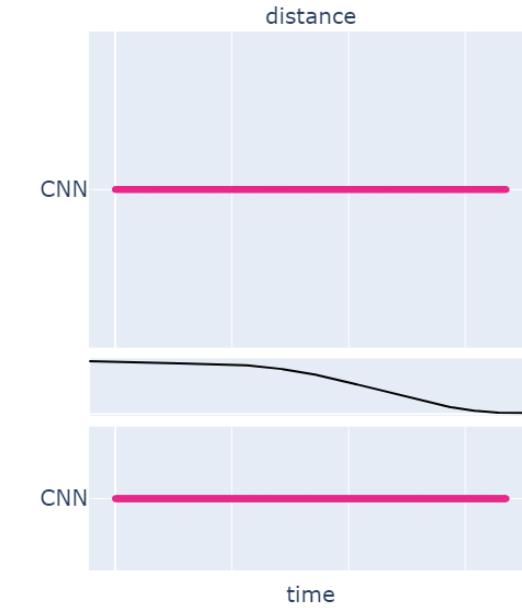
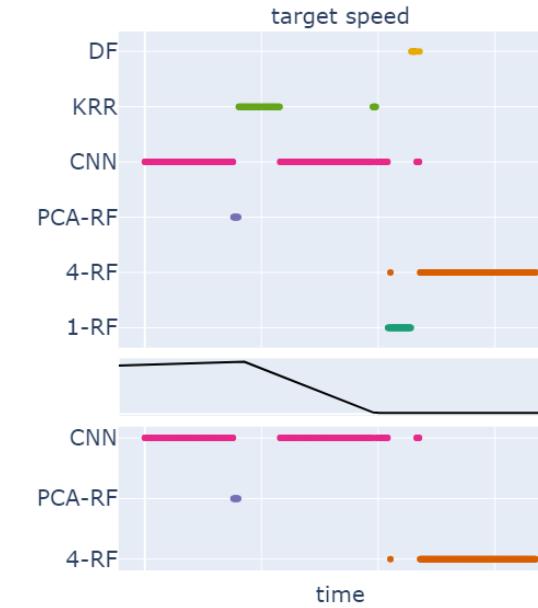
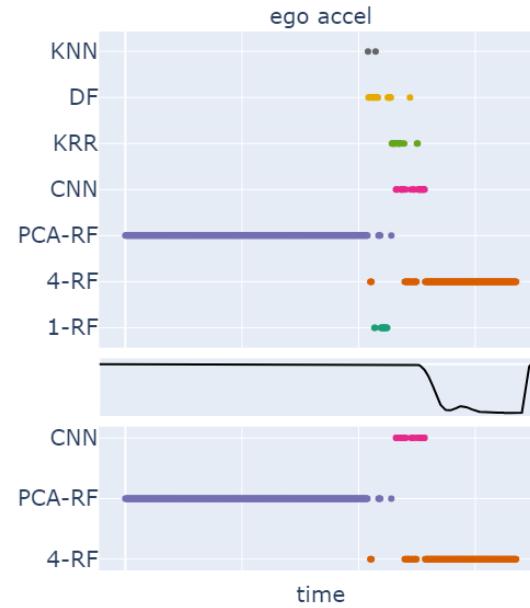
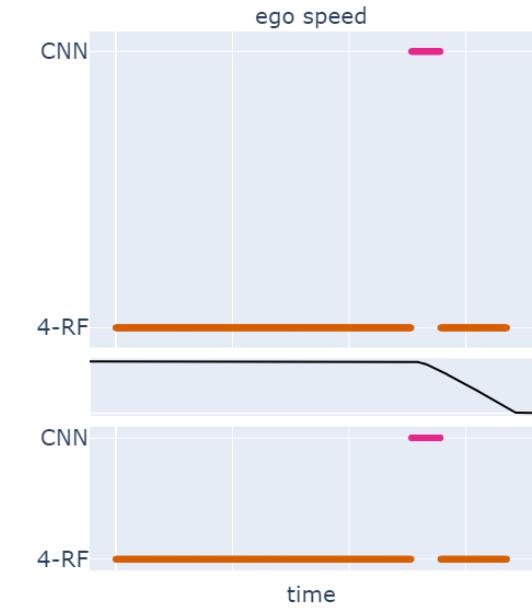
## RMSE ( $\times 10^{-2}$ ) between true and predicted values at each time step



## First idea: hybrid models

VALIDATION

One method selected for each time step



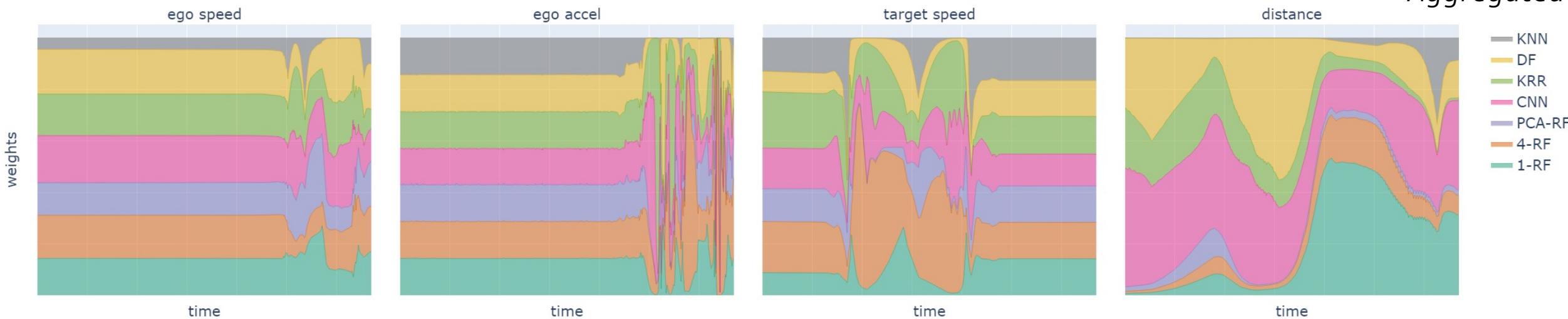
Hybrid 1

Hybrid 2

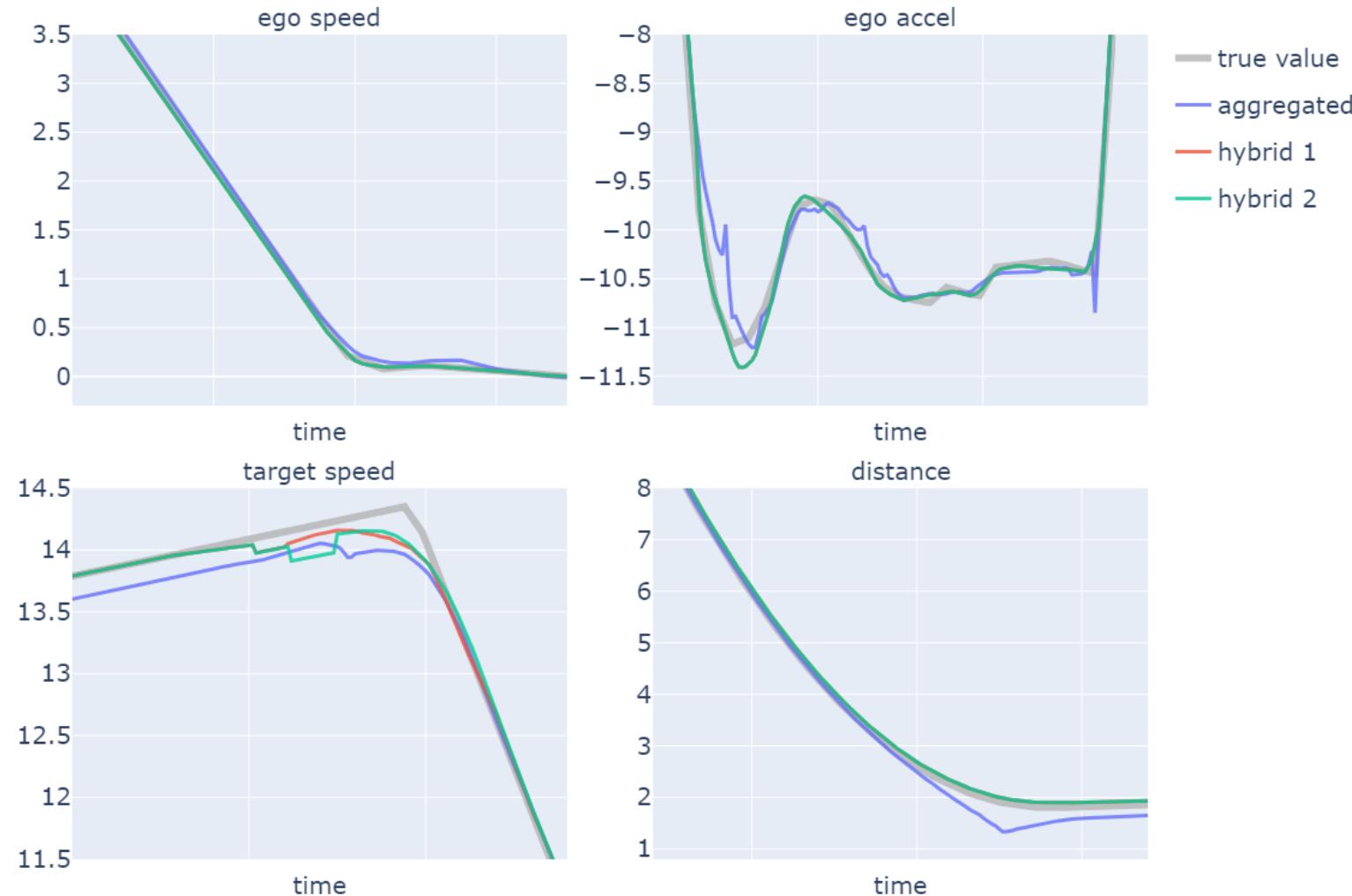
## Second idea: expert aggregation

VALIDATION

Retain all methods by assigning them higher or lower weights at each time step



## Final results: one concrete example



## Final results: RMSE ( $\times 10^{-2}$ ) between true and predicted values

- With validation set (data used to construct the new models):

VALIDATION

	Ego speed	Ego accel	Target speed	Distance	Mean
CNN	0.18	3.85	1.06	4.16	2.31
4-RF	0.12	1.47	2.34	10.84	3.69
Hybrid 1	0.11	1.46	0.93	4.16	1.66
Hybrid 2	0.11	1.46	1.04	4.16	1.69
Aggregated	<b>0.07</b>	<b>0.59</b>	<b>0.24</b>	<b>1.36</b>	<b>0.56</b>

- With test set (never-used data):

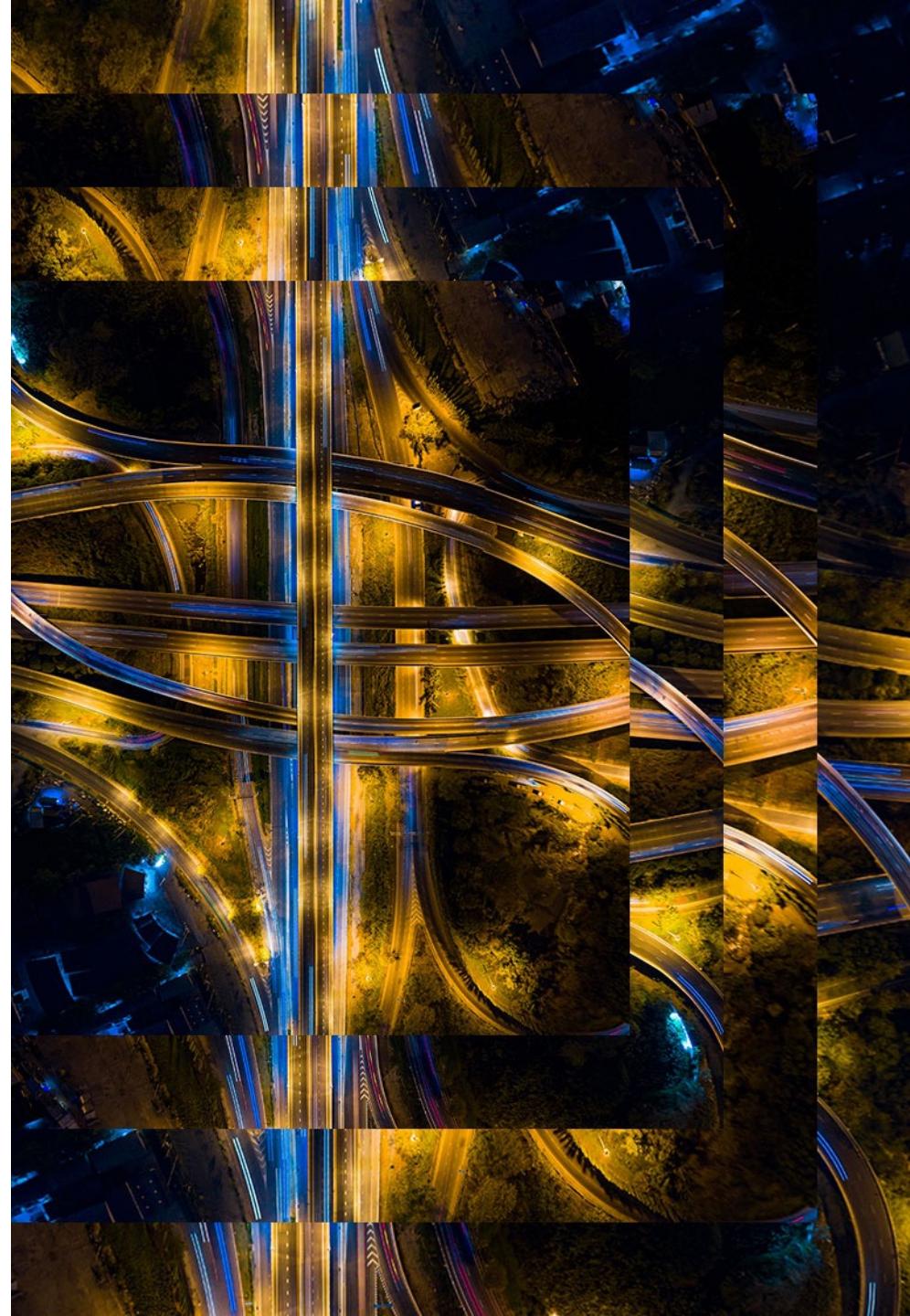
TEST

	Ego speed	Ego accel	Target speed	Distance	Mean
CNN	0.23	3.34	1.00	2.52	1.77
4-RF	0.13	1.36	2.36	9.84	3.42
Hybrid 1	0.12	1.35	1.12	2.52	1.28
Hybrid 2	0.12	1.35	<b>1.00</b>	2.52	<b>1.25</b>
Aggregated	0.64	3.81	1.35	4.97	2.69

# 04

## Conclusion

NUMERICAL RESULTS AND COMPUTATION TIMES



- **Calculation times:**

	CNN	4-RF	Hybrid 1(+)	Hybrid 2 (+)	Aggregated (+)
Training	59 min	53 sec	0.29 sec	0.18 sec	2 min 13
Prediction	1.39 sec	0.15 sec	10.25 sec	8.59 sec	2 min 17

(+) add the prediction times for each method

- **To generate 50,000 experiments in several stages:**

4-RF	Hybrid 2	Simulator *
1 minute	1 hour	5 days

\*stop and restart the simulator at each step

- **To conclude :**

- Mixing models gives better results.
- The whole process increases calculation times but is still faster than using the simulator.

Choice to do: trade-off between calculation time and accuracy

A nighttime photograph of a winding road through a dense forest. The road curves through the center of the frame, with its surface appearing bright due to the long exposure. Light trails from moving vehicles are visible as streaks of red and white along the curves. The surrounding trees are dark silhouettes against a deep blue night sky. The overall atmosphere is serene and dynamic.

Thank you