

# aimsun.

### Breaking new ground in how we manage large transport networks: data analytics, AI and real-time simulation prediction

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## Multi-resolution modelling



Static models Long term planning Demand modelling State/Region-wide models

#### Dynamic models

City-wide models (meso) Medium-long term planning

Driver behaviour model (micro) Short term planning Operation analysis Unexpected events Environmental analysis Congestion charging ITS, AV, CAV



## From Aimsun Next to Aimsun Live

#### Offline

- Micro-meso-macro integration
- Large-scale static and dynamic models
- Faster than real time
- Planning/operational model



#### Real-time

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- Monitoring
- Prediction for next 60-120min
- Decision support

## Monitoring

• From a limited set of detectors to a full network coverage

Traffic Monitoring in Real Time Quality indicators Daily reports Monthly reports



Real Time Monitoring t = 8:00:00



## Prediction





## **Decision support**



#### Assessment of different response plans



Indicators and choice of the best strategy

#### Proactive traffic management





# Research to improve Aimsun Live processes

Aimsun Live launches one simulation every 5min

- What demand to go with that simulation
- What routes will the vehicle choose
- What are the optimum network parameters

#### **Research objectives**

- Partially automatize model calibration tasks
- Improve model performance
- Faster processes





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## 1-027: Advanced data analytics for real-time demand calibration and prediction in large scale networks



Activity 1: Develop a systematic framework to establish a library of traffic states and OD matrices from real traffic data



Activity 2: Develop short term traffic demand matching model for real time traffic simulation of urban networks



Activity 3: Enrich simulation model by offline calibration of route choice models using real traffic observations



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Activity 4: Integration, testing, and further refinement of the developed modules on TMR node



1-025: Advanced data analytics for real-time demand calibration and prediction in large scale networks



Activity 1: Develop proxy Aimsun Live model for offline development



Activity 2: Develop the Al-assisted calibration module based on data-driven machine learning techniques for analysing disparities between Aimsun Live's model results and observed data



Activity 3: Develop the pattern refinement module that maps traffic patterns with associated optimal parameter set for Aimsun Live



Activity 4: Develop the prediction confidence module



Activity 5: Integration, testing and further refinement of the developed modules





## Test bed for the research

- Aimsun has developed two Aimsun Live test-beds
- Analysed/cleaned the historical data
- Calibrated the base models
- Deployed Aimsun Live pilot for the clients
- Transferred all the historical data, model and related information to the respective universities







## 1-025 Activity 2: Develop the Al-assisted calibration module

5.0

4.5

50

75

100

maxDesiredSpeed\_Truck



0.8 Cost egula Inf 3.5 3.5 3.0 3.0 3.0 0.0 0.0 0.0 1.5 1.5 2.0 0.5 1.0 2.0 2.5 1.0 2.5 3.0 0.5 1.0 1.5 reactionTime\_Global reactionTimeTrafficLight\_Global jamDensity\_Global 5.0 5.0 5.0 2.0 2.0 4.5 4.5 45 surrogate model Cost 4.0 observations expected best × 3.5 regularisation 3.0 3.0 3.0 0.0 0.0 0.8 1.0 1.2 1.4 1.6 0.8 1.0 1.2 1.4 1.6 50 75 100 125 150 speedLimitAcceptance Car speedLimitAcceptance Truck maxDesiredSpeed Car 5.0 4.5 9.6 Cost 3 5 3.0

5.0

4.5

4.0

20

2 0

Simulation start time: 09:00:00. Min of expected: 3.273.

2.0

0.0

125 150









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## Conclusion

- The research methodologies will be implemented in the Aimsun software/processes
- This should reduce manual model calibration effort
- The model performance for both offline and live models should improve
- These projects shows how we can combine AI, Data analytics and Simulations together to get the best outcome for short term traffic predictions



Thank you!

