

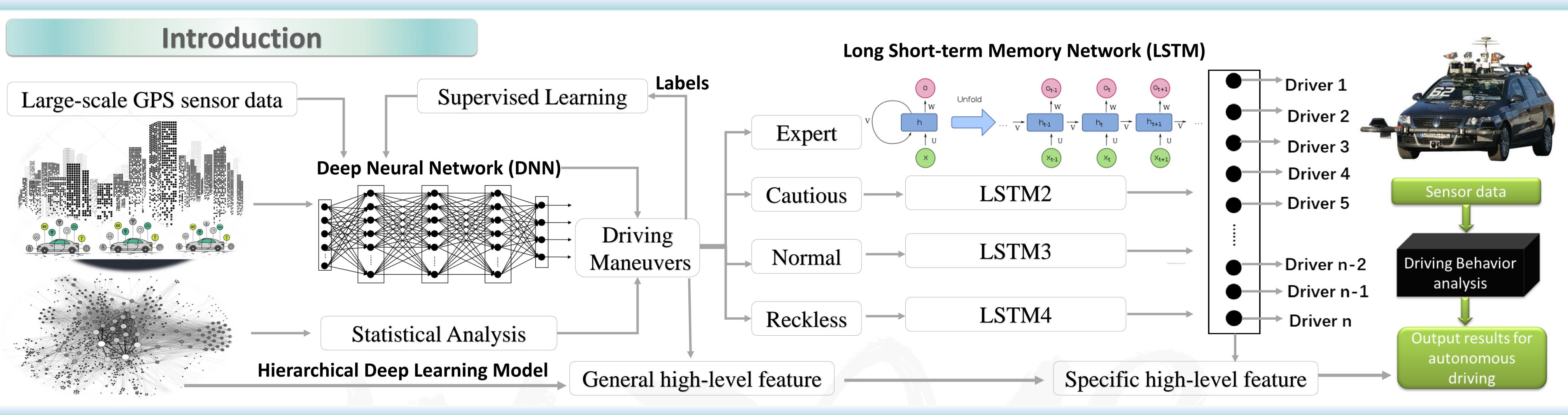
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Joint Deep Neural Network Modelling and Statistical Analysis on Characterizing Driving Behaviors

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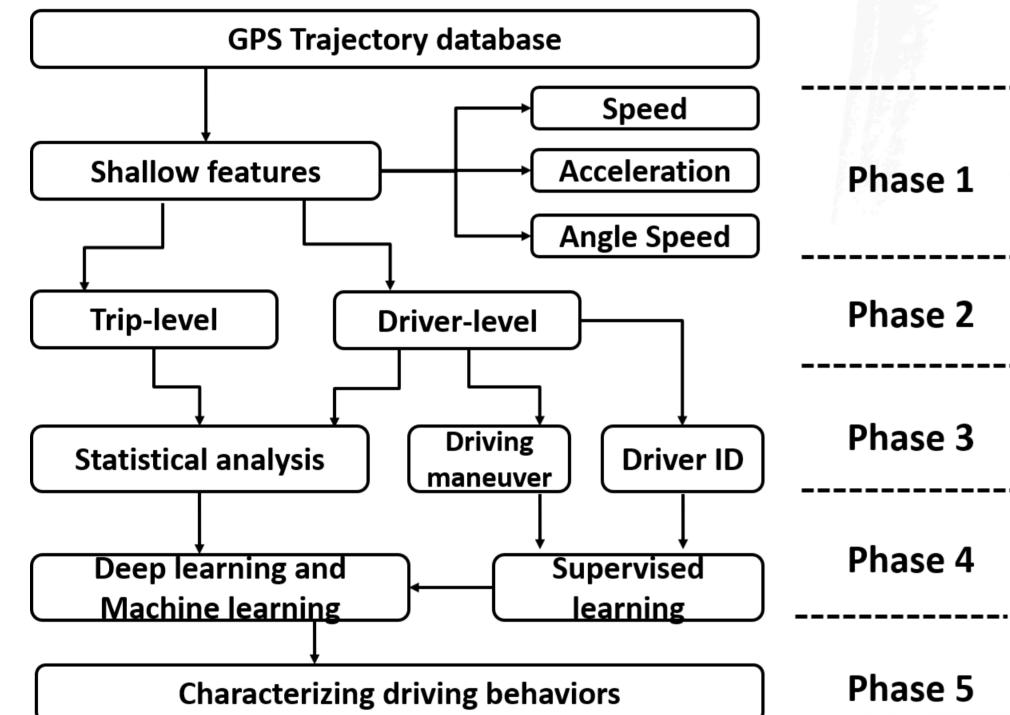
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System Architecture

Driving Maneuver Analysis



The major contributions:

The **general architecture**:

- The shallow driving features can be calculated from a large GPS trajectory database;
- The statistical analysis of driving maneuvers and
- the driver identity will be treated as the
- sematic-level driving features;
- We then introduce a joint histogram feature map as the input to characterize the driving features with the assistance of the artificial neural network;

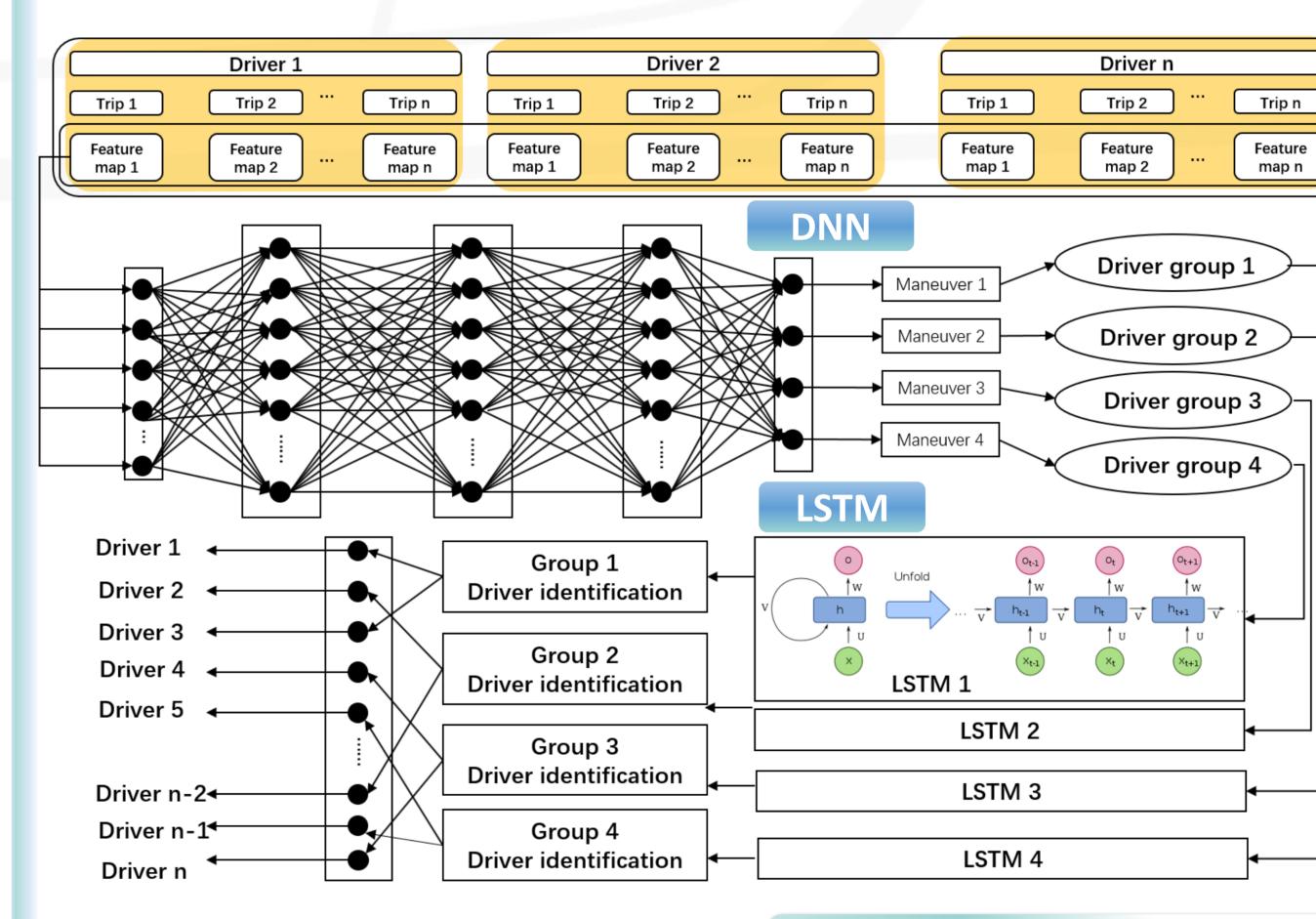
Simulation Results and Discussion

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Simulation results

		Driver Identification			Driving maneuver
Joint- histogram feature maps	Model types	Trip Accuracy	Top5 Accuracy	Driver Accuracy	Classification accuracy
	PCA+SVM	3.13%		2%	20%
	T-SNE+SVM	6.13%		10%	44.24%
	Random Forest	17.73%		18%	39.80%
	DNN	12.35%	35.10%	14%	94.66%
	RNN	28.62%	58.98%	66%	68.01%
	LSTM	36.54%	67.91%	92%	72.38%

And a hierarchical deep learning model for the large-scale driver identification task.



Hierarchical Learning Model DNN: Talented at the general highlevel feature classification; • **RNN:** Talented at the specific highlevel feature classification; **Hierarchical deep learning model:**

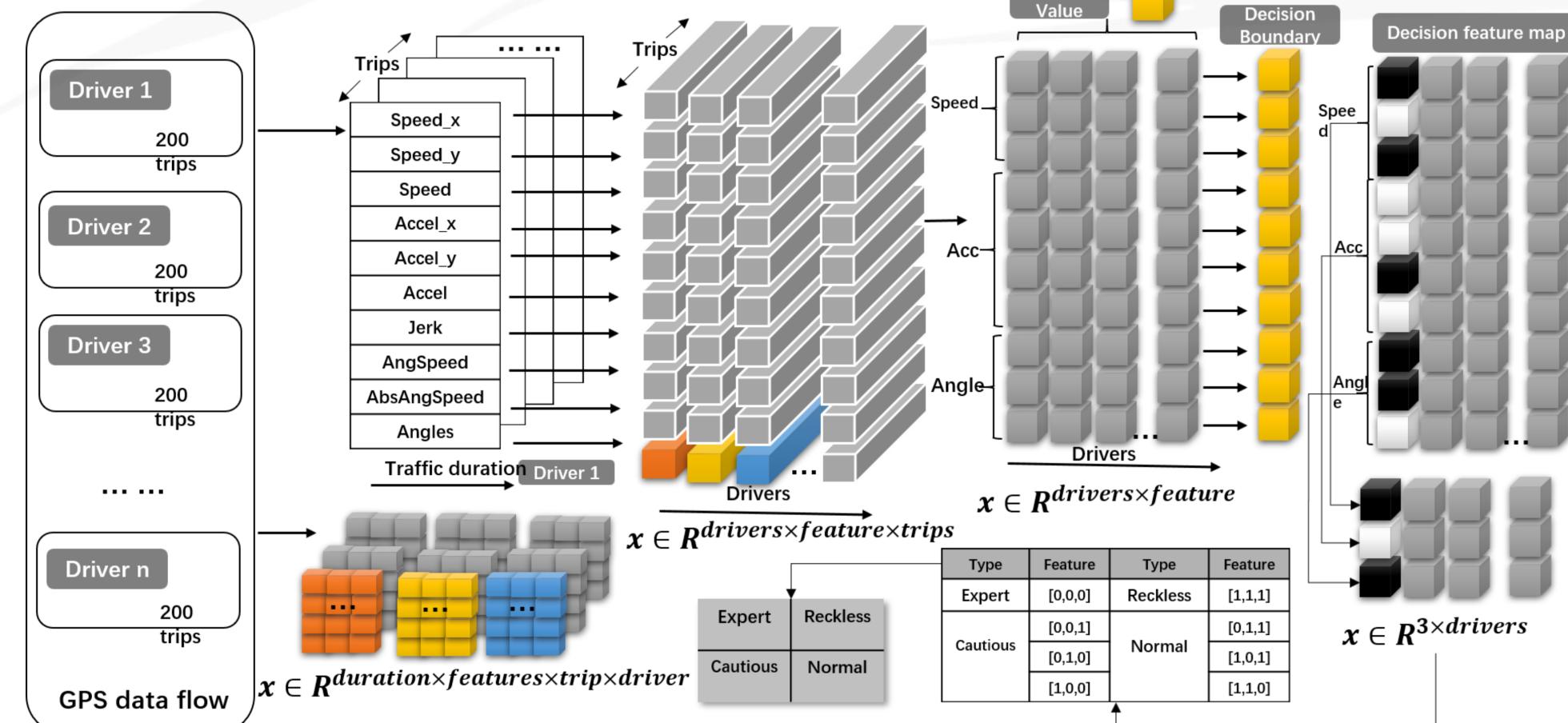
Classify source dataset into the general feature with DNN, and identify the specific feature with multiple RNNs.

Accuracy	LSTM			
	Trip-level	Driver-level		
50 Drivers	36.54%	67.91%		
200 Drivers	17.04%	36.64%		
	Hierarchy deep learning			
50 Drivers	42.08%	78.55%		
200 Drivers	41.52%	73.38%		

Conclusion and Feature Work

Acknowledgement

This study is **detecting the sematic-level driving behaviours** from **GPS sensor data:**



- For **real-time** application, the statistical analysis from both a huge amount of dataset and numerous drivers is not available;
- we aim to propose a working methodology to train autonomous vehicles so that they can have stronger ability to recognize the statistical-analyzed driving maneuvers as well as **to identify a specific driver** under different driving conditions with a **robust** performance.

rip-level GPS coordinates [x_n, y_n, t instantaneous driving behaviors Vel=[speed₁, speed₂,..., speed_{n-1}] $Acc=[acc_1, acc_2, \dots, acc_{n-1}]$ AngSpeed=[angSpeed1, angSpeed2,...,angSpeedn Statistical Driving Feature Map Trip Acceleratio 0.0 2.5 5.0 7.5 10.0 12.5 15.0 17.5 - Neg_Acc Driver, AngSpee Tripk

- We classified different driving maneuvers through a statistical analysis method; The identified maneuver information with the corresponding driver ID is useful for the supervised learning of high-level feature abstraction with neural network;
- We propose a joint histogram feature map to analyze the sensory data with deep **learning** in a consumable form;
- DNN is suitable for the driving maneuver classification task, while LSTM performs well in identifying a specific driver.
- We proposed a hierarchical deep learning model which can well maintain the prediction accuracy even when the scale of the recognition task is four times larger.

For future work, we intend to characterize the driving features under a more

sophisticated condition.



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