

# **Use of a Neural Network Algorithm in Multiple- Sensor Weigh-In-Motion**

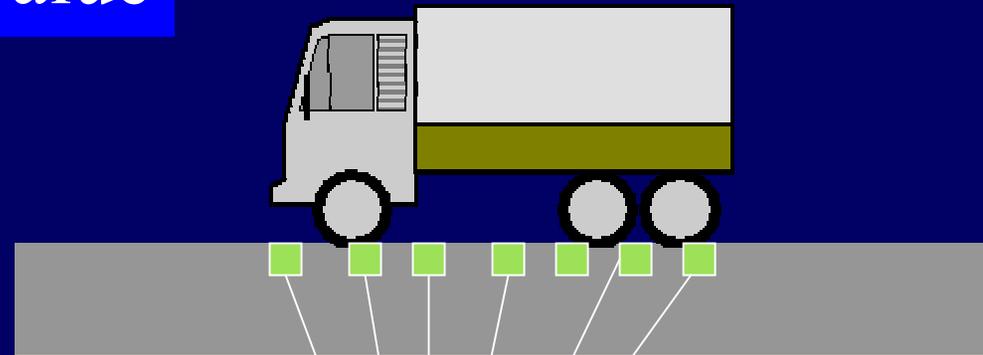
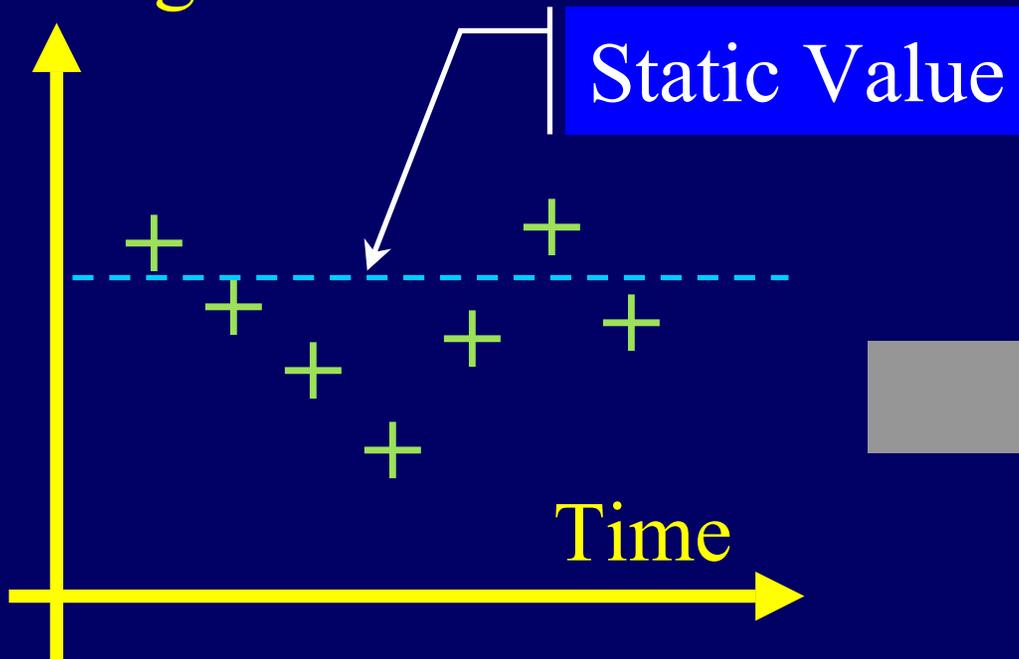
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**<sup>(1)</sup> University College Dublin, Ireland**

**<sup>(2)</sup> The University of Birmingham, United  
Kingdom**

# Multiple-Sensor WIM Systems

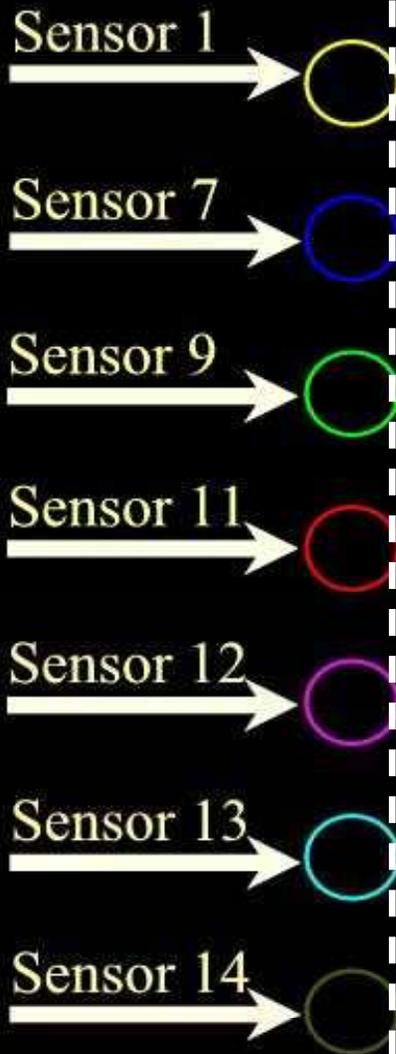
Measured  
Weight



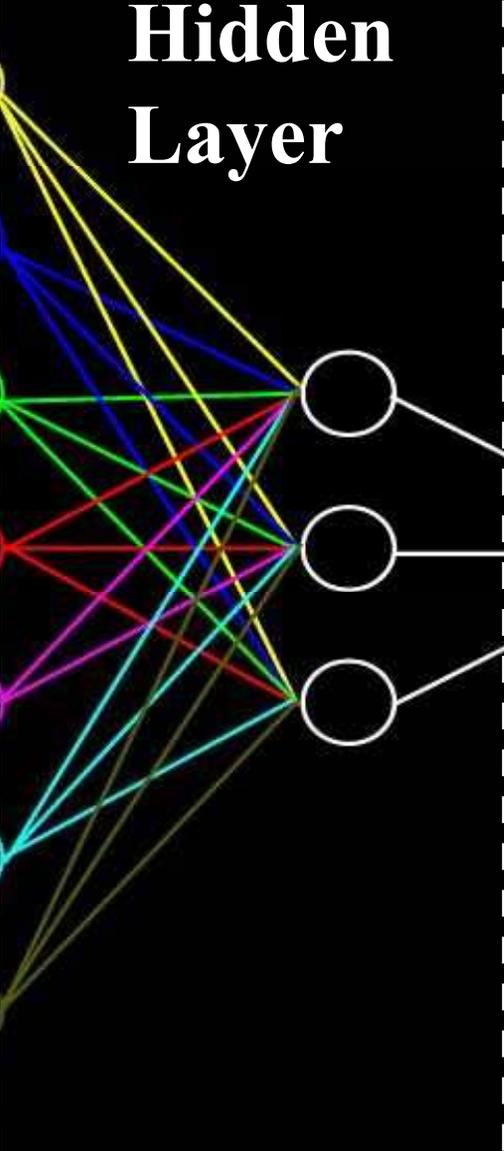
**SENSORS**

# Neural Network Algorithm

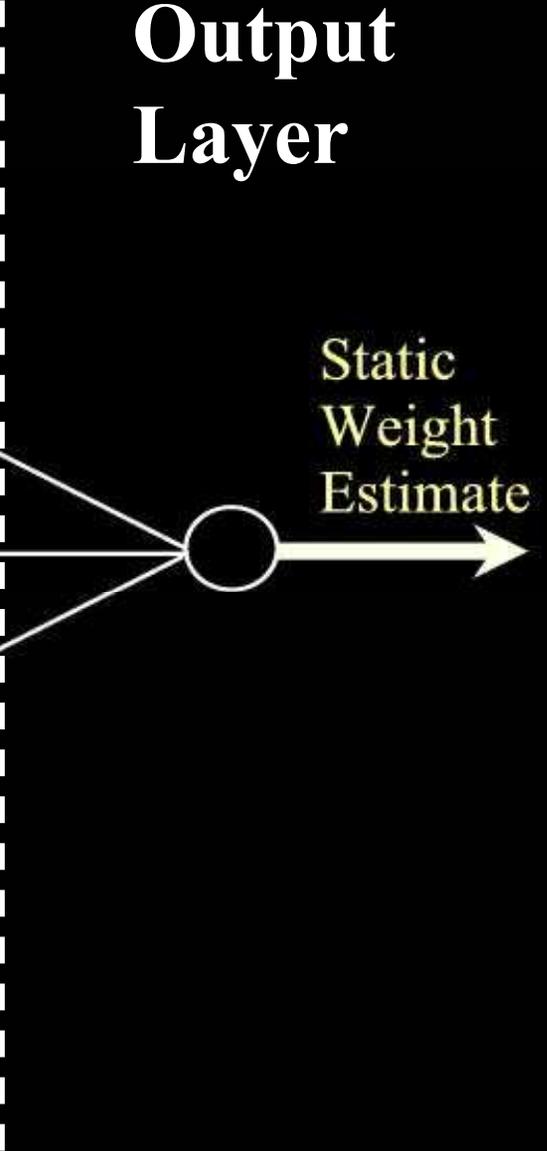
**Input  
Layer**



**Hidden  
Layer**



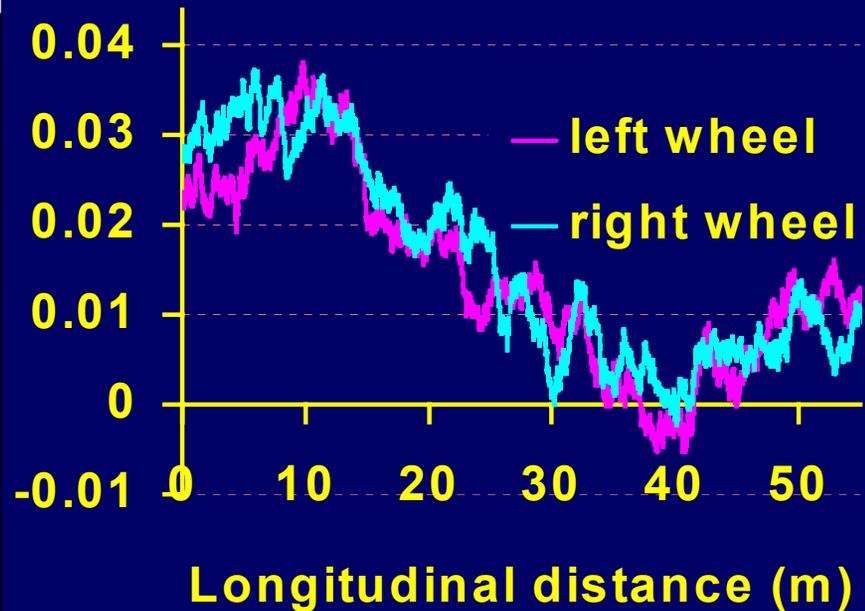
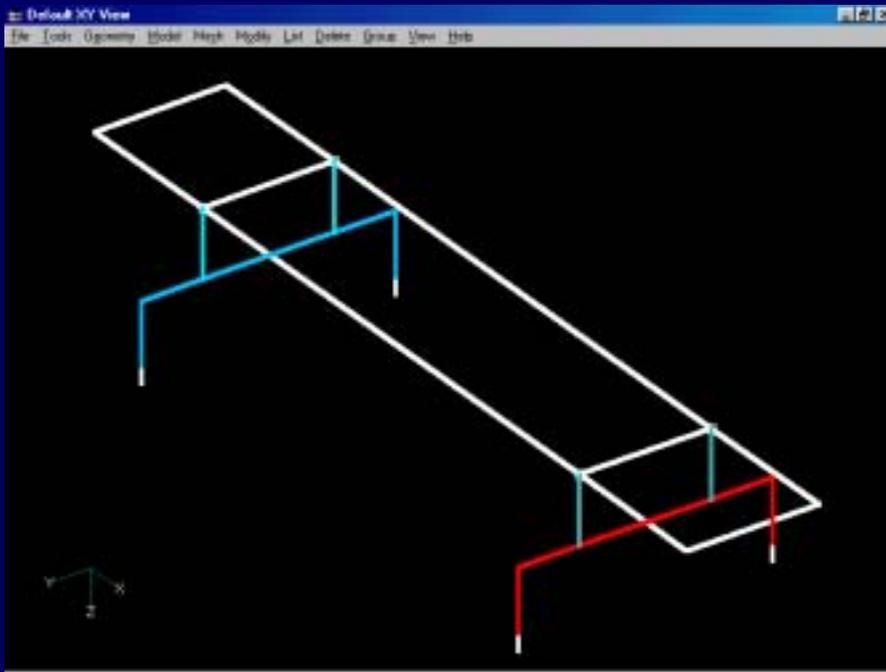
**Output  
Layer**



# Simulations

- 400 runs of a 3-D rigid two-axle truck model.

- Class B Road Profile



- Dynamic Wheel Forces calculated at 7 sensor locations.

# Accuracy and Stability

	Neural Network Output trained with			Average
	Set 1	Set 2	Set 3	
Mean Error	<b>1.59</b>	1.61	1.69	3.39
Stand. Dev.	<b>0.57</b>	0.57	0.57	1.48
Delta min.	<b>3.0</b>	3.0	3.1	7.0

# Influence of a Faulty Sensor on Neural Network

	Biased train. Set		Unbiased Set	
Trained	Set 1-0.3	Set 1-1.2	Set 1	Set 1
Tested	Set 4-0.3	Set 4-1.2	Set 4-0.3	Set 4-1.2
Mean	1.78	1.44	2.82	1.25
Stand.	0.41	0.62	0.44	0.60
Del.Min	2.8	3.0	3.9	2.7

# Influence of Noise on Neural Network Estimate

	Number of Noisy Sensors		
	1	3	7
Mean Error	<b>1.59</b>	1.54	1.60
Standard Dev.	<b>0.57</b>	0.76	0.90
Delta min.	<b>3.0</b>	3.4	3.9

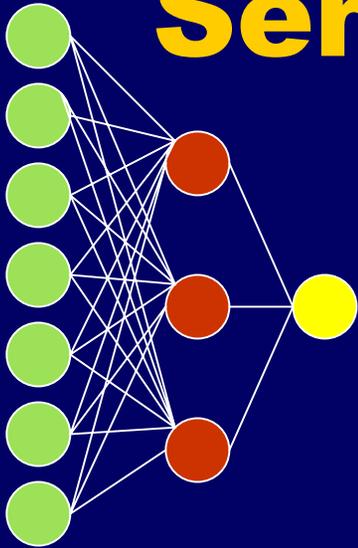
# Testing with Field Data

- Data from 11 two-axle trucks, 21 four-axle trucks and 42 five-axle trucks (Trappes, France).



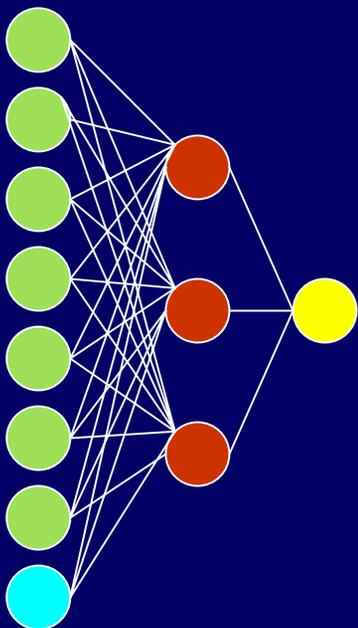
- From a large 24-sensor array, the most accurate sub-array of **7-sensors** is selected to train the Network.

# Accuracy using as Input Sensor Weights Only



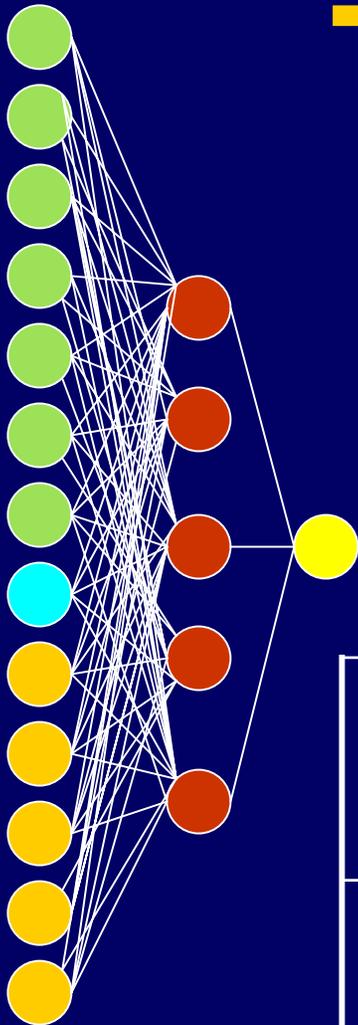
	Neural Network		Average Method	
	Delta min.	Class	Delta min.	Class
<b>GVW</b>	<b>5.8</b>	B+(7)	7.5	B+(7)
<b>Single Axles</b>	<b>9.1</b>	B+(7)	9.2	B+(7)
<b>All Axles</b>	<b>11.4</b>	B(10)	11.7	B(10)

# Accuracy using Speed as an Input



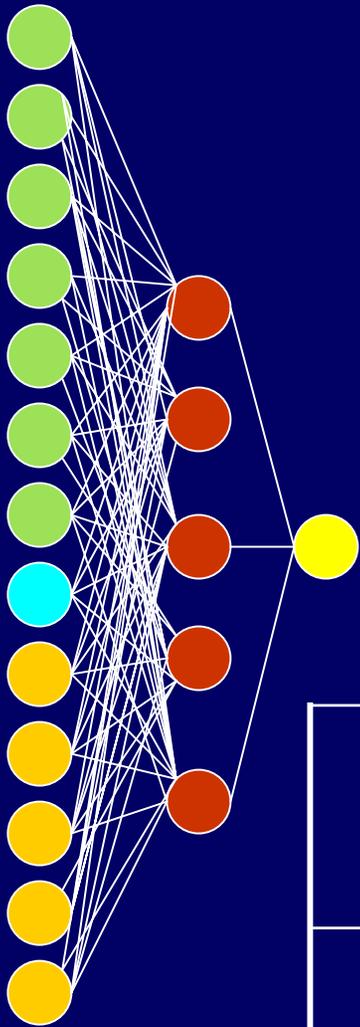
	Neural Network		Average Method	
	Delta min.	Class	Delta min.	Class
<b>GVW</b>	<b>5.9</b>	B+(7)	7.5	B+(7)
<b>Single Axles</b>	<b>8.6</b>	B+(7)	9.2	B+(7)
<b>All Axles</b>	<b>12.6</b>	B(10)	11.7	B(10)

# Accuracy using Type of Truck as an Input



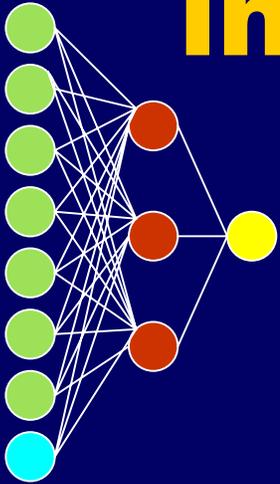
	Neural Network		Average Method	
	Delta min.	Class	Delta min.	Class
Single Axles	<b>9.7</b>	B+(7)	11	B(10)
All Axles	9.3	B+(7)	<b>8.1</b>	B+(7)

# Accuracy using Rank of Axle as an Input



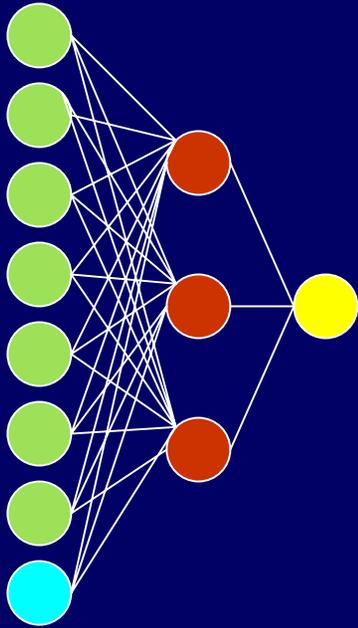
	Neural Network		Average Method	
	Delta min.	Class	Delta min.	Class
Single Axles	7.3	A(5)	11.9	B(10)
All Axles	12.6	B(10)	12.8	B(10)

# Accuracy Separating Data into Weight Divisions



	Neural Network		Average Method	
	Delta min.	Class	Delta min.	Class
<b>Heavy ( &gt; 6 tonne)</b>	<b>8</b>	A(5)	9.6	B+(7)
<b>Medium (From 4 to 6 t)</b>	<b>7.4</b>	A(5)	8.5	B+(7)
<b>Light ( &lt; 4 t)</b>	27	D(25)	<b>24.1</b>	D+(20)

# Accuracy using Calibrated and Uncalibrated Data



	Uncalibrated Data		Calibrated Data	
	Delta min.	Class	Delta min.	Class
GVW	5.7	B+(7)	5.9	B+(7)
Single Axles	8.7	B+(7)	8.6	B+(7)

# Conclusions

- **Simulated Traffic Data** indicated that the Neural Network remained **Stable** and **Largely Unaffected by Bias** or **Noise** in the sensor readings.
- From **Field Data**, Neural Networks were generally **More Accurate than the Average Method**. Accuracy can be further improved by considering as an Input, not only Sensor Weights, but also **Speed, Rank of Axle, Type of Truck and Separating Axle Weights into Weight Divisions**.