

asefma XV Jornada Nacional de Asefma 1-2 de diciembre de 2020 Exploring the impact of road surface conditions on truck fleet fuel consumption using Big Data

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Background

- Experiments (e.g. Sandberg et al 2011) and modelling (e.g. Chatti and Zaabar 2012) suggest that road surface characteristics (evenness and texture) can influence fuel consumption by up to 5%
- But maybe only 2% or 3% for UK highways
- So road maintenance could save fuel consumption, costs and emissions for the whole vehicle fleet (unlike new vehicle technologies)
- But these experiments and models are soon out of date
- And the findings are difficult to validate for real vehicle fleets, at network level and for real driving conditions where measurements are confounded by so many variables.

Big Data



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Data from sensors installed on modern trucks inform fleet managers (SAE J1939)



A model for:

highways england

- Routing of vehicles;
- Help in the design and maintenance of pavements;
- Improve Life Cycle Assessment and Life Cycle Cost Analysis of road pavements
- Live network, truck and fuel consumption information

England Case Study

During 1 week in October 2016:

- ~ 300 km of motorway
- Constant speed (+/- 2.5 km/h)
- 1,110 articulated trucks (Euro 6 with 6 axles)
- 14,281 records
- 1 minute or 1 mile
- Fuel used in 0.001 l

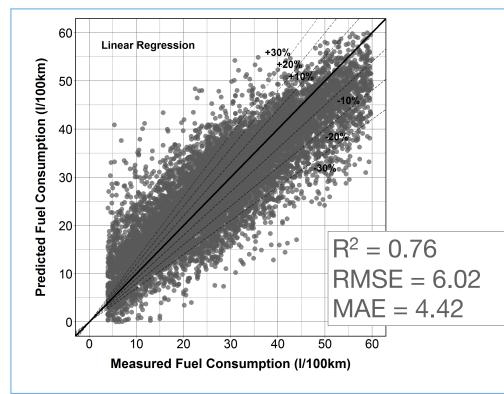


Included Variables

- Gross Vehicle Weight
- Vehicle constant speed (+/- 2.5 km/h)
- Acceleration
- Torque Start
- Torque End
- Revolutions
- Gear
- Cruise control

- Road gradient
- Radius of curvature
- Unevenness LPV03 m
- Unevenness LPV10 m
- Unevenness LPV30 m
- Macrotexture (SMTD)

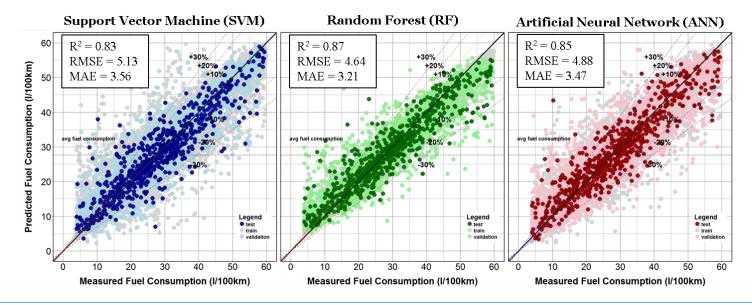
Multiple Linear Regression



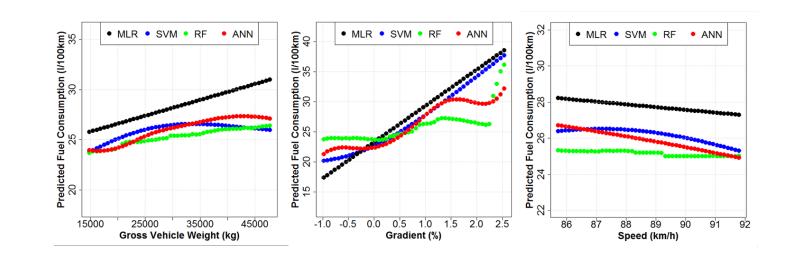
Can we improve this using machine learning?

Machine Learning

72% records for training, 23% records for validation, 5% records for test 10-fold cross-validation: R², RMSE, and MAE calculated on test sets (average of 10).

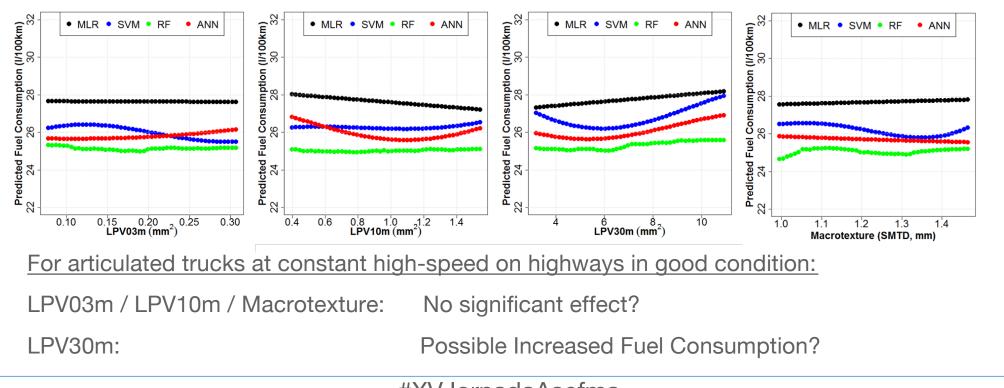


Parametric Analysis



Parametric Analysis

Perrotta et al (2017) DOI: 10.1109/BigData.2017.8258382



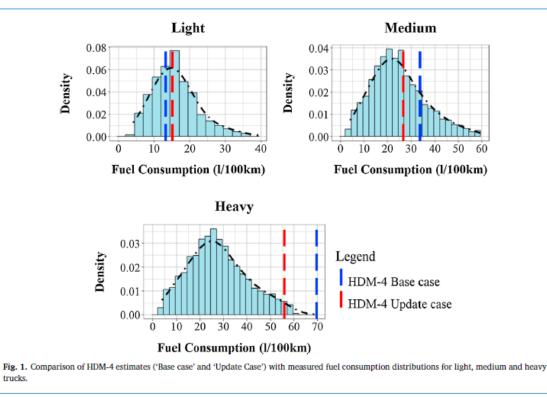
Ongoing before-and-after maintenance studies





HDM4 Calibration

Perrotta et al (2019) DOI: 10.1016/j.trd.2018.11.001



Roundabout Collision Risk

Kamla et al (2019) DOI: 10.1016/j.aap.2018.04.31



Fig. 1. Grade-separated roundabout (J28 on M1 motorway, UK) with accident (left) and truck HBI (right) positions.

Roundabout Collision Risk

Kamla et al (2019) DOI: 10.1016/j.aap.2018.04.31

Table 7

Roundabout category	Variable	Total accident numbers	Truck accident numbers	HBI numbers
Whole roundabouts	ln(AADT)	<u> </u>	<u>^</u>	<u>^ </u>
	Percentage of average annual daily truck traffic	86% 1	$\wedge \uparrow$	$\Lambda \Lambda$
	Un-signalised roundabouts compared to partially signalised roundabouts	$\checkmark \checkmark$	44	51%
	Signalised roundabouts compared to partially signalised roundabouts	-	76%	48% \downarrow 🗸
	Two-lane roundabouts compared to three-lane roundabouts	-	66% 🗸	-
	ICD	1	1	-
	Three-arm roundabouts compared to six-arm roundabouts	-	↓↓	48% √√
	Circulatory roadway width	-	\downarrow \downarrow	$\sqrt{\sqrt{1}}$
	Entry width	-	-	^

Conclusions

- Case study shows the feasibility of using Big Data to:
 - Monitor the impact of road condition on vehicle fuel consumption
 - For real vehicles and driving conditions at highway network level
 - Continuously
- Many other potential uses for this Big Data approach, including:
 - Calibration of pavement-vehicle interaction models
 - Life Cycle Assessment use phase / Life Cycle Cost Analysis of road maintenance
 - Collision risk



ASCEACCON ESPARIOLA DE FARRICANTES DE MEZCLAS ASFALTICAS XV Jornada Nacional de Asefma 1-2 de diciembre de 2020



Gracias por su atención!!

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