



Can you "sniff" what <u>type</u> of car is passing by?

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Remote sensing measurements are great tools e.g. for fleet monitoring -

but...mass photographing of vehicle number plates may infringe persona data (GDPR)

=> No <u>routine</u> use of remote sensing for fleet monitoring to date 😕







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Can fuel type and emission class be """, sniffed" from emission records only?





"Sniff" fuel type & emission standard of a car

Approach: Use chassis and PEMS data for cars Train two machine learning algorithms

Transfer to remote sensing data

Aim: Identify fuel type (petrol or diesel) and emission standard (Euro 5, 6b, 6d-TEMP, 6d) from pollutant measurement

When successful, we can also sniff high-emitters!!!





Data sources and treatment

"Reference data" (ERMES/ARTEMIS db):

Emissions, fuel flow, driving parameters from chassis dynamometer (hot ARTEMIS cycles) and PEMS RDE trips second-by-second [g/sec].

- Take moving 3-sec averages to improve time alignment and calculate ratio pollutant/fuel flow => 22'000 records [g/kg fuel]_{sec}
- Filter for v-a/driving situations & temperature comparable to RS values

"Field data" (CONOx/CARES remote sensing db):

Instantaneous remote sensing records from campaigns across Europe, mostly urban driving conditions: 63'000 records [g/kg fuel]_{sec}





Identification of fuel type on reference data

Given: Instantaneous pollutant ratios from reference data. => Train algorithms on reference data, then make blind test.



With reference data:

- Estimate of petrol or diesel type possible,
- based on NOx, CO and HC as measured,
- no gain from v, a, VSP
- Up to 90+% accuracy of detection



Identification of fuel type on field data

Given: Instantaneous RS records drawn from field data. => Apply algorithms on RS data (where fuel type is known).



With RS field data:

- Estimate of petrol or diesel type possible,
- based on NOx, CO and HC as measured,
- no gain from v, a, VSP
- using 5 RS records per vehicle up to 90% accuracy of detection





Identification of emission class on reference data

Given: Instantaneous pollutant ratios from reference data. => Train algorithms on reference data, then make blind test.



With reference data:

• Estimate of Euro 5, 6b, 6d-TEMP, 6d

emission standard possible,

- based on NOx, CO and HC as measured,
- no gain from v, a, VSP
- Up to 85% accuracy of detection





Identification of emission class on field data

Given: Instantaneous RS records drawn from field data. => Apply algorithms on RS data (whose emission class is known).



With RS field data:

- Estimate of Euro 5, 6b, 6d-TEMP, 6d emission standard possible,
- based on NOx, CO as measured,
- no gain from v, a, VSP
- gets bad, when also HC included 😣
- Up to 80% accuracy of detection



Results – and open questions / your input needed!

Proof of concept: Estimate 80%+ accurate of fuel type and emission class possible based on a few emission records.

=> Apply to identify vehicles with suspiciously high emissions!

For enforcement no need to photograph ALL number plates, but only those 10%+, where emissions are suspiciously high. (technical data though important for other detailed analyses)

Open questions:

- Mismatch or accuracy issue for HC measurements!?
- More reference data (PEMS) needed!
- Test other algorithms and vehicle classes





Further reading / information

- MSc thesis (2023, in English) by Lukas Unterschütz (TU Dresden)
- Poster by Unterschütz et al. at TAP 2023 Conference: "Data Protection in Remote Sensing through Profiling High-Emitting Vehicles"
- Jens.Borken-Kleefeld@TU-Dresden.de www.verkehrsoekologie.de



Data Protection in Remote Sensing through Profiling High-Emitting Vehicles

Inference from Machine Learning Models Trained on PEMS and Chassis Data



