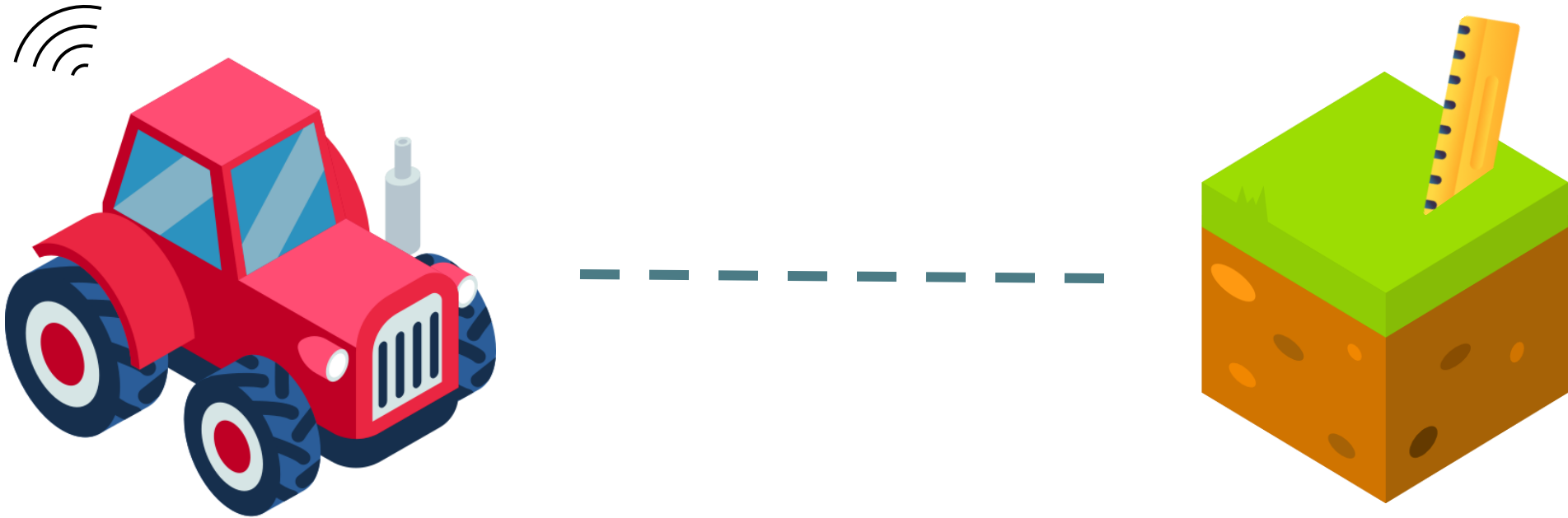




The potential of sensor technology for soil condition assessment

Heli Leskelä, Petri Junttila 12.12.2024

Is there a connection between tractor's sensor data and soil conditions?



Case example: from field to results

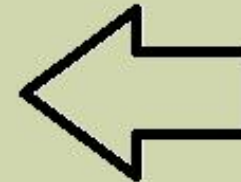
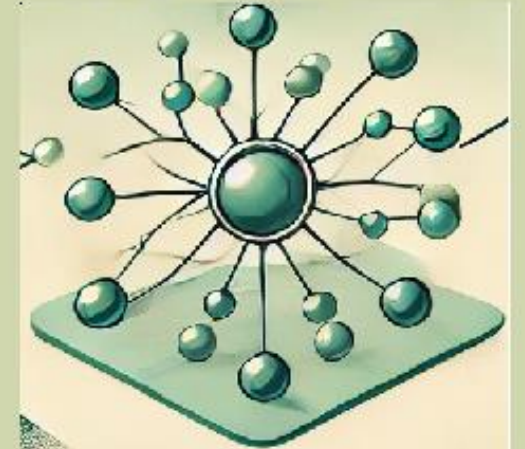
Collecting data from the field



Data analysis and preparation



Training Neural Networks



Examining results



Test drive

- Multiple test tracks
 - Total 1 km on a soft soil
 - Total 600 m on sand
- Standardized test runs
 - Tire pressures 2.1 bar
 - 15 km/h speed
 - Same driver every time



Penetrometer

(The variable to be predicted)

- Measures penetration resistance (MPa)
 - Cone-index
- Measured before driving
- Around 140 measurements in total
- Soft soil and sand

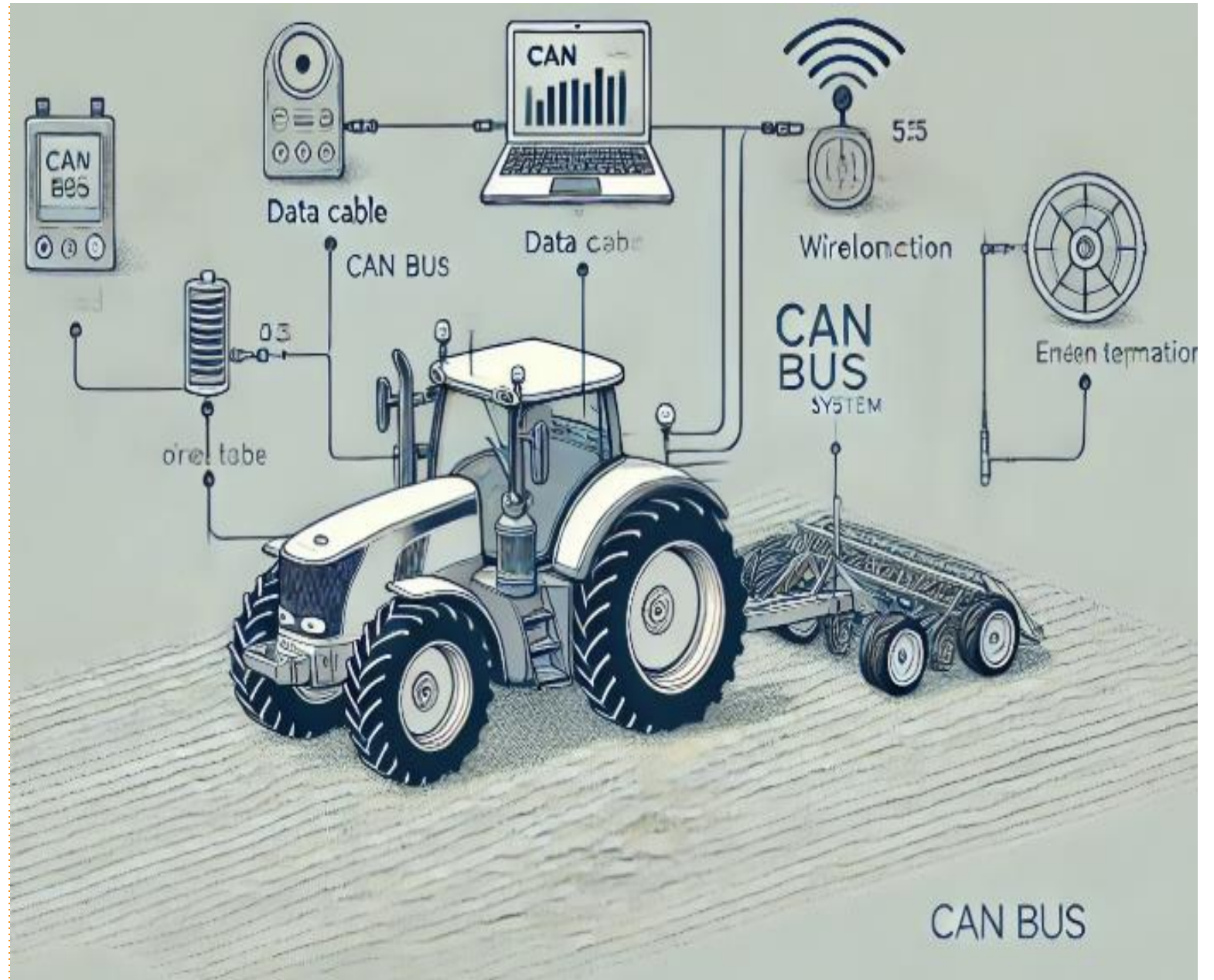


Petri Junttila (left) ja Otto Läspä (right)

CAN-bus

(Controller Area Network)

- GPS
- Wheel speed
- Ground speed
- Slip
- Engine speed
- Fuel rate
- Etc...





Data preparation

- Important to accurately combine data
 - CAN-bus and penetrometer from the same point of the field
- GPS points were used to combine everything into one Excel
- Normalization for neural networks
 - Example: Engine rpm range is 0-3000, then 1500rpm is 0.5

Training Neural Networks

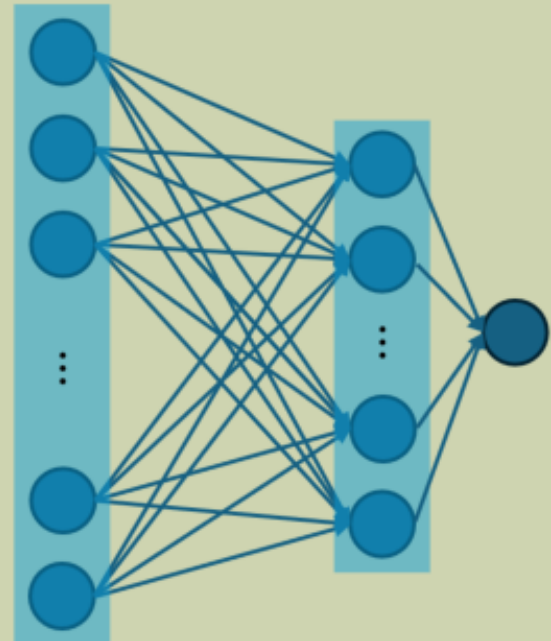
$$\omega_1 \cdot x_1 + \omega_2 \cdot x_2 + \omega_3 \cdot x_3 + b = z$$

$$f(z) = \text{output of neuron}$$

Can-bus data { Variables:
Vehicle speed
Torque
Engine power
Etc.

Value:	Normalized:
5 km/h	x1=0.1
100 Nm	x2=0.3
30 Kw	x3=0.3

Neural Network

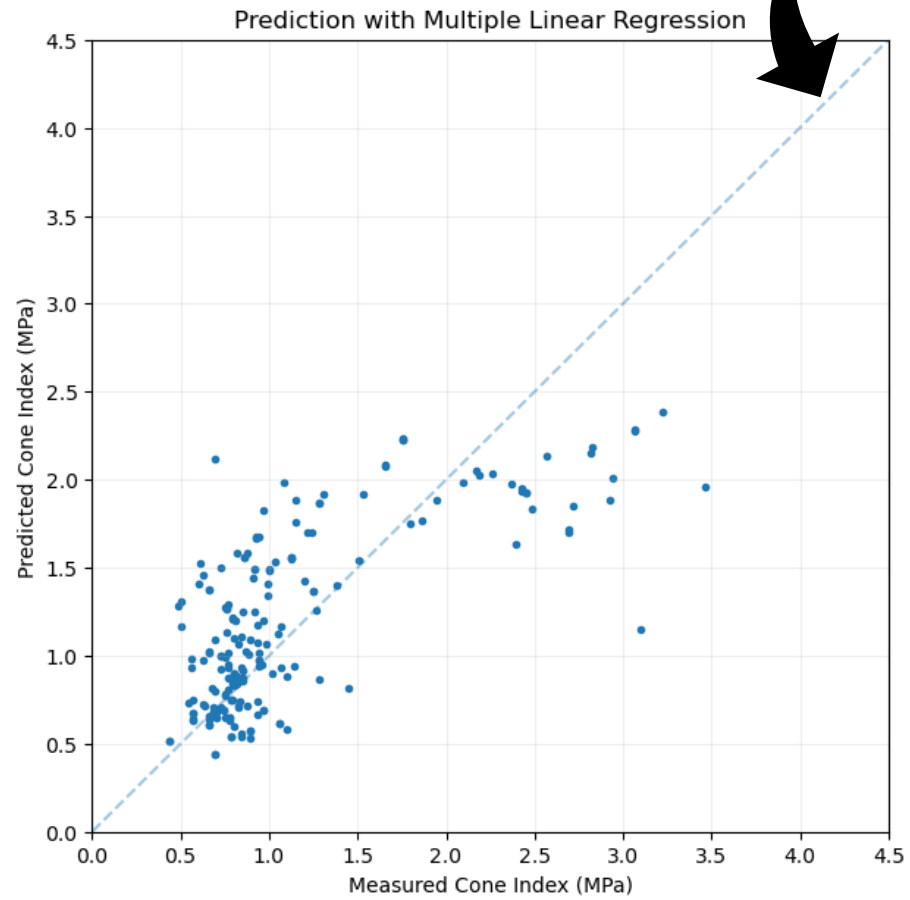


Prediction:	Measured
0.1 Mpa	Cone index:
	2.2 MPa

Step2:
Backpropagation ←

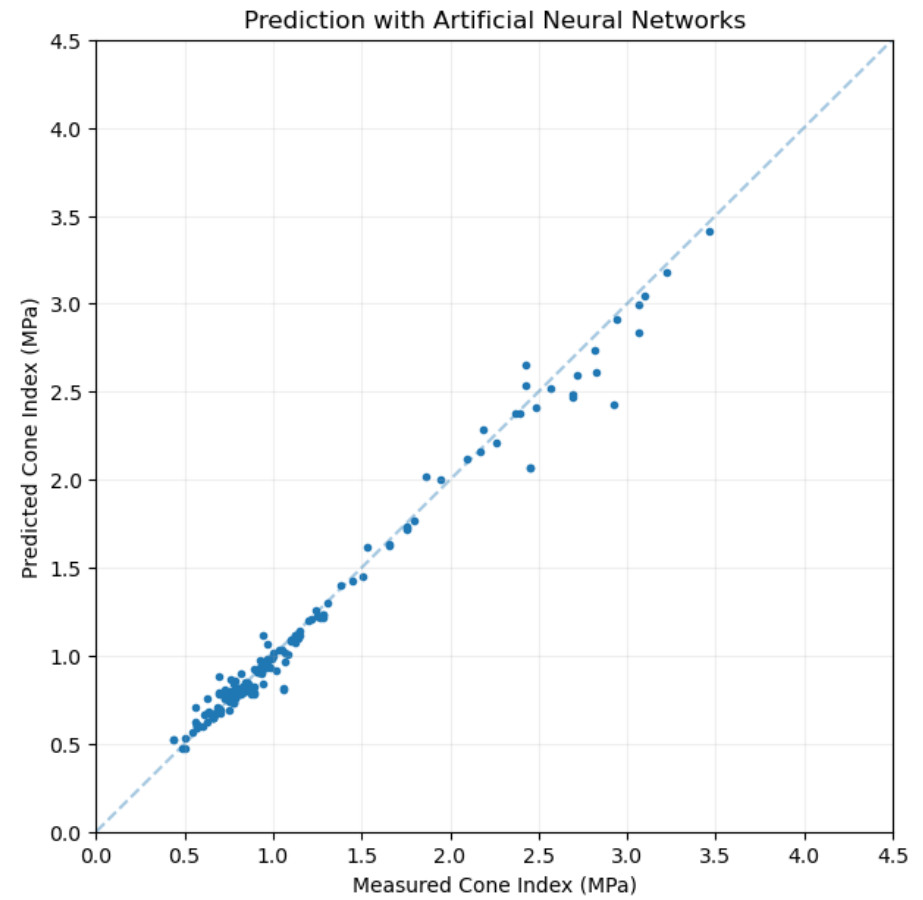
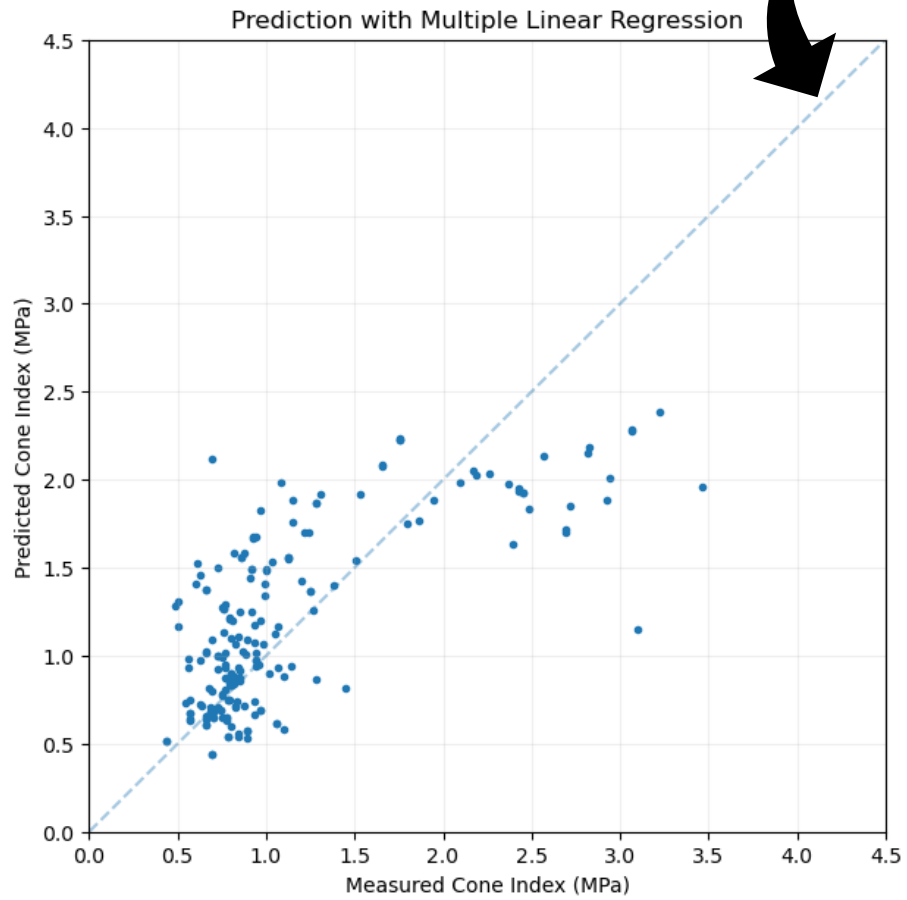
Results

The line represents
a perfect prediction



Results

The line represents a perfect prediction



Conclusions

- Very good results
- Indication that it is possible to predict soil conditions from sensor data
- The next step: finalize the scientific paper



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