

# Soutenance de thèse de doctorat (PhD)

Application of Artificial Intelligence in the exploitation of deflection measurements  
for the mechanical characterization of pavement structures

Abdelgader Abdelmuhsen, Doctorant (LAMES – MAST)

- Rapporteur : Prof. Khalil Drissi
- Rapporteur : Prof. Simona Fontul
- Examinateur : Prof. Fekri Meftah
- Examinateur : Prof. Yide Wang
- Directeur de thèse : Dr Amine Ihamouten
- Co-directrice de thèse : Dr Franziska Schmidt
- Encadrant : Dr Jean-Michel Simonin

# Outline

1. Global problem statement
2. The state of the art
3. Research gaps = Thesis objective
4. AI-based intelligent solution
5. Systematic methodology
6. Results & discussion
7. Research limitations & challenges
8. Conclusion, perspective & valorization



# Introduction

- Context and implications
- Causes and consequences

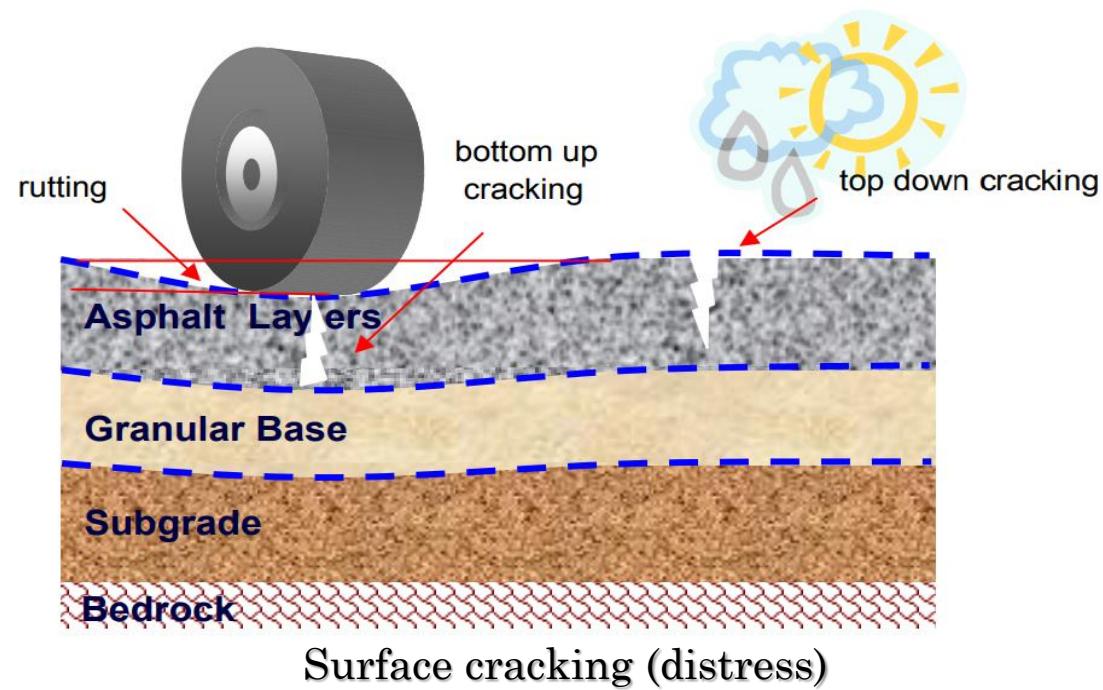
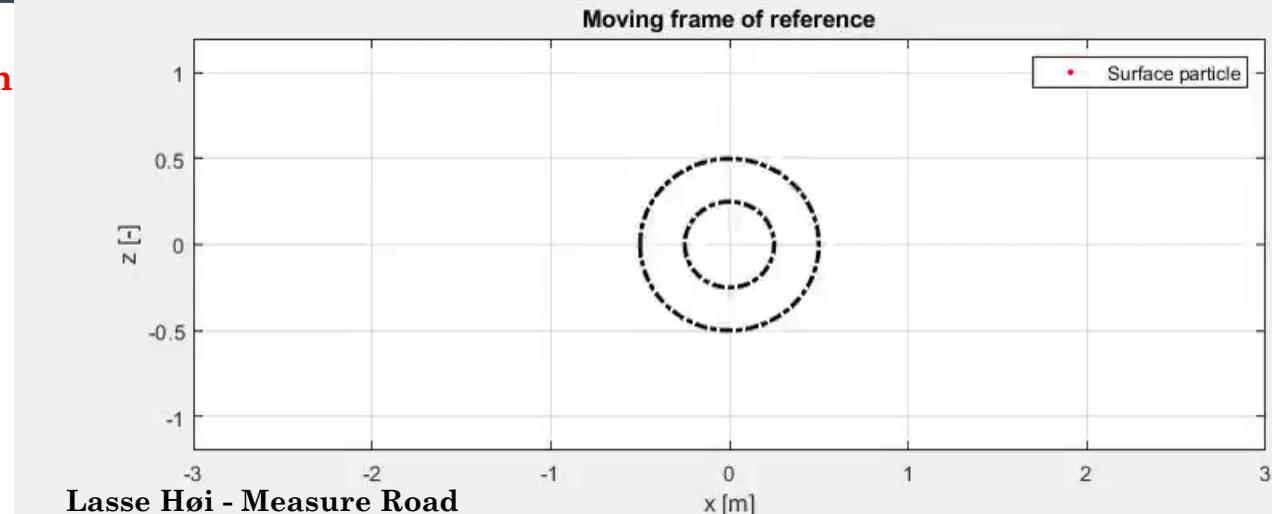
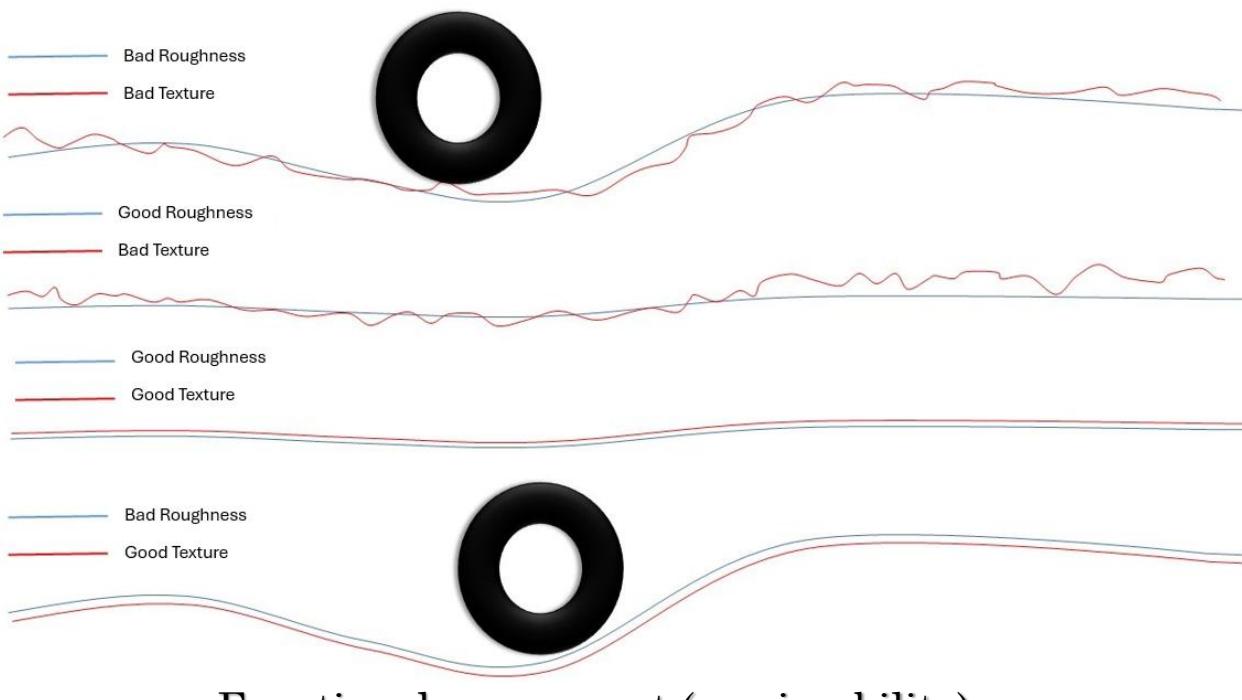
# The state of the art

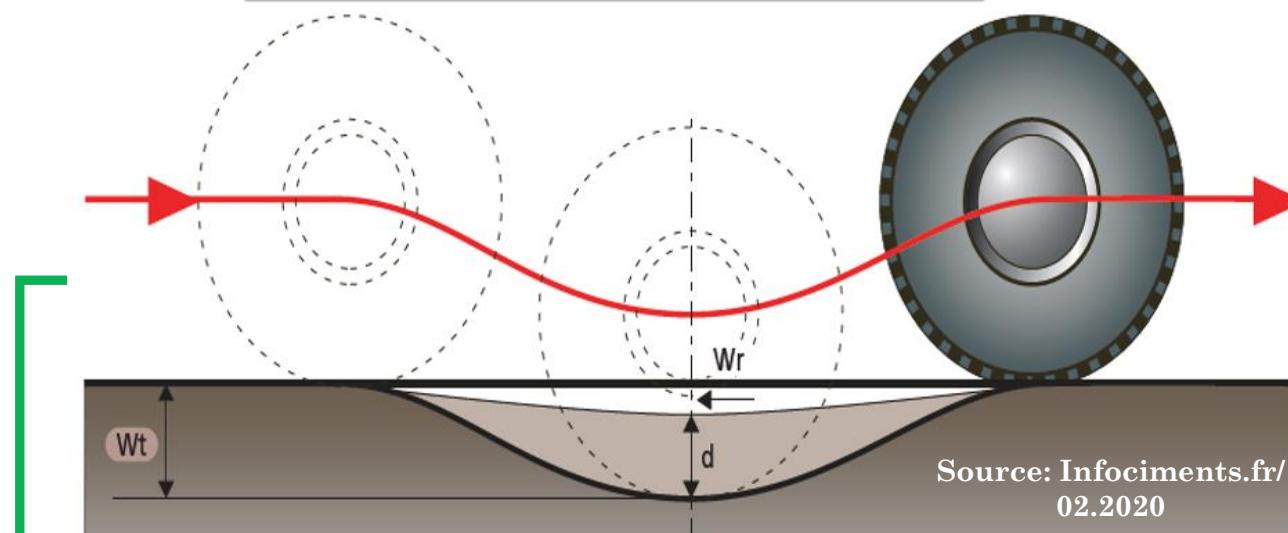
## Context

## Challenges

## Motivation

- Road Authorities: critical progressive infrastructure **deterioration**
- Insufficient **DATA**: allowing for long-term predictive maintenance
- PMS indicators**: Relies on functional assessment & distress data
- Both might not completely capture the true **structural integrity**
- Gaps: Structural **bearing capacity** assessment at Network-level



NDT Deflection measurement

Reference (2) : J-M, Simonin, 2021, correction

$(E_i)$ : Fundamental mechanical property

Inverse Model (Backcalculation)

$(E_i)$

Modulus Estimation

Traditional NDT deflection devices operational limitations

Static  
Benkelman



Stationary (Impulse)  
FWD



Slow move 8 km/h  
Deflectograph

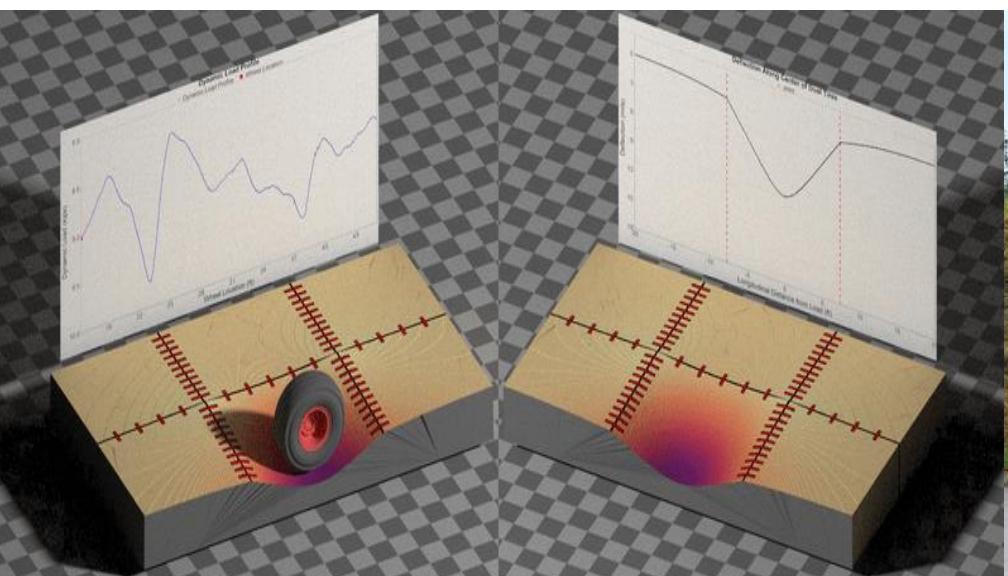


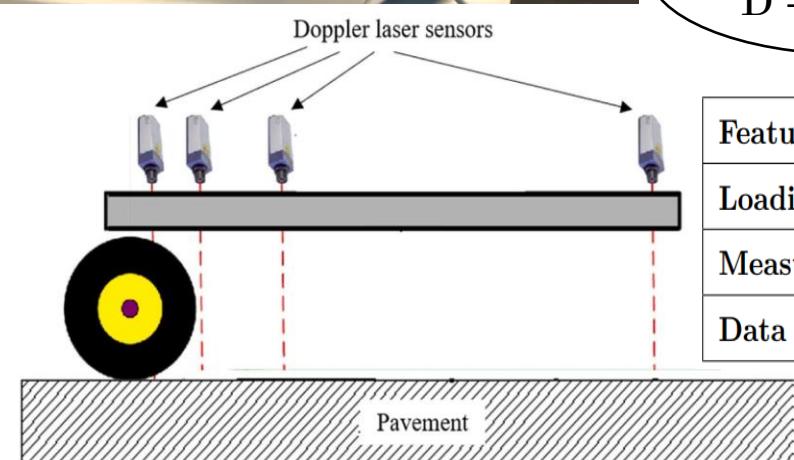
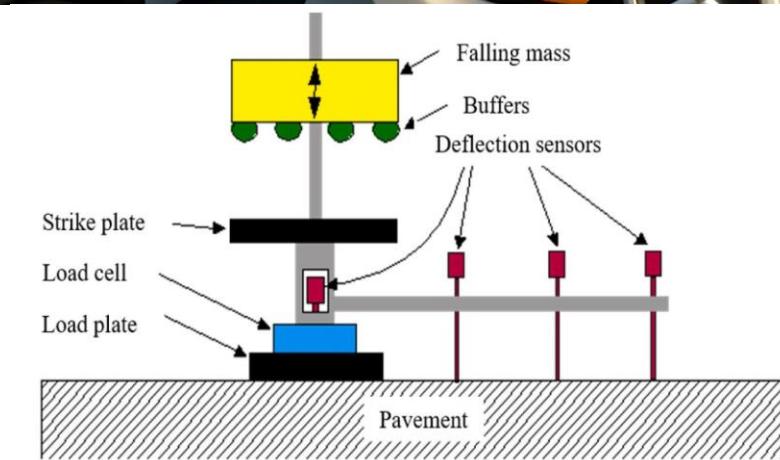
Constant move 8 km/h  
Curviameter

## TSDDs overcome FWD limitations via:

- ❖ Continuous bearing capacity assessment
- ❖ Measurement at traffic speed (+80 km/h)
- ❖ Secure with no traffic disruption
- ❖ Cost-effective for projects and networks levels

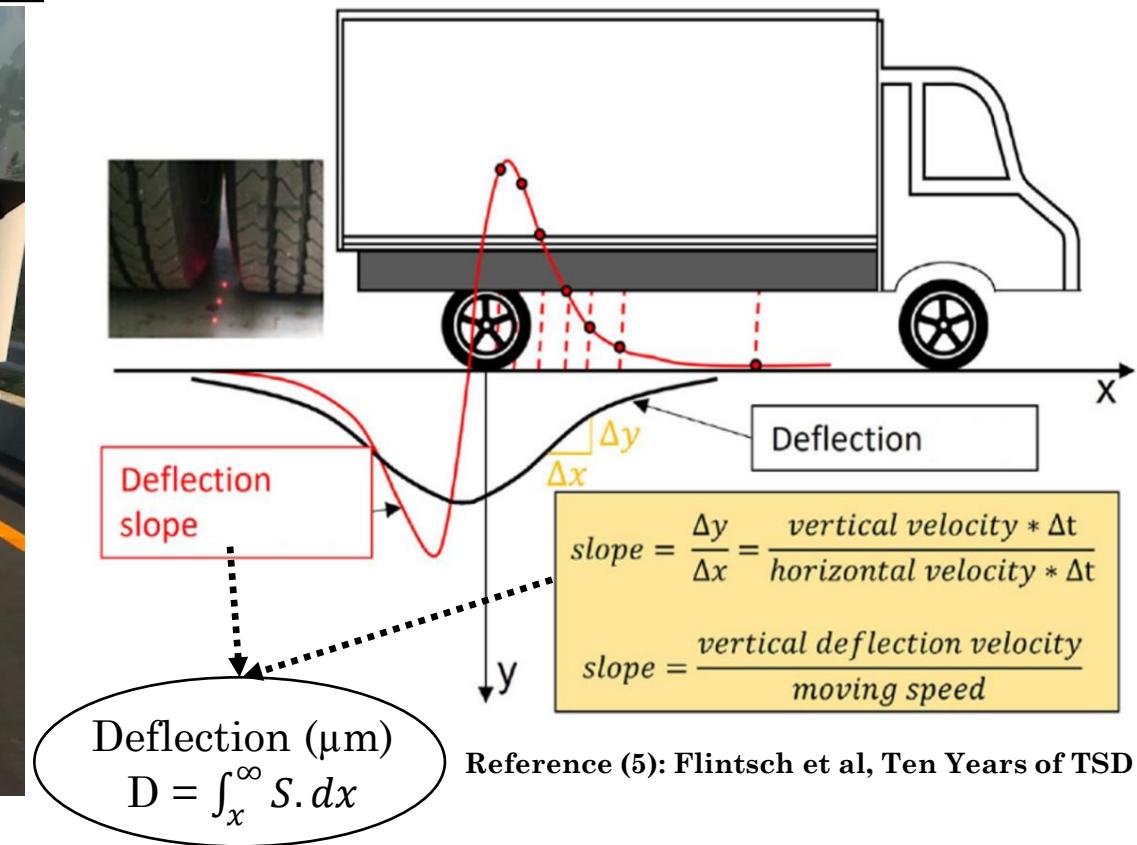
Reference (3): Katicha, and Flintsch, Ten years TSD research.....





(a) The measuring schematic of the FWD

(b) The measuring schematic of the TSD



Reference (5): Flintsch et al, Ten Years of TSD

| Feature            | (FWD)                     | (TSD)                           |
|--------------------|---------------------------|---------------------------------|
| Loading mechanism  | Single plate (stationary) | Dynamic or moving tires         |
| Measured parameter | Pavement deflection       | Deflection slope                |
| Data conversion    | No conversion required    | Integration, (information loss) |

Reference (4): F. Xiao, Tongji University, Measurement, 2021

✓ TSD efficiently collects structural data at traffic speeds

✗ Processing, analyzing & interpreting TSD data: key challenges!

Strategic

Industrial

Research

Priorities

# This is a workshop!



**What is your opinion about ...**

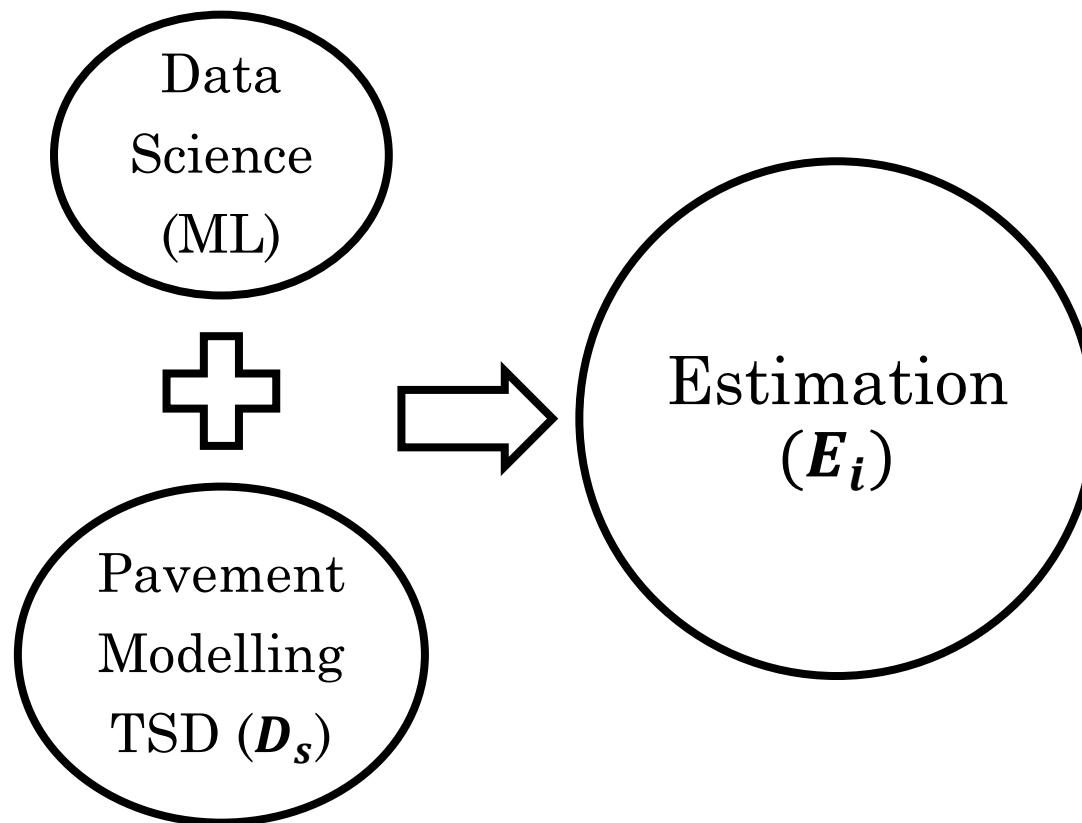
1. Choosing a method to calculate deflections?
2. Using slope values instead of deflections?
3. Establishing international standards for TSD measurements?

Dr. Dirk Jansen

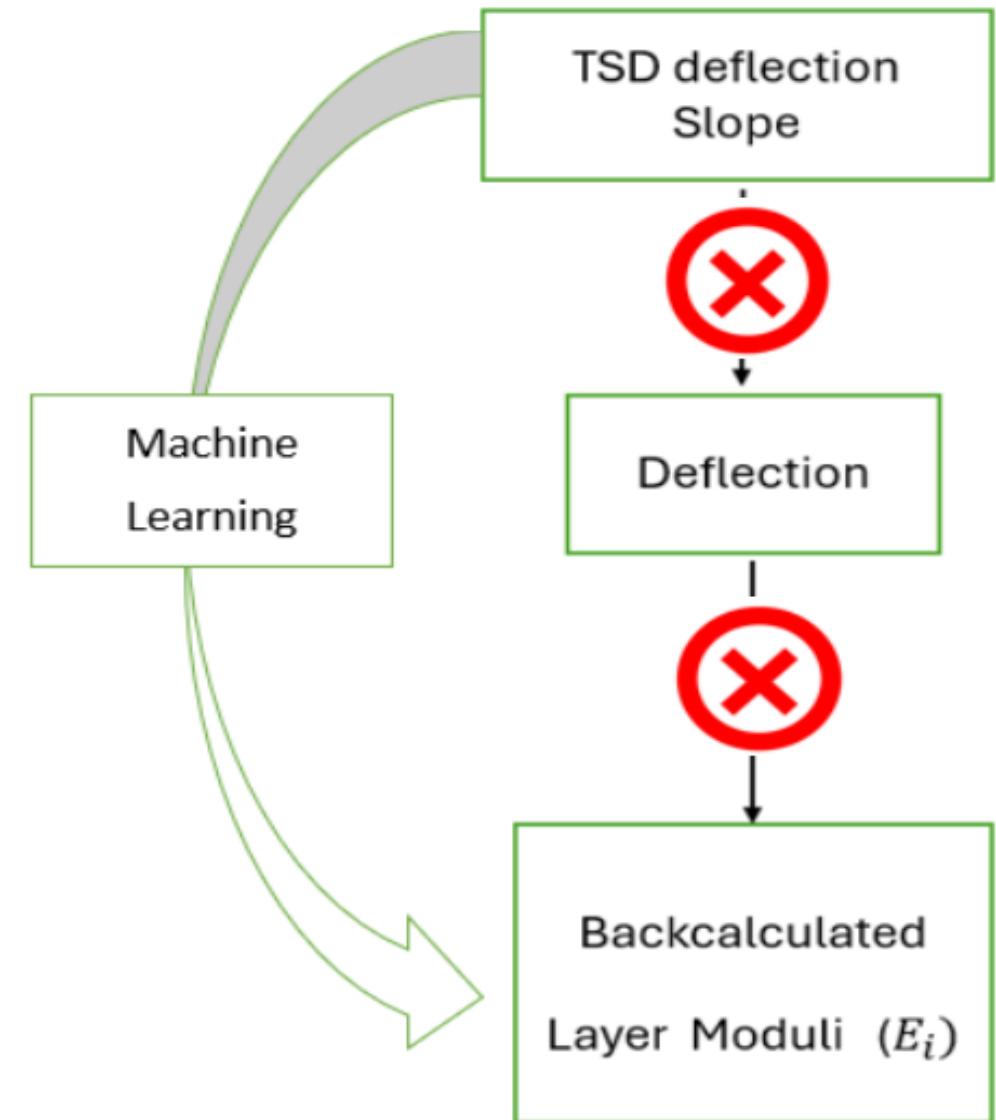
TRB Annual Meeting - US, Washington, D.C, January 2019

► Main Objective :

Develop an ML-based intelligent framework



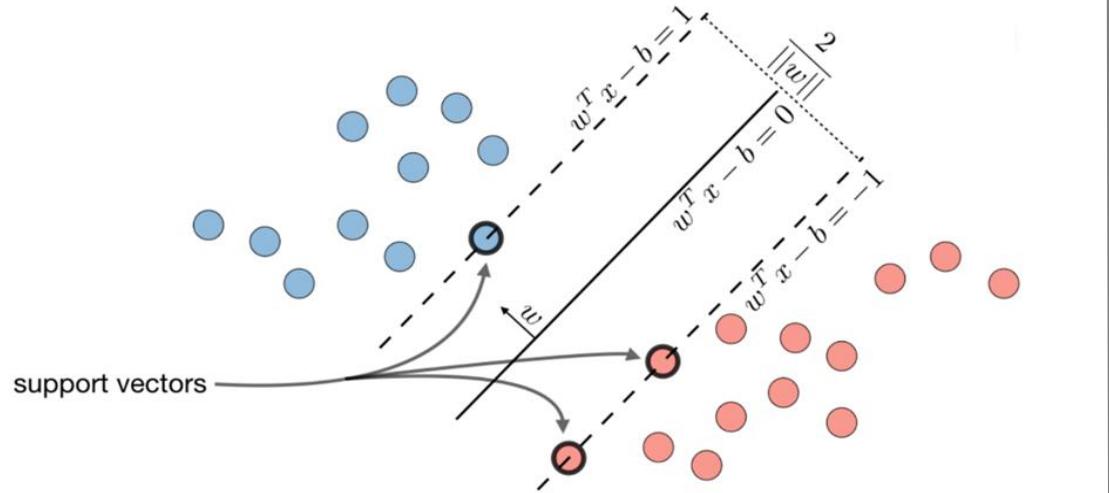
► Thesis proposition :



# Method

- ML: Rational and significance
- Research methodology

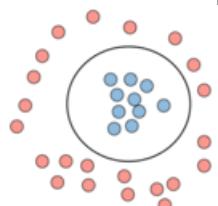
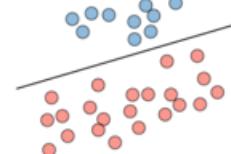
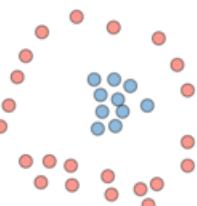
- Generalized model
- Adaptable to varying data population; quality and quantity
- Effective in identifying complex patterns in data
- High accuracy and high computational efficiency



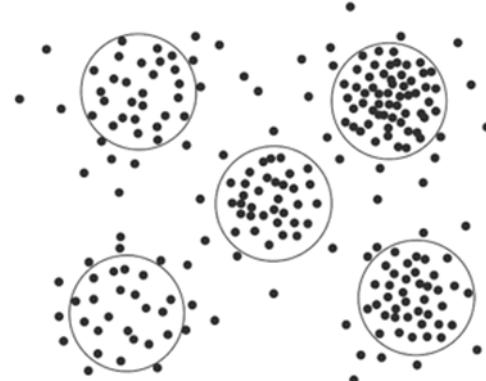
Finds the optimal hyperplane that maximizes/approximates Hard and soft **margins**

RBF (Kernel)

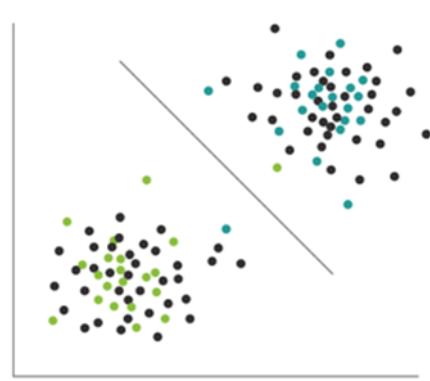
Higher dimensional



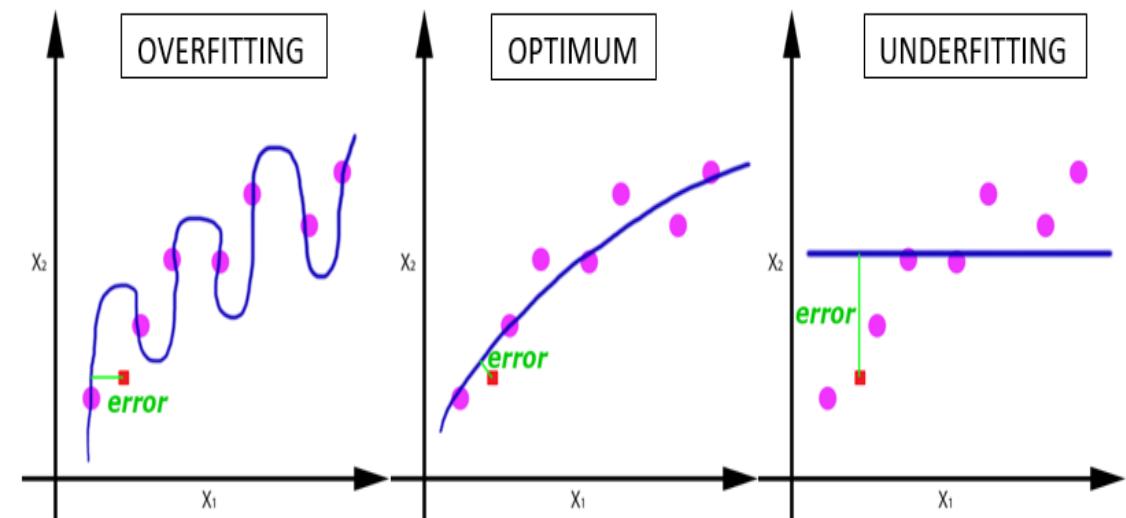
## Unsupervised



## Supervised



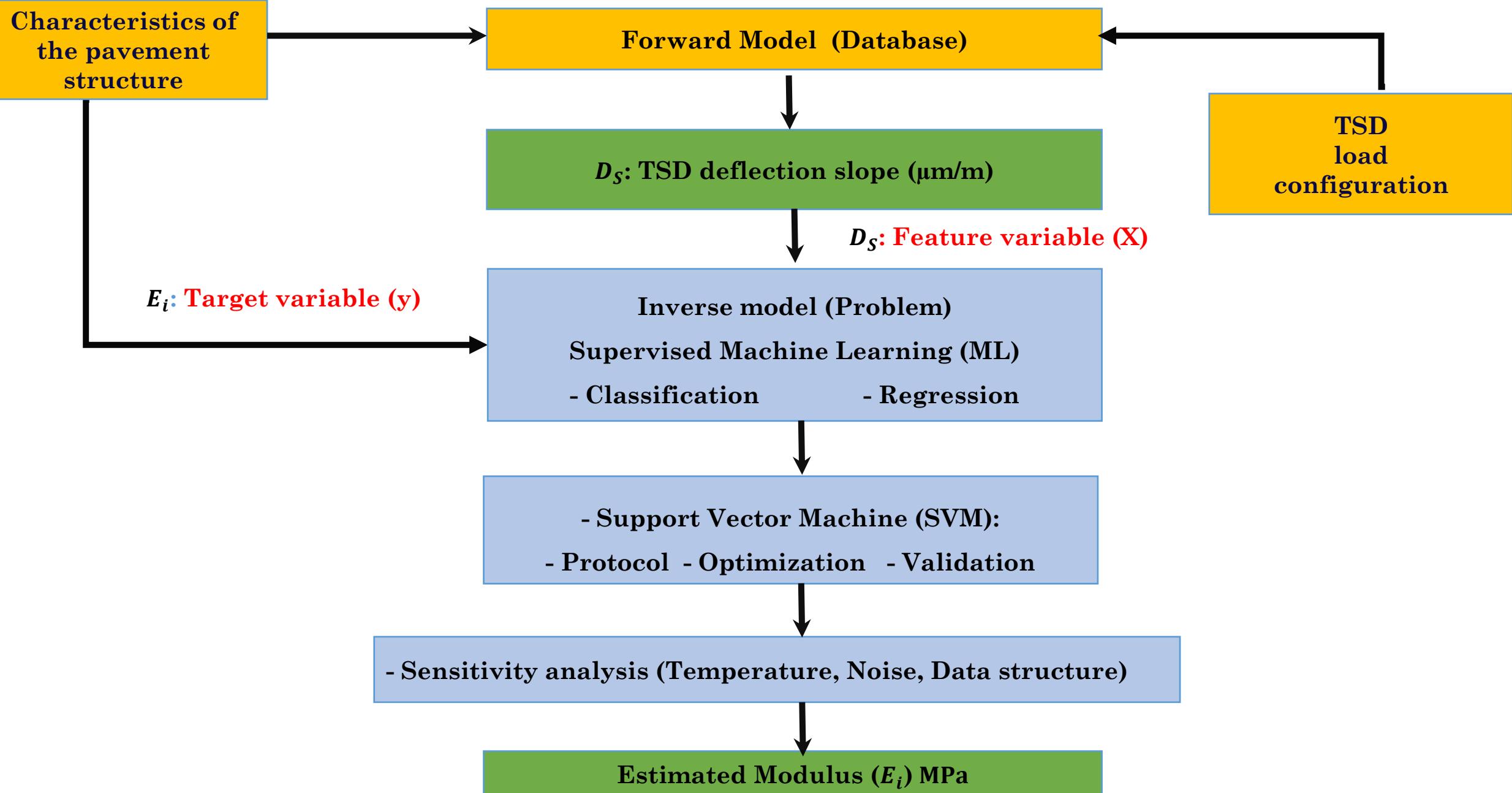
SVM GridSearchCV Scikit-learn hyperparameter tuning



# Method

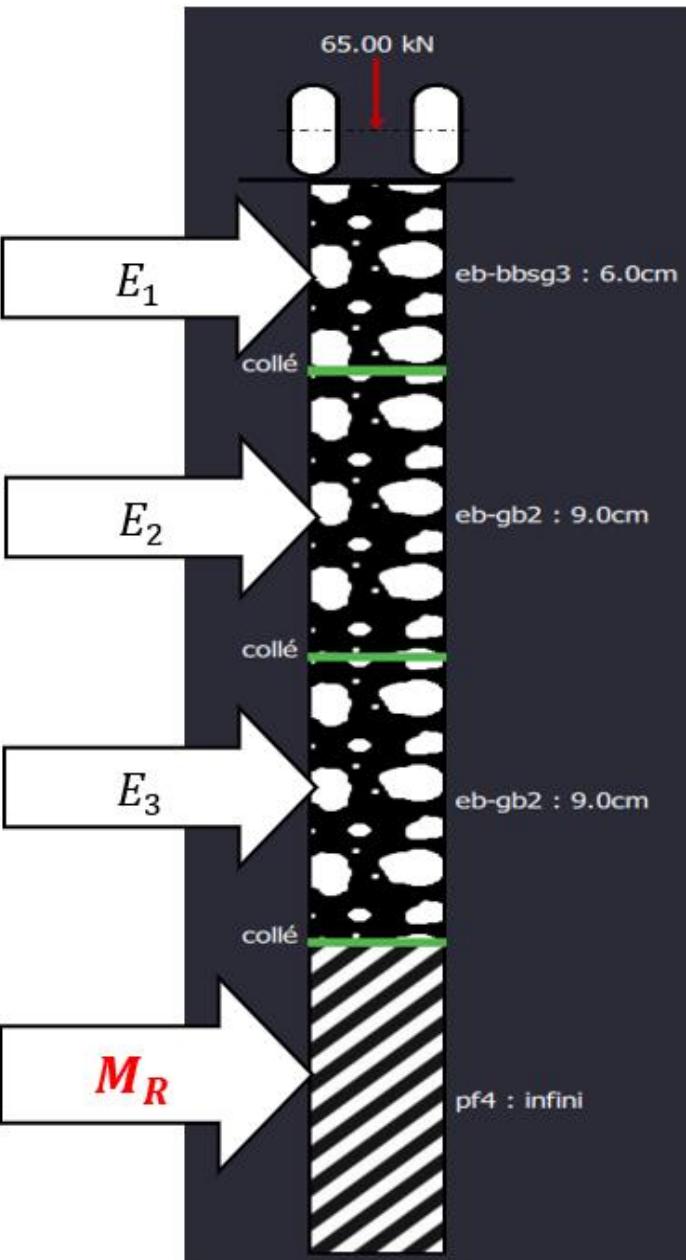
## AI-Model

## Methodology



# Forward Model

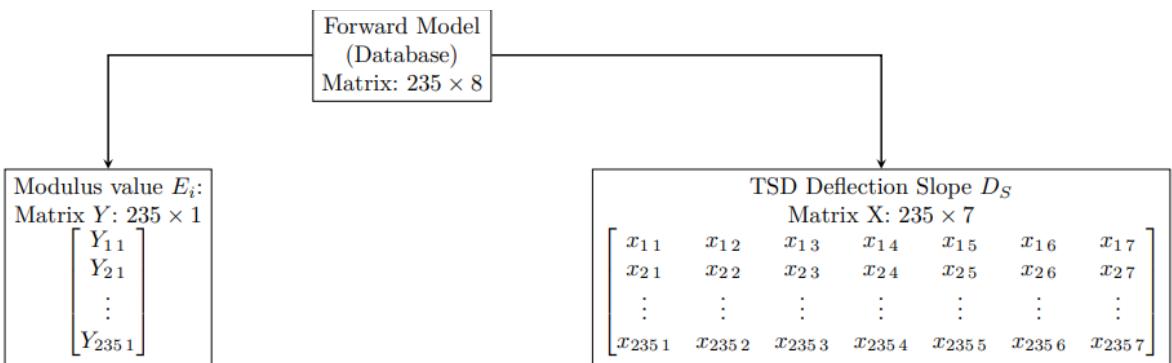
- Numerical simulation
- Synthetic database



- ▶ Alize-LCPC: Trademark developed by the Univ. Gustave Eiffel
- ▶ French industry standard for pavement modeling/design for +30 years
- ▶ **Burmister** pavement model: Multi-layered, linear elastic, isotropic approach with infinite in-plane layers

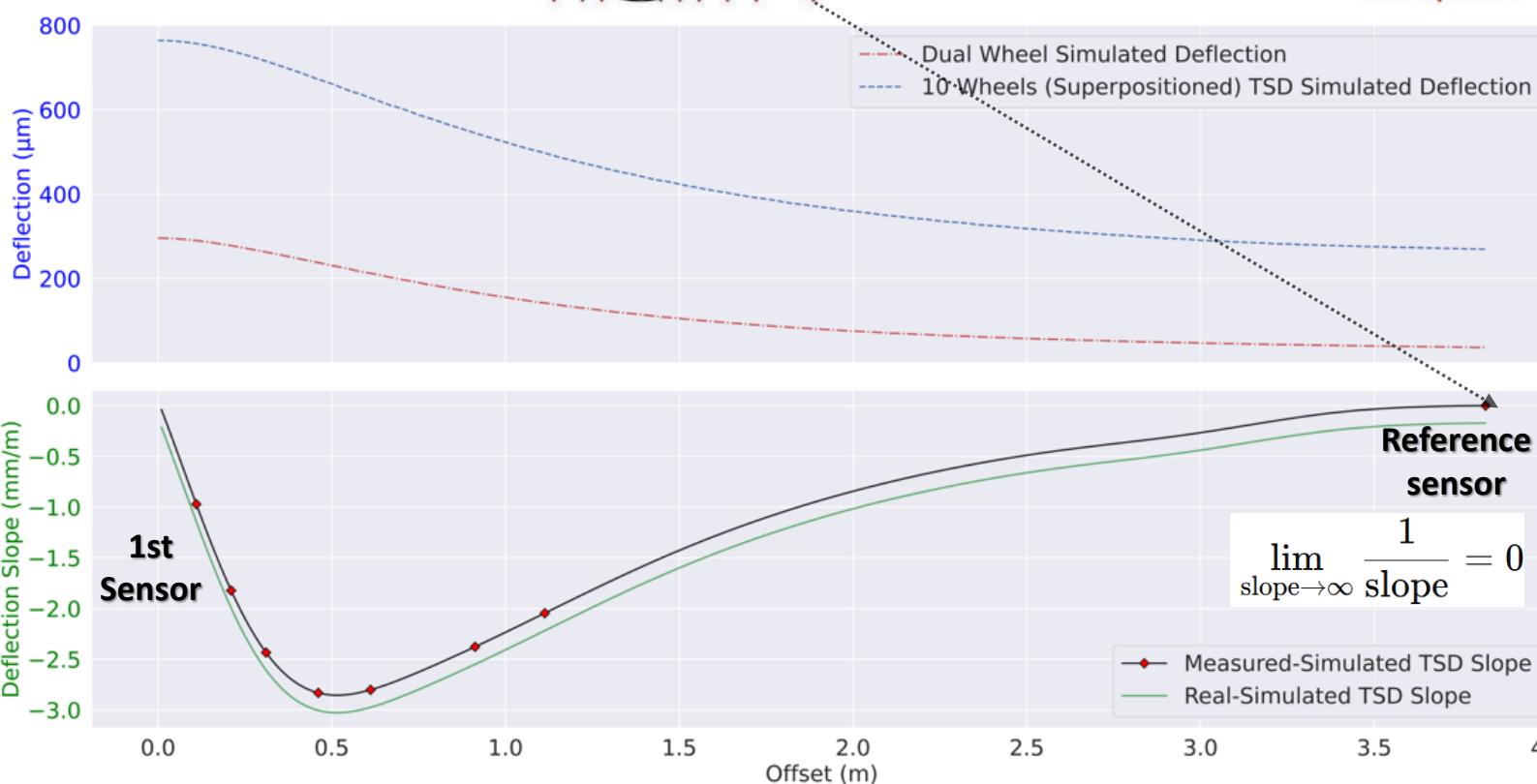


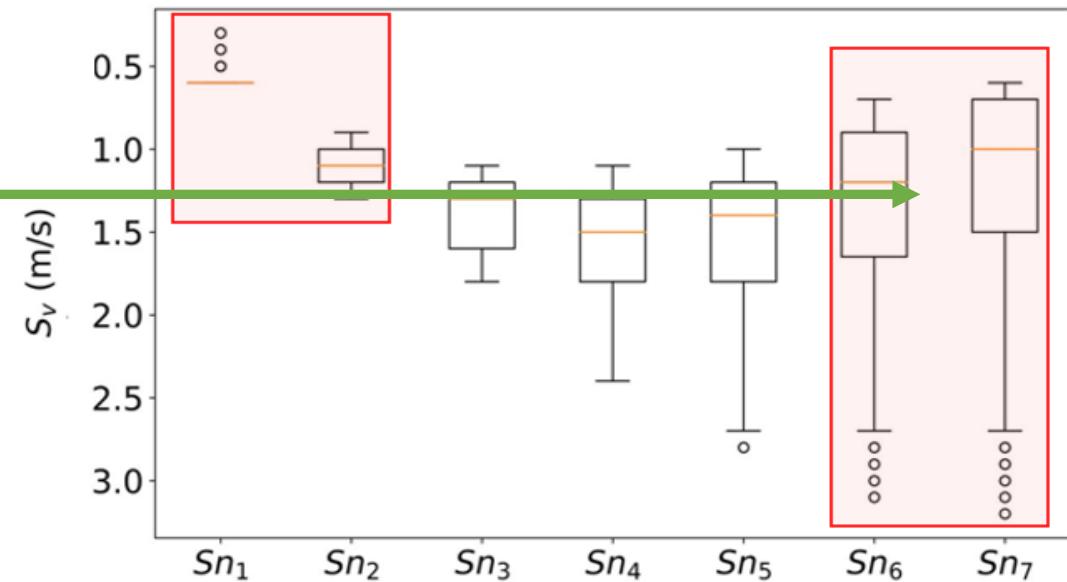
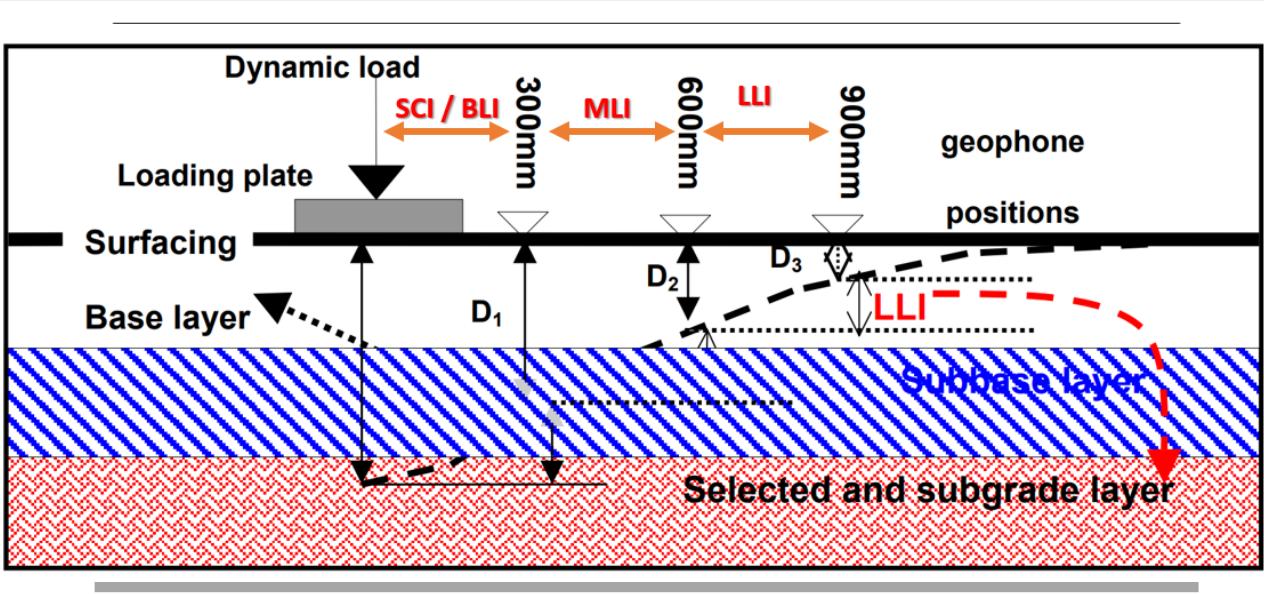
| Layer (i) | Thickness (m) | Material | $E_i$ (MPa) | $\nu$ | Interface | °C |
|-----------|---------------|----------|-------------|-------|-----------|----|
| 1         | 0.06          | HMA      | 7000        | 0.35  | (1) bound | 15 |
| 2         | 0.09          | BC-g2    | 9300        | 0.35  | (1) bound | 15 |
| 3         | 0.09          | BC-g2    | 9300        | 0.35  | (1) bound | 15 |
| 4         | $\infty$      | $M_R$    | 16-250      | 0.35  | (1) bound | 15 |



## TSD assumptions & observations

- ▶ **Loading mechanism:** Superposition
- ▶ **TSD Reference Sensor:** Uncertainty
- ▶ **Sensor Positioning:** Variability

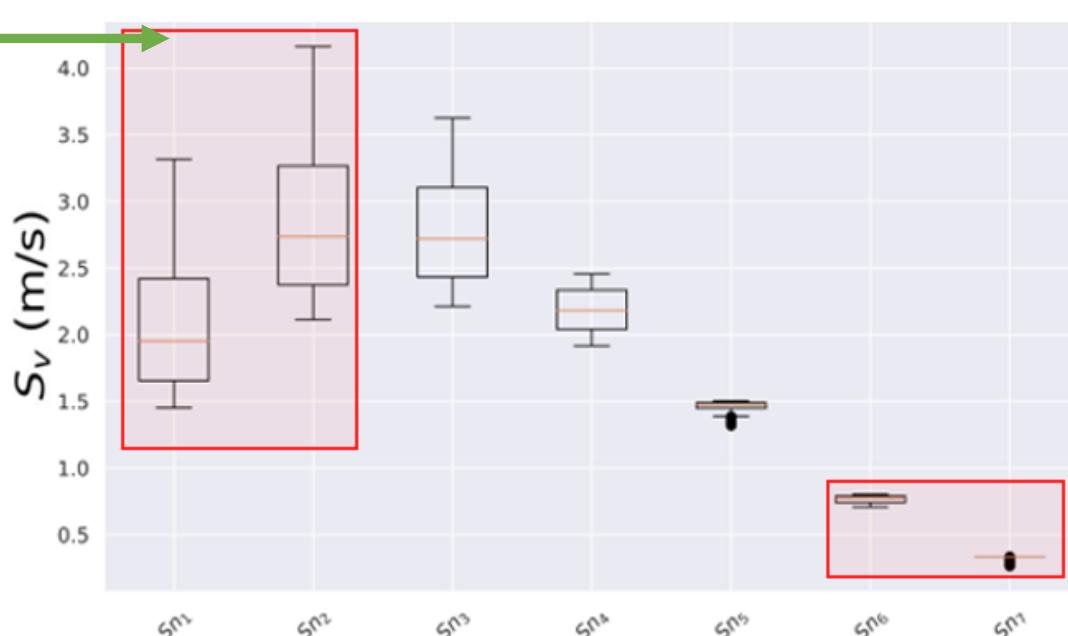




Deflection bowl parameter structural condition rating criteria for various pavement types

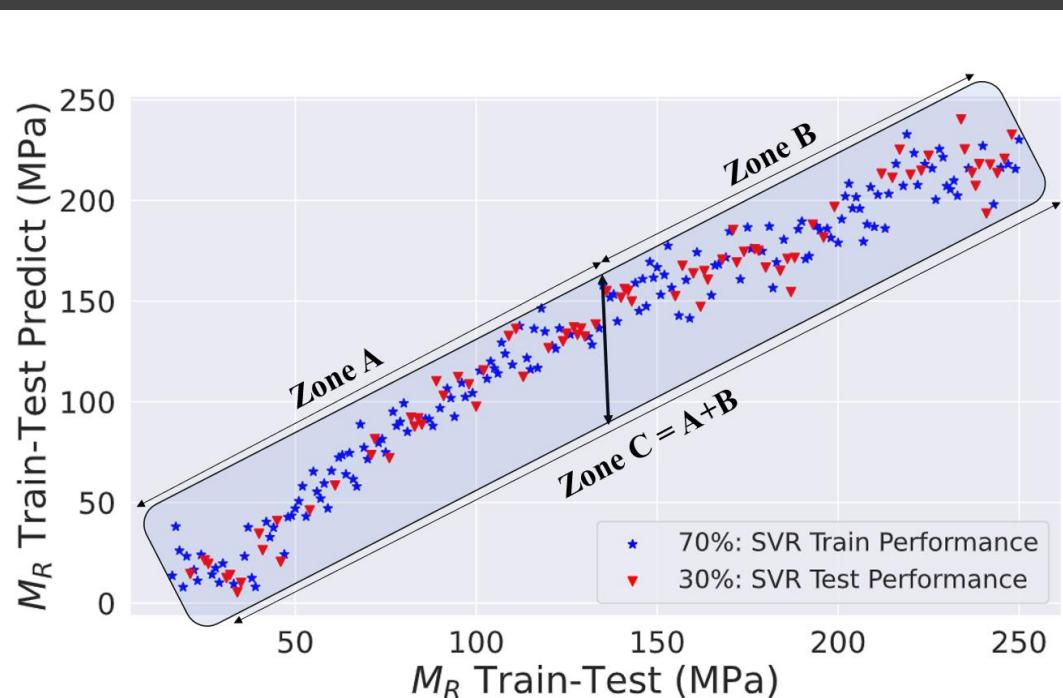
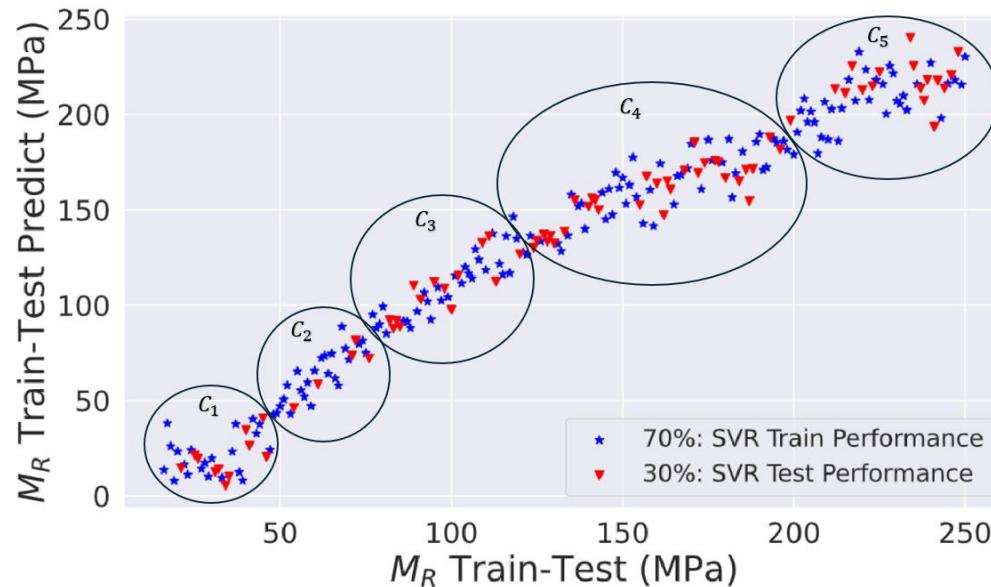
|                   | Structural condition rating | Deflection bowl parameters |         |                       |                       |                       |
|-------------------|-----------------------------|----------------------------|---------|-----------------------|-----------------------|-----------------------|
|                   |                             | $D_0$ ( $\mu\text{m}$ )    | RoC (m) | BLI ( $\mu\text{m}$ ) | MLI ( $\mu\text{m}$ ) | LLI ( $\mu\text{m}$ ) |
| Granular base     | Sound                       | <500                       | >100    | <200                  | <100                  | <50                   |
|                   | Warning                     | 500–750                    | 50–100  | 200–400               | 100–200               | 50–100                |
|                   | Severe                      | >750                       | <50     | >400                  | >200                  | >100                  |
| Cementitious base | Sound                       | <200                       | >150    | <100                  | <50                   | <40                   |
|                   | Warning                     | 200–400                    | 80–150  | 100–300               | 50–100                | 40–80                 |
|                   | Severe                      | >400                       | <80     | >300                  | >100                  | >80                   |
| Bituminous base   | Sound                       | <400                       | >250    | <200                  | <100                  | <50                   |
|                   | Warning                     | 400–600                    | 100–250 | 200–400               | 100–150               | 50–80                 |
|                   | Severe                      | >600                       | <100    | >400                  | >150                  | >80                   |

Note: These criteria can be adjusted to improve sensitivity of benchmarking.



# Inverse problem

- Modulus estimation observations
- Advanced parametric study (optimization)



| $M_R$ Value (MPa) | (16 - 49) | (50 - 79) | (80 - 119) | (120 - 199) | (200 - 250) |
|-------------------|-----------|-----------|------------|-------------|-------------|
| N° Class          | $C_1$     | $C_2$     | $C_3$      | $C_4$       | $C_5$       |

### (TTV) Framework:

- **Generalizability** and **Robustness** test of the model
- Stratified Train-Test-Split: 70% training, 30% testing
- Five iterations with 5/10 fold cross-validation for accuracy

Regression analysis: Training Testing Validation Framework

| Iteration | Training Zone | Testing Zone | Testing Objective   |
|-----------|---------------|--------------|---------------------|
| 1         | [C] (70%)     | [C] (30%)    | Overall Performance |
| 2         | [A]           | [B]          | Highest band        |
| 3         | [B]           | [A]          | Lowest band         |

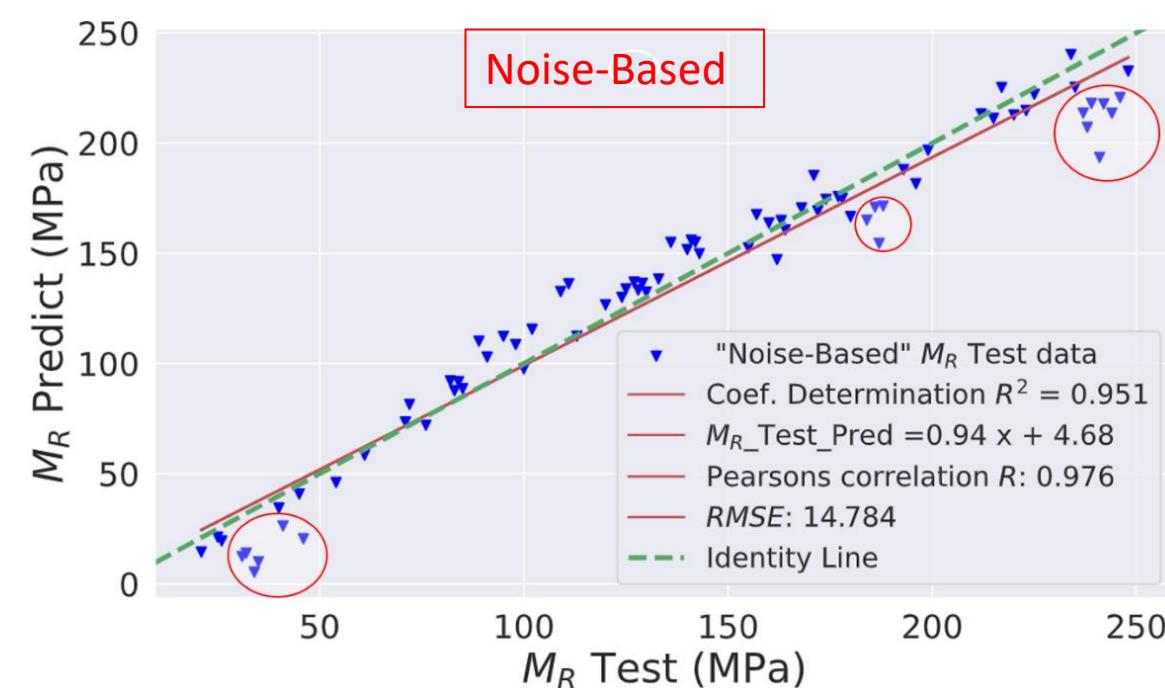
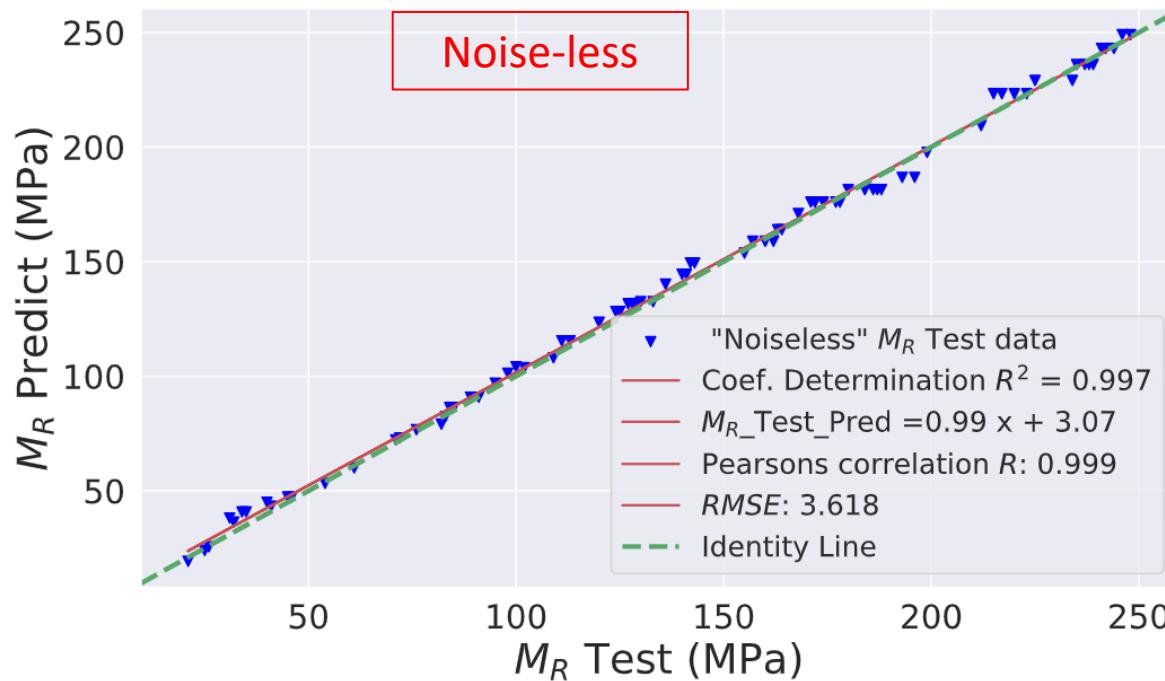
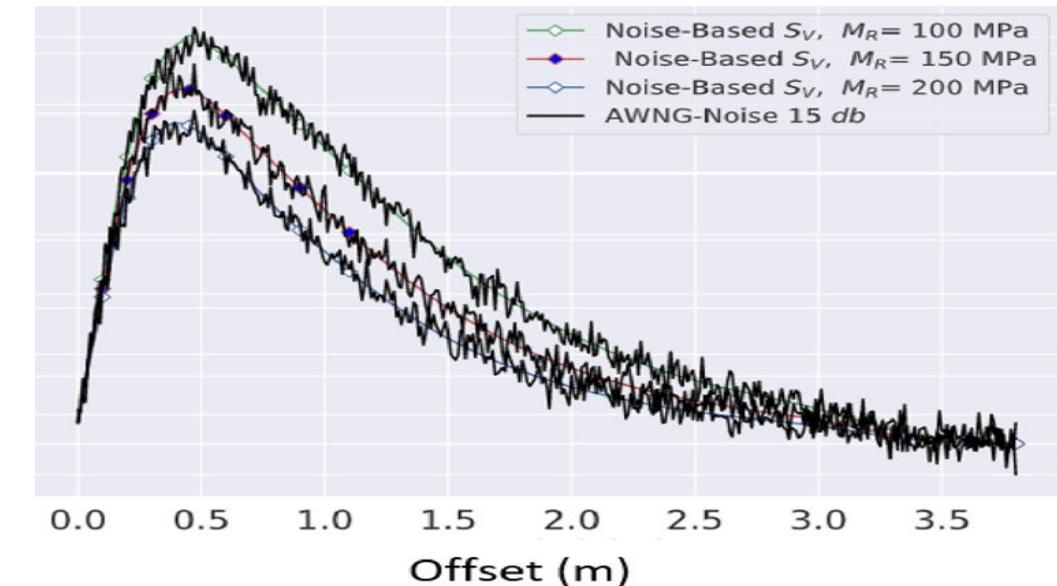
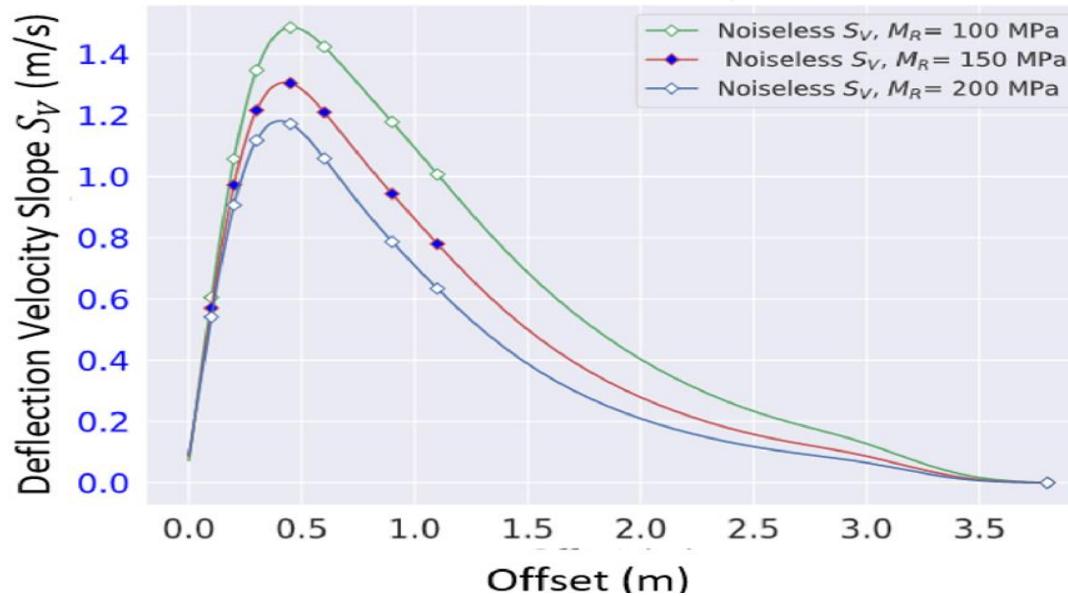
# Numerical: Inverse problem

Regression

Classification

Optimization

15/30



Calculation of evaluation metrics

| Metric         | Formula                                     | Description  |
|----------------|---|--|
| Precision      | $\frac{TP}{TP+FP}$                          | Proportion of correctly identified positive cases        |
| Recall         | $\frac{TP}{TP+FN}$                          | Proportion of actual positive cases identified correctly |
| DSC (F1 score) | $\frac{2 \times TP}{2 \times TP + FP + FN}$ | Harmonic mean of precision and recall                    |

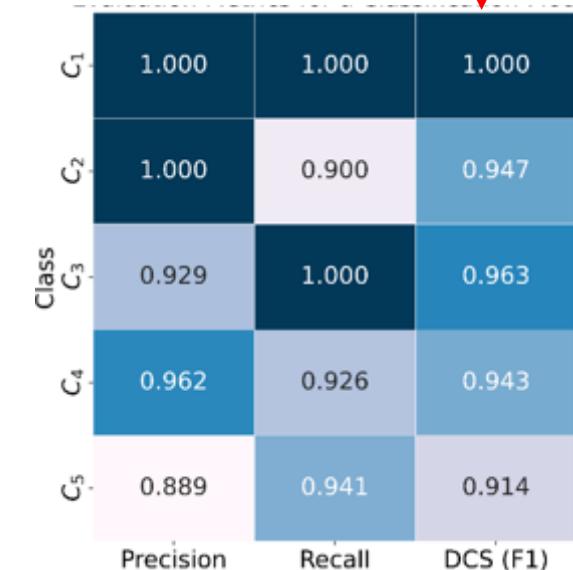
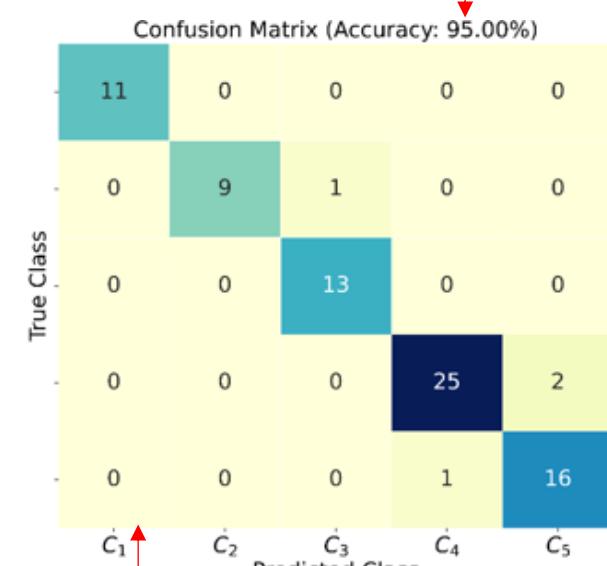
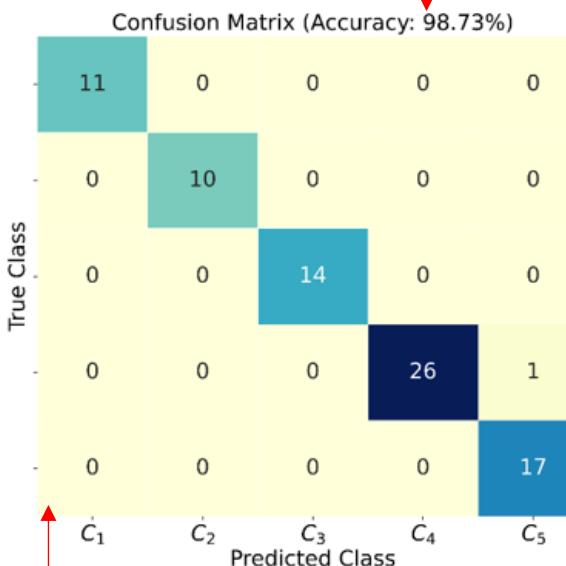
- Class (5) misclassifications : (High MPa)

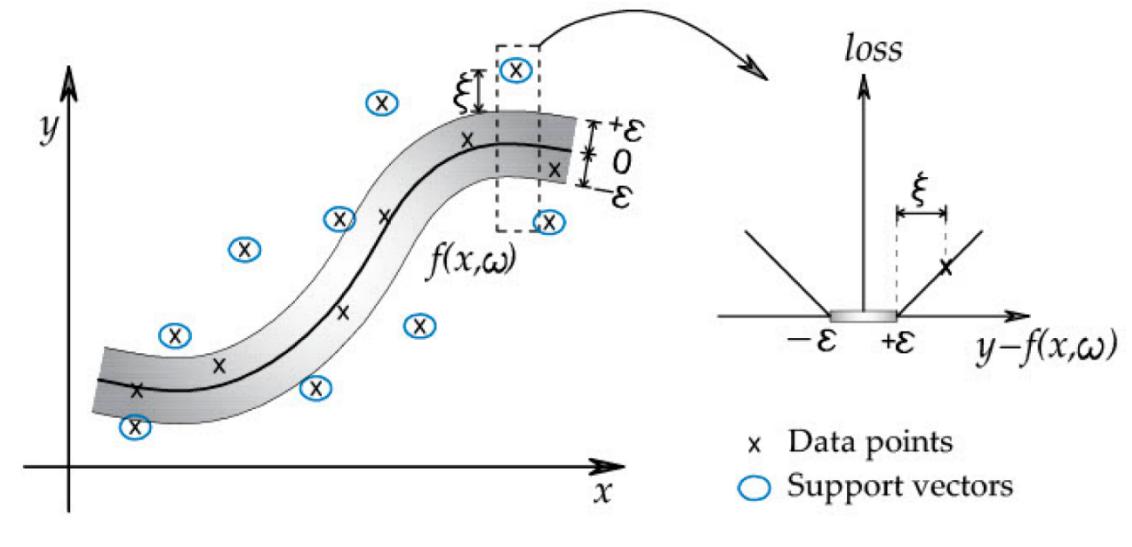
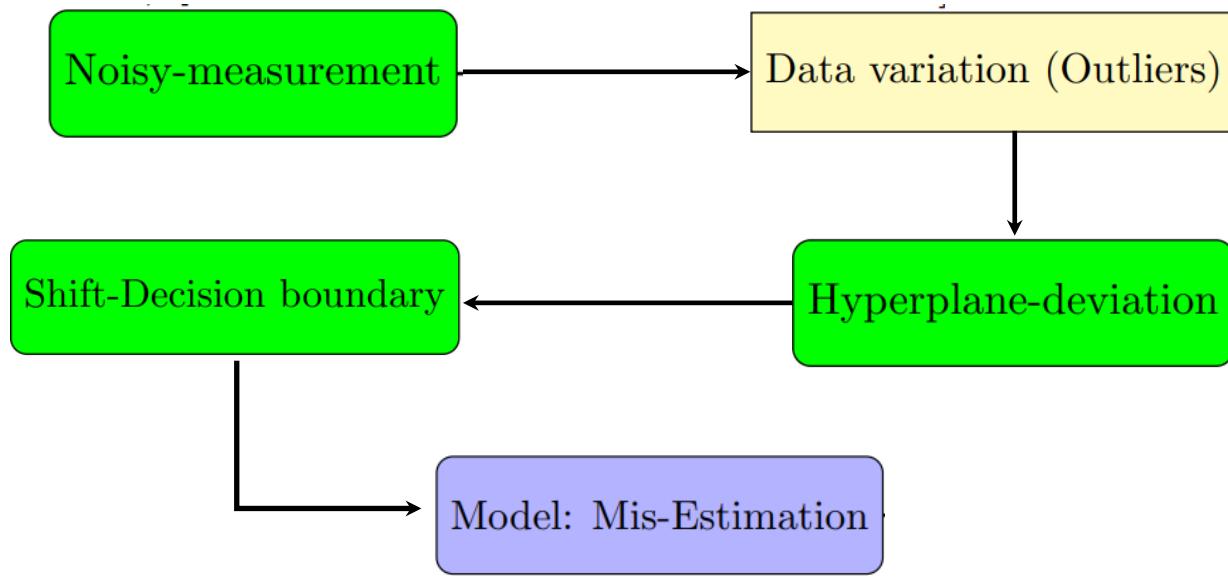
Relatively less critical

- Class (2) misclassifications: (Lower-MPa)

More critical

- Why the misclassification happened?





SVM GridSearchCV Scikit-learn hyperparameter tuning

| Parameter                     | Flexibility  | Accuracy     | Functionality                                  |
|-------------------------------|--------------|--------------|--|
|                               |              |              |  |
| $\uparrow C$                  | $\downarrow$ | $\uparrow$   | Regularization: Balance Over/underfitting      |
| $\uparrow \gamma$ (gamma)     | $\downarrow$ | $\uparrow$   | Control outliers impact on decision boundaries |
| $\uparrow \epsilon$ (epsilon) | $\uparrow$   | $\downarrow$ | Tolerance margin without error-penalty         |

| SVM | Parameter  | Initial Values Range    | Optimized | Initial Values Range    | Optimized |
|-----|------------|-------------------------|-----------|-------------------------|-----------|
|     |            |                         |           |                         |           |
|     | Kernal     | Linear/Poly/RBF/Sigmoid | RBF       | Linear/Poly/RBF/Sigmoid | RBF       |
|     | $C$        | [0.1 ; 1000]            | 100       | [0.1 ; 1000]            | 10        |
|     | $\gamma$   | [0.01 ; 100]            | 1         | [0.01 ; 100]            | 0.5       |
|     | $\epsilon$ | [0.001 ; 10]            | 0.01      | [0.001 ; 10]            | 0.2       |

# Sensitivity analysis

- Correlation analysis ( $\Theta \text{ } ^\circ\text{C} \propto E_i \propto D_S$ )
- Temperature uncertainty investigations

# Temperature sensitivity analysis

Forward Model

Inverse Problem

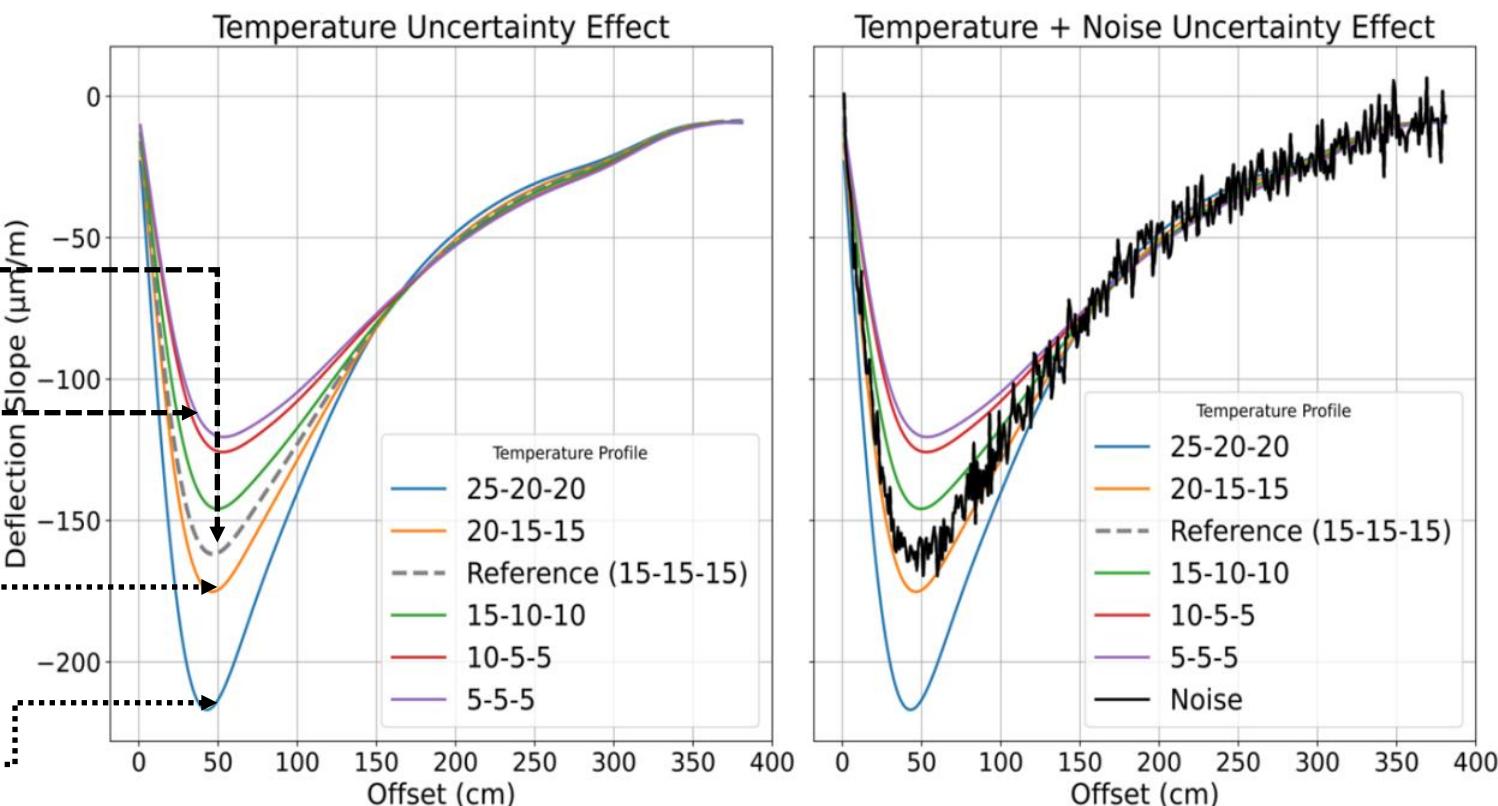
18/30

| Layer (i) | Thickness (m) | Material | $E_i$ (MPa) | $\nu$ | Interface  | °C |
|-----------|---------------|----------|-------------|-------|------------|----|
| 1         | 0.06          | HMA      | 7000        | 0.35  | (1) bound  | 15 |
| 2         | 0.09          | BC-g2    | 9300        | 0.35  | (1) bonded | 15 |
| 3         | 0.09          | BC-g2    | 9300        | 0.35  | (1) bonded | 15 |
| 4         | $\infty$      | $M_R$    | 16-250      | 0.35  | (1) bonded | 15 |

| Temp-Profil $\Theta$ °C |       |       | $E_i$ |       |       | $M_R$ |
|-------------------------|-------|-------|-------|-------|-------|-------|
| $L_1$                   | $L_2$ | $L_3$ | $E_1$ | $E_2$ | $E_3$ | 125   |
| 5                       | 5     | 5     | 11405 | 15090 | 15090 | 125   |
| 10                      | 5     | 5     | 9310  | 15090 | 15090 | 125   |
| 15                      | 10    | 10    | 7000  | 11880 | 11880 | 125   |
| 20                      | 15    | 15    | 4690  | 9300  | 9300  | 125   |
| 25                      | 20    | 20    | 3245  | 6120  | 6120  | 125   |

Correlation Analysis:  $\Theta$  °C  $\propto E_i \propto D_S$

- Temperature **uncertainty** influence
- Experimental representative cases



# Temperature sensitivity analysis

Forward Model

Inverse Problem

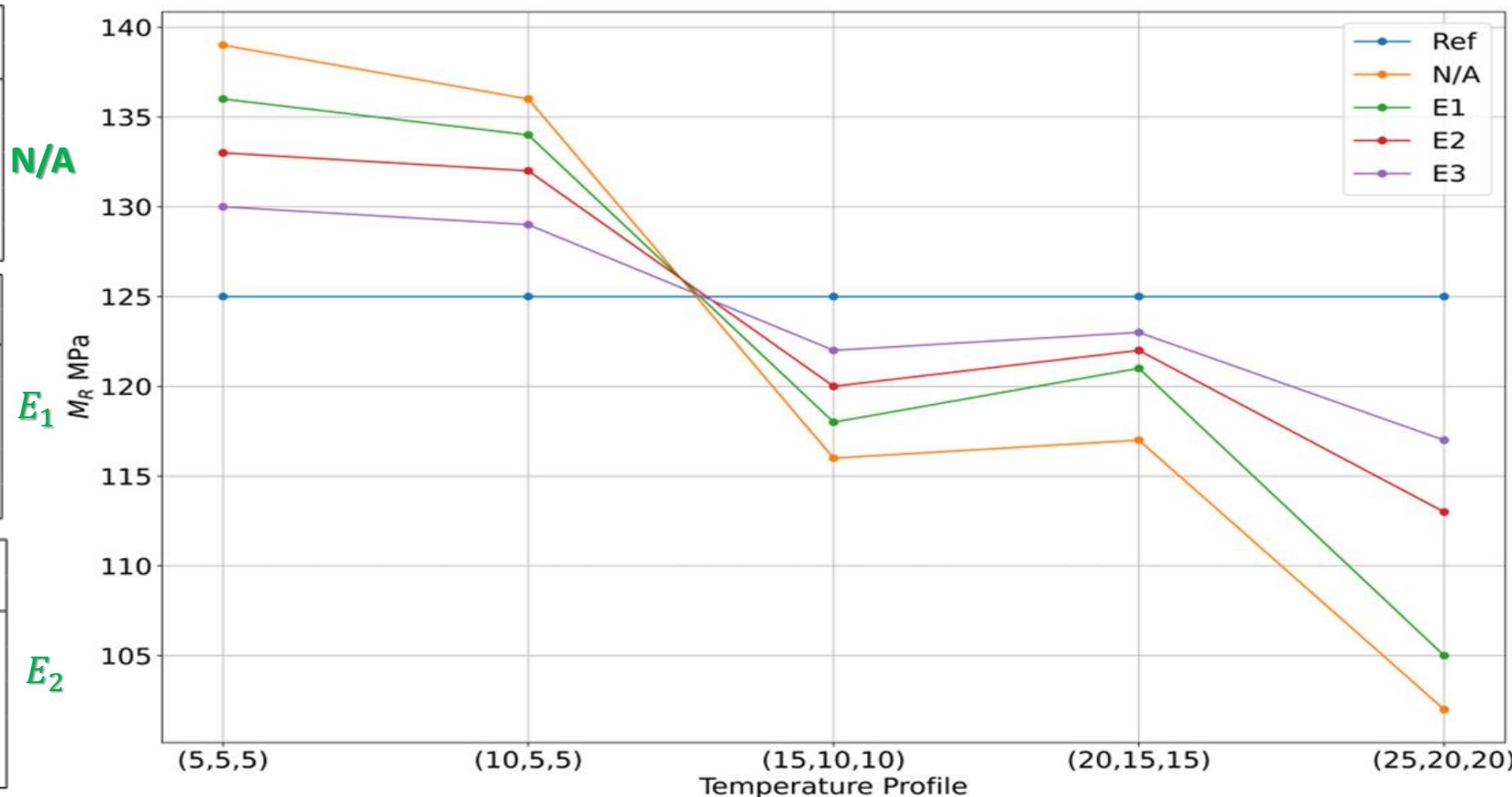
19/30

| $\Theta$ °C |       |       | $E_i$ (MPa) |       |       | $M_R$ | $\Delta$ |
|-------------|-------|-------|-------------|-------|-------|-------|----------|
| $L_1$       | $L_2$ | $L_3$ | $E_1$       | $E_2$ | $E_3$ | (MPa) | %        |
| -           | -     | -     | -           | -     | -     | 139   | 11.2%    |
| -           | -     | -     | -           | -     | -     | 136   | 8.8%     |
| -           | -     | -     | -           | -     | -     | 116   | 7.2%     |
| -           | -     | -     | -           | -     | -     | 117   | 6.4%     |
| -           | -     | -     | -           | -     | -     | 102   | 18.4%    |

| $\Theta$ °C |       |       | $E_i$ (MPa) |       |       | $M_R$ | $\Delta$ |
|-------------|-------|-------|-------------|-------|-------|-------|----------|
| $L_1$       | $L_2$ | $L_3$ | $E_1$       | $E_2$ | $E_3$ | (MPa) | %        |
| 5           | -     | -     | 11405       | -     | -     | 136   | 8.8%     |
| 10          | -     | -     | 9310        | -     | -     | 134   | 7.2%     |
| 15          | -     | -     | 7000        | -     | -     | 118   | 5.6%     |
| 20          | -     | -     | 4690        | -     | -     | 121   | 3.2%     |
| 25          | -     | -     | 3245        | -     | -     | 105   | 16%      |

| $\Theta$ °C |       |       | $E_i$ (MPa) |       |       | $M_R$ | $\Delta$ |
|-------------|-------|-------|-------------|-------|-------|-------|----------|
| $L_1$       | $L_2$ | $L_3$ | $E_1$       | $E_2$ | $E_3$ | (MPa) | %        |
| -           | 5     | -     | -           | 15090 | -     | 133   | 6.4%     |
| -           | 5     | -     | -           | 15090 | -     | 132   | 5.6%     |
| -           | 10    | -     | -           | 11880 | -     | 120   | 4%       |
| -           | 15    | -     | -           | 9300  | -     | 122   | 2.4%     |
| -           | 20    | -     | -           | 6120  | -     | 113   | 9.6%     |

| $\Theta$ °C |       |       | $E_i$ (MPa) |       |       | $M_R$ | $\Delta$ |
|-------------|-------|-------|-------------|-------|-------|-------|----------|
| $L_1$       | $L_2$ | $L_3$ | $E_1$       | $E_2$ | $E_3$ | (MPa) | %        |
| -           | -     | 5     | -           | -     | 15090 | 130   | 4%       |
| -           | -     | 5     | -           | -     | 15090 | 129   | 3.2%     |
| -           | -     | 10    | -           | -     | 11880 | 122   | 2.4%     |
| -           | -     | 15    | -           | -     | 9300  | 123   | 1.6%     |
| -           | -     | 20    | -           | -     | 6120  | 117   | 6.4%     |



$E_3$

► Layer effects:  
Knowledge of known pavement properties (temperature), enhances soil modulus estimation

# Experimental validation

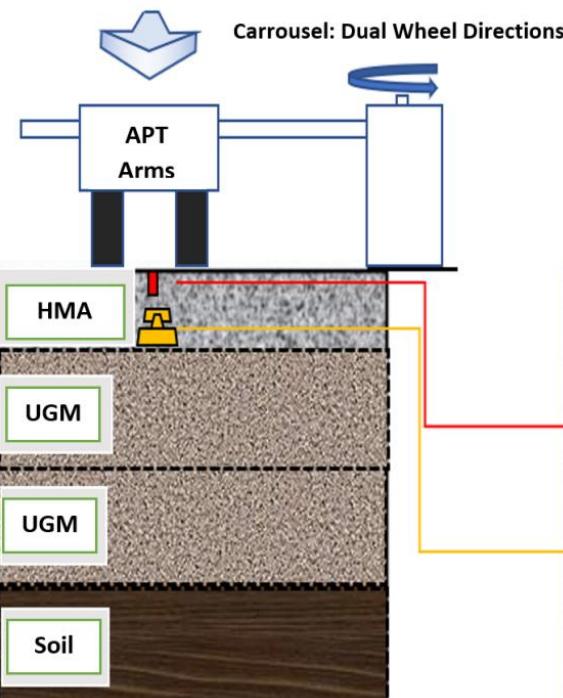
- Data acquisition & preprocessing
- Feature engineering (Global to Local)



| $L_i$ | $h$ (m)  | Material | $E_i$ (MPa)  | $\nu$ | Interface | (°C)   |
|-------|----------|----------|--------------|-------|-----------|--------|
| 1     | 0.10     | HMA      | 2000 - 10000 | 0.35  | (1) bound | 5 - 35 |
| 2     | 0.05     | UGM      | 126          | 0.35  | (1) bound | 5 - 35 |
| 3     | 0.25     | UGM      | 126          | 0.35  | (1) bound | 5 - 35 |
| 4     | $\infty$ | $M_R$    | 104          | 0.35  | (1) bound | 5 - 35 |

## Data Acquisition

- Location: Univ Gustave Eiffel, fatigue carousel
- Testing method: Accelerated Pavement Test (APT)
- Project: I-Street project
- Instruments: Geophone and Strain Gauge
- Purpose: Simulate real-world traffic conditions
- Bottom line: Collect fatigue data for pavement analysis



Geophone Sensor, specifically Model GS-ONE LF from Geospace Technology

| Model         | GS-ONE LF       |
|---------------|-----------------|
| Orientation : | Vertical        |
| Drawing No. : | CS 480-00360-01 |
| Revision :    | B               |
| Date :        | 12/01/16        |



H-shaped DYNATEST® PAST-IIA strain sensor

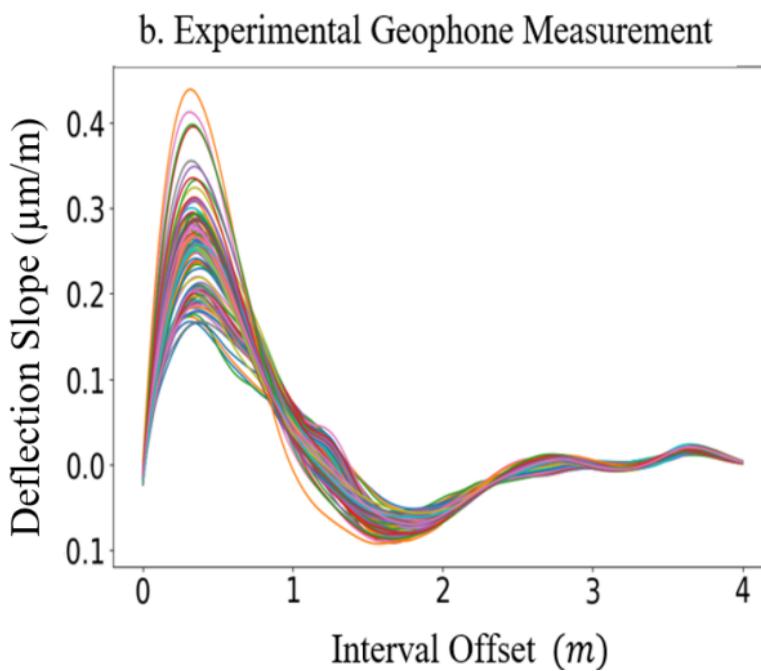
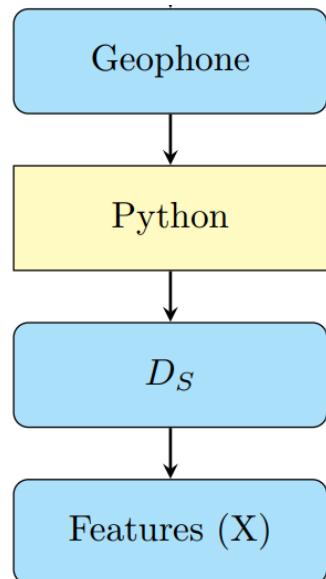
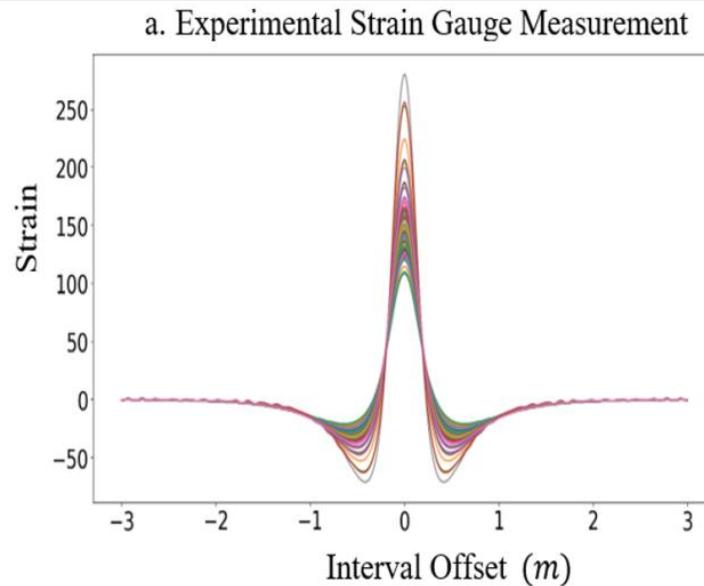
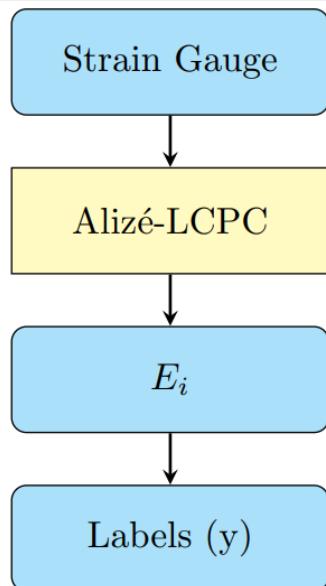
# Experimental validation

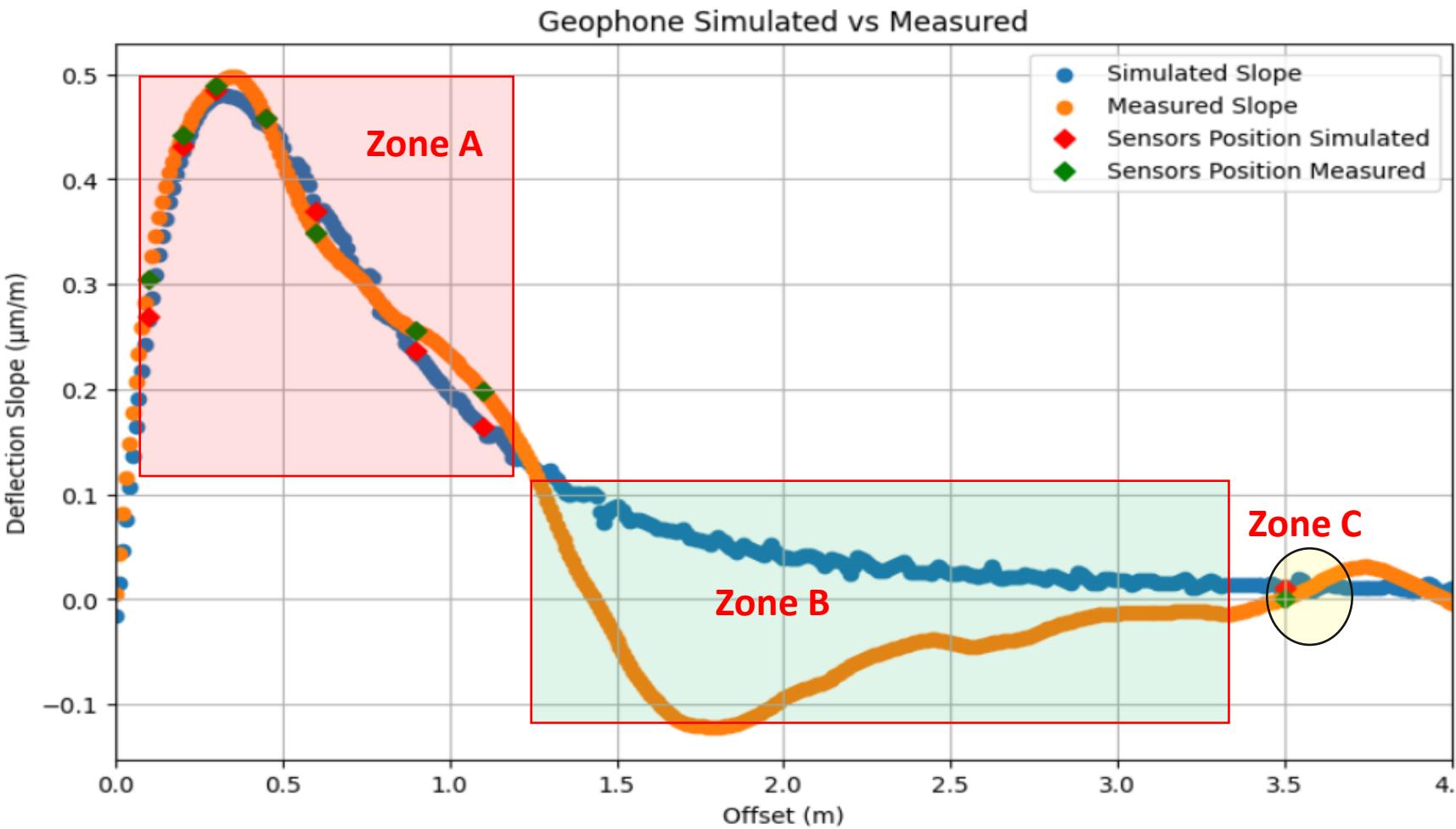
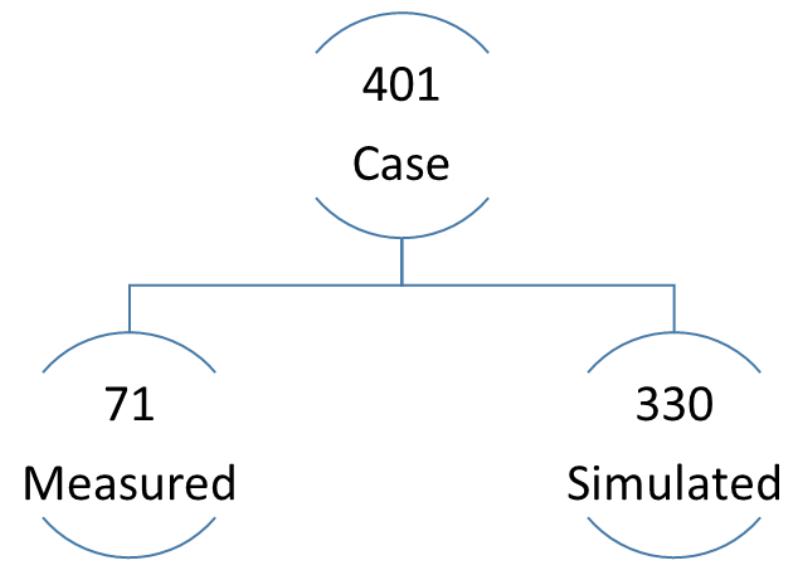
DataSet

Data Augmentation

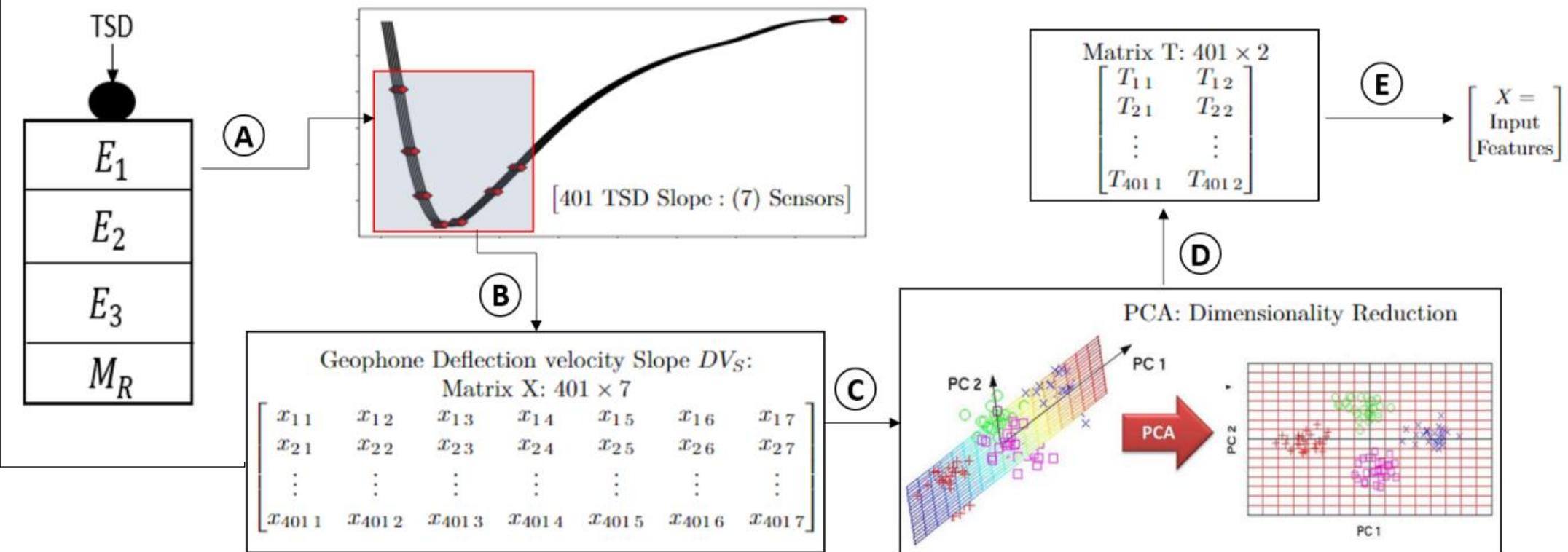
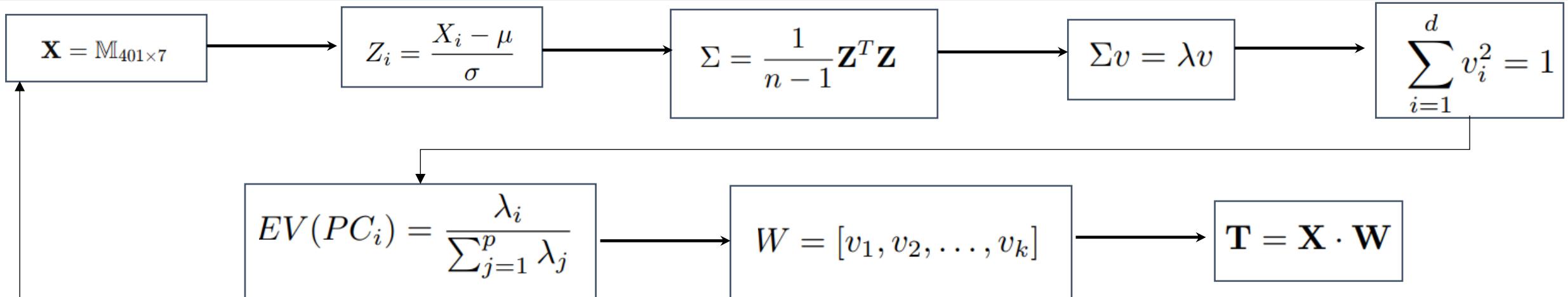
Dimensionality Reduction

21/30





$$\lim_{\text{slope} \rightarrow \infty} \frac{1}{\text{slope}} = 0$$

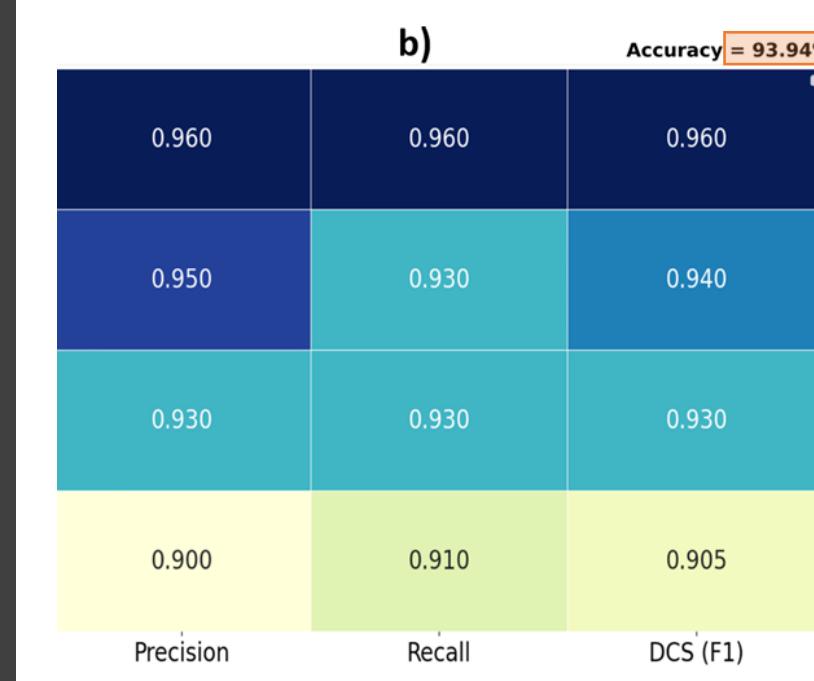
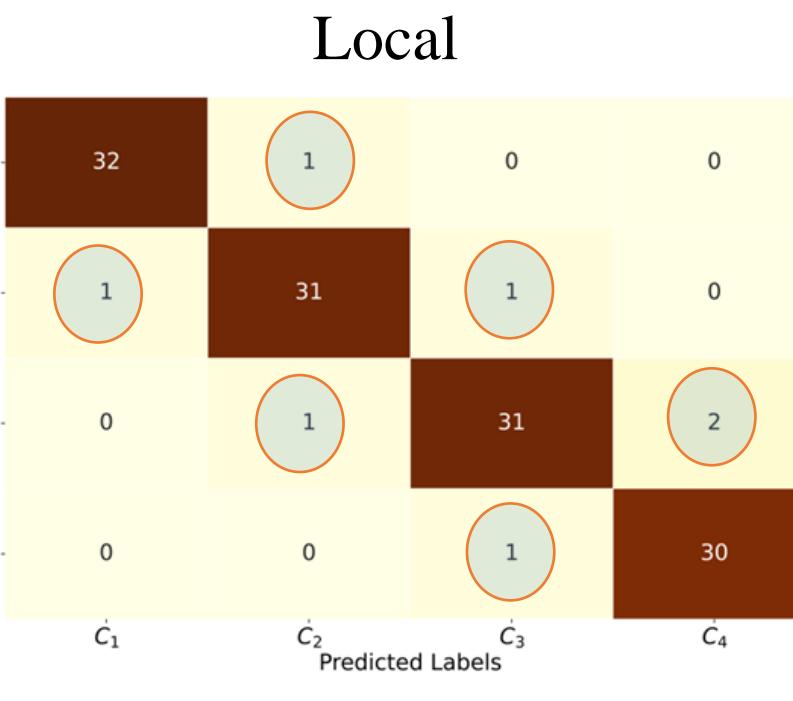
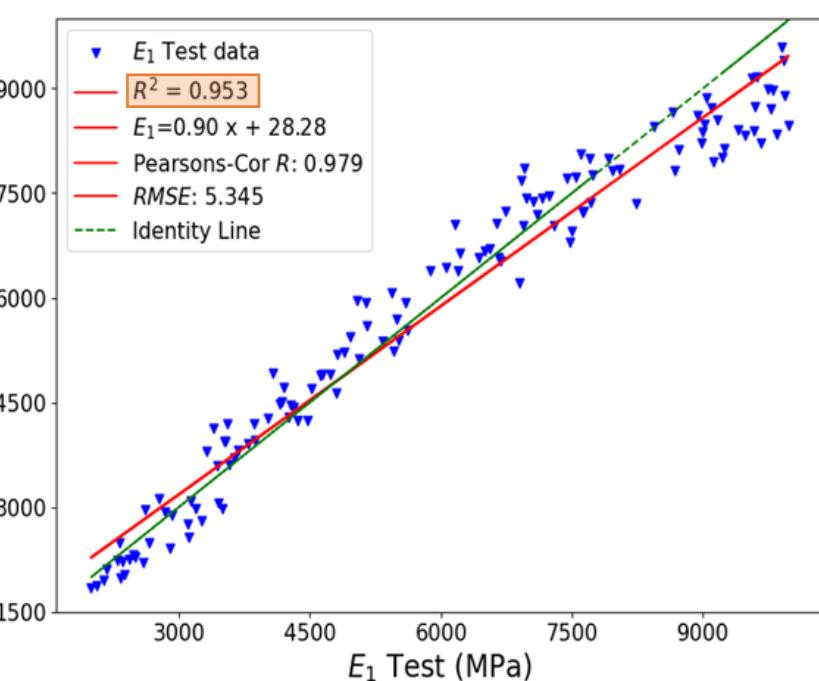
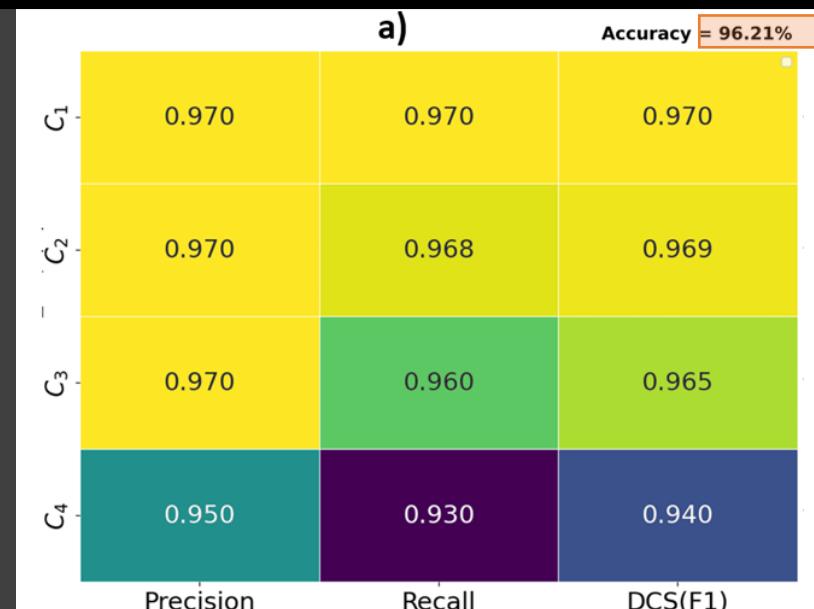
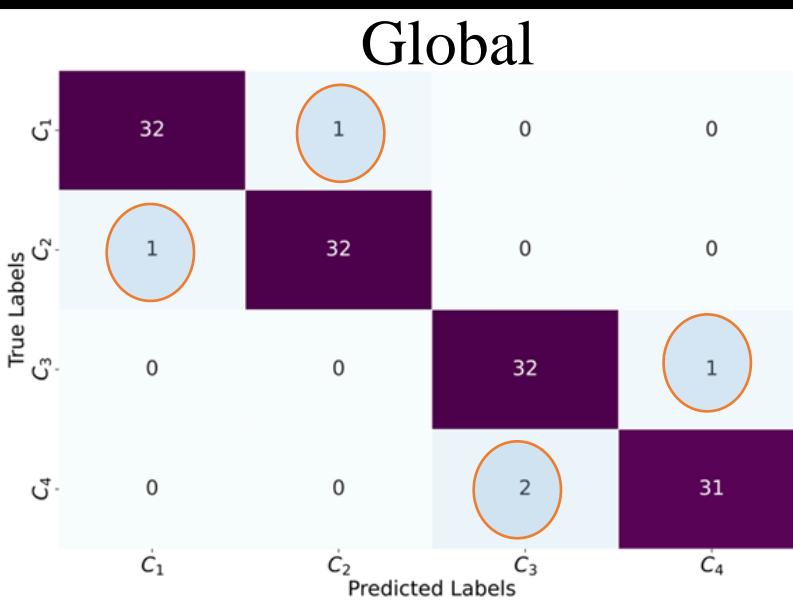
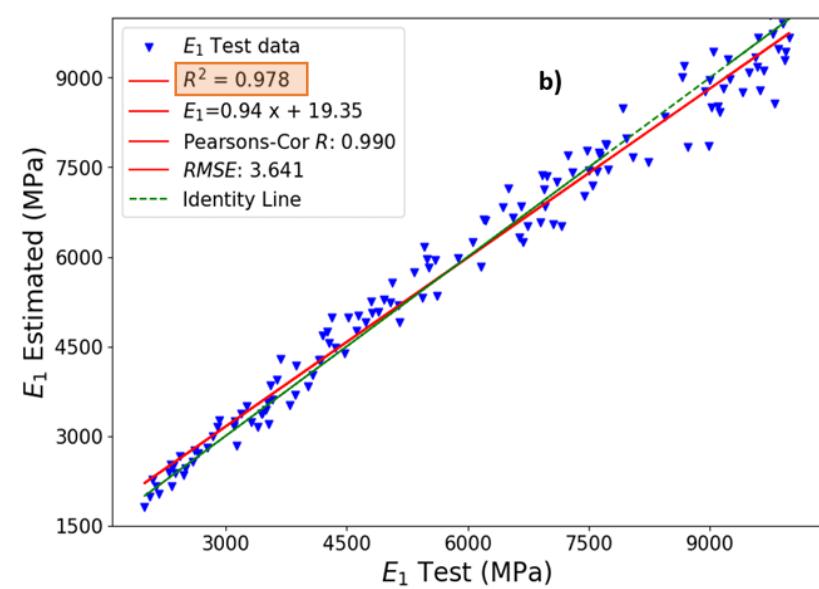


# Experimental validation

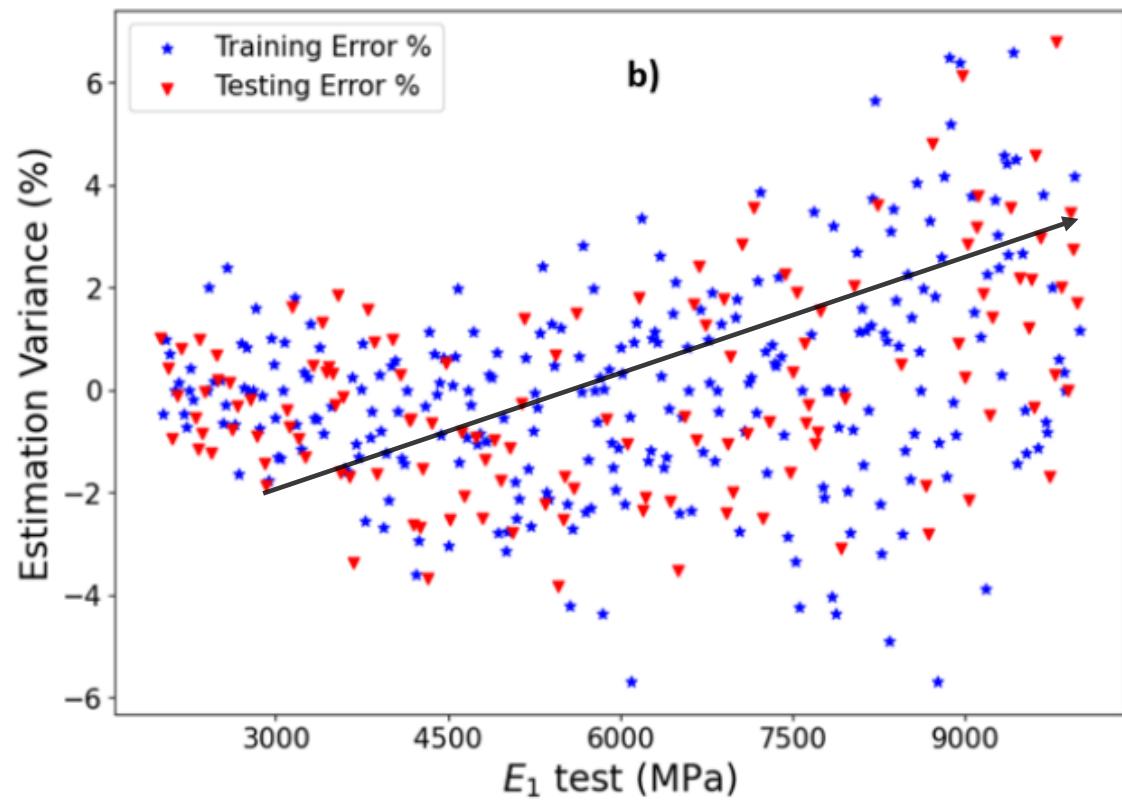
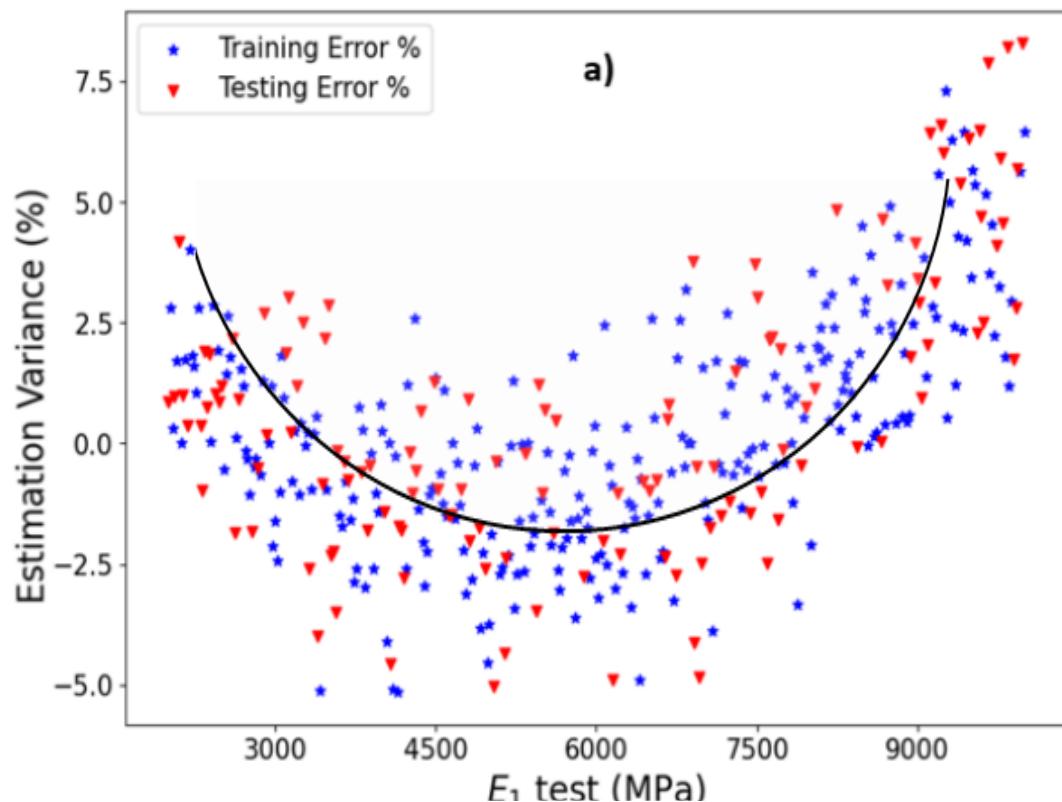
Case Study

Data Augmentation

Dimensionality Reduction



## Variance analysis a) Local b) Global



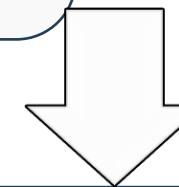
| Factor                          | Local (PCA)                       | Global                    |
|---------------------------------|-----------------------------------|---------------------------|
| Accuracy                        | Slightly ↓                        | ↑                         |
| Data Efficiency                 | ↑ (faster processing)             | ↓ (computation-intensive) |
| Generalizability (limited data) | Potentially ↑                     | Potentially ↓             |
| Strength                        | Balancing accuracy and efficiency | Mitigate uncertainty      |

# Conclusion

- Challenges and limitations
- Perspective (future Work)
- Valorization & outreach

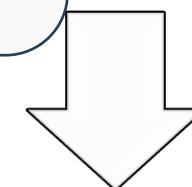
► **Numerical Forward Model :**

SCI, MLI, LLI, and are key pavement indices, incorporating other sensors with proportional significance is essential for a more accurate structural assessment



► **Numerical Inverse problem :**

- I. ML effectively estimates soil modulus from the deflection slope
- II. Noise impacts measurement accuracy, necessitating noise reduction techniques
- III. Incorporating temperature thermal model leads to enhancing modulus estimation



► **Experimental validation :**

- I. Experimental approach validates the accuracy and reliability of the numerical model
- II. Local approach balances accuracy, while offering computational efficiency

## Research limitations and challenges

- Limited labeled data for supervised learning
- Applicability across diverse environments (Generalizability)

## Future work :

- Validate with real TSDDs measurement data
- Explore TSD applications for concrete and composite pavements
- Vertical temperature analysis and thermal modeling
- Integrate TSD and Ground Penetrating Radar (GPR) data for (Thickness, Interface...)

# General conclusion

Conclusion

Limitation

Perspective

Valorization

## Journal articles



## Career perspective



R&D Project Manager

## International conferences



11th Young Researchers Seminar  
May 15-17, 2023 – Lisbon, Portugal

Supporting organisations: ectri, FEHRL, FERSI, EURNEX, Humanist

## National conferences



## Scientific outreach (vulgarization)



Research visits



## LIST OF PUBLICATIONS

- [1] A. Abdelmuhsen et al. "ML-based Model for Numerical Pavement Elastic Modulus Estimation Using TSD Deflection Slope Data". In: *Discover Applied Sciences* (2024). DOI: SubmissionID:1dcf31a2-54f4-4780-820c-4174ec74bdde. URL: <https://submission.springernature.com>.
- [2] A. Abdelmuhsen et al. "On the variants of SVM method for the estimation of soil elastic modulus from TSD model: Numerical parametric study". In: *Transportation Engineering* 13 (2023), p. 100187. ISSN: 2666-691X. DOI: <https://doi.org/10.1016/j.treng.2023.100187>. URL: <https://www.sciencedirect.com/science/article/pii/S2666691X23000271>.
- [3] Abdelgader Abdelmuhsen et al. "Database Construction based upon Numerical Modelling of Pavement Deflection Behaviour under the Traffic Speed Deflectometer". In: *25e Congrès Français de Mécanique*. 2022.
- [4] Abdelgader Abdelmuhsen et al. "Machine Learning-based modelling for numerical evaluation of the traffic speed deflectometer performance". In: *YRS2023-11th Young Researchers Seminar*. Zenodo. 2023.
- [5] Abdelgader Abdelmuhsen et al. "ML-Based Model for the Estimation of the Pavement Elastic Modulus via Deflection". In: *Transport Research Arena (TRA)*. 2024.
- [6] Abdelgader Abdelmuhsen et al. *Numerical modeling of pavement deflection behavior under The Traffic Speed Deflectometer*. Version V1. 2024. DOI: [10.57745/QNUCI6](https://doi.org/10.57745/QNUCI6). URL: <https://doi.org/10.57745/QNUCI6>.

# Annex

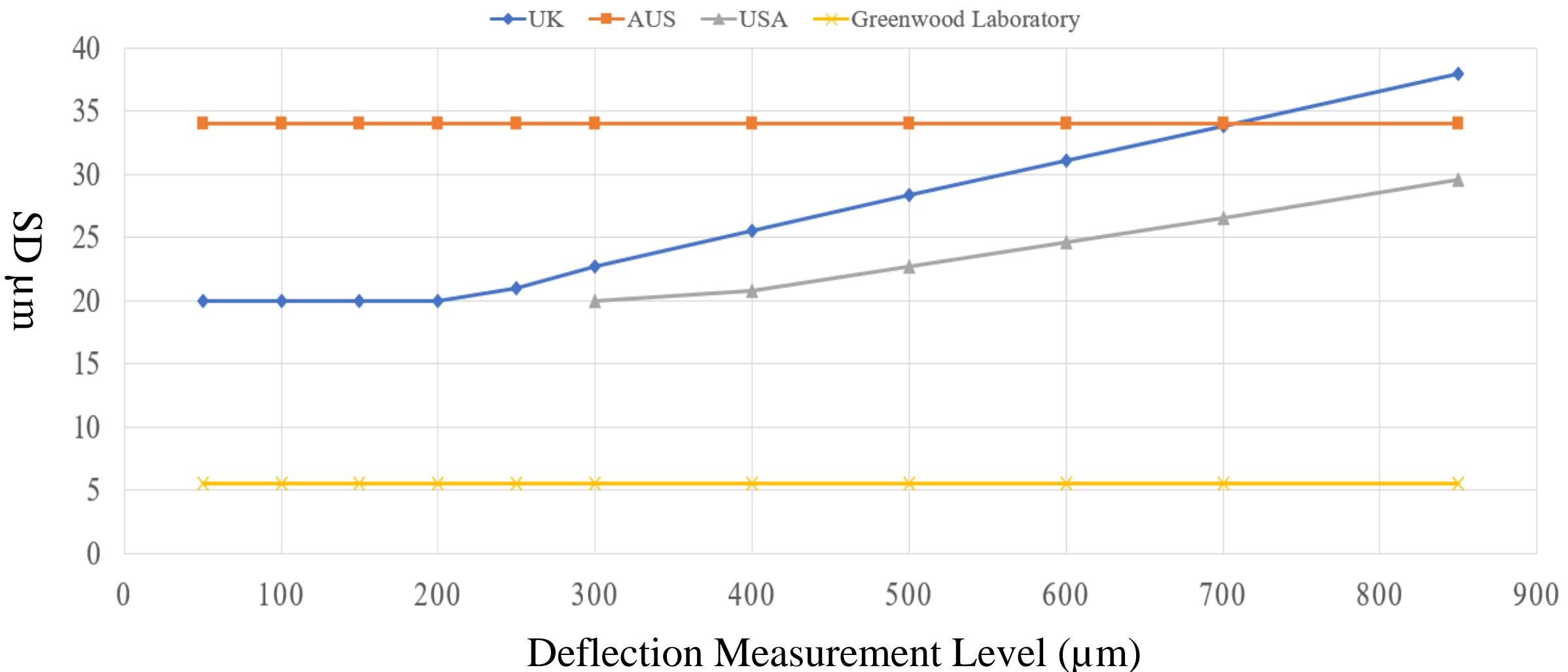
- Repeatability test (Noise level)
- Geophone Signal Processing
- Principle Components Analysis

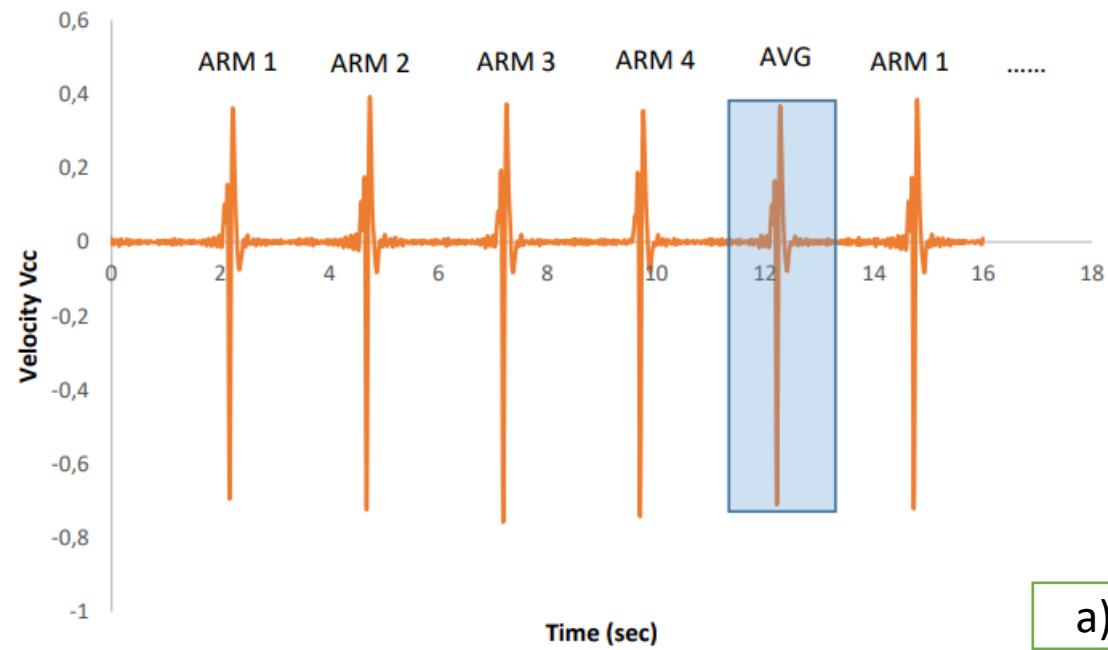
$$\text{SNR} = 10 \log_{10}(\text{SD})^2$$

$$20 \cdot \log_{10}(5.5) \approx 20 \cdot 0.740$$

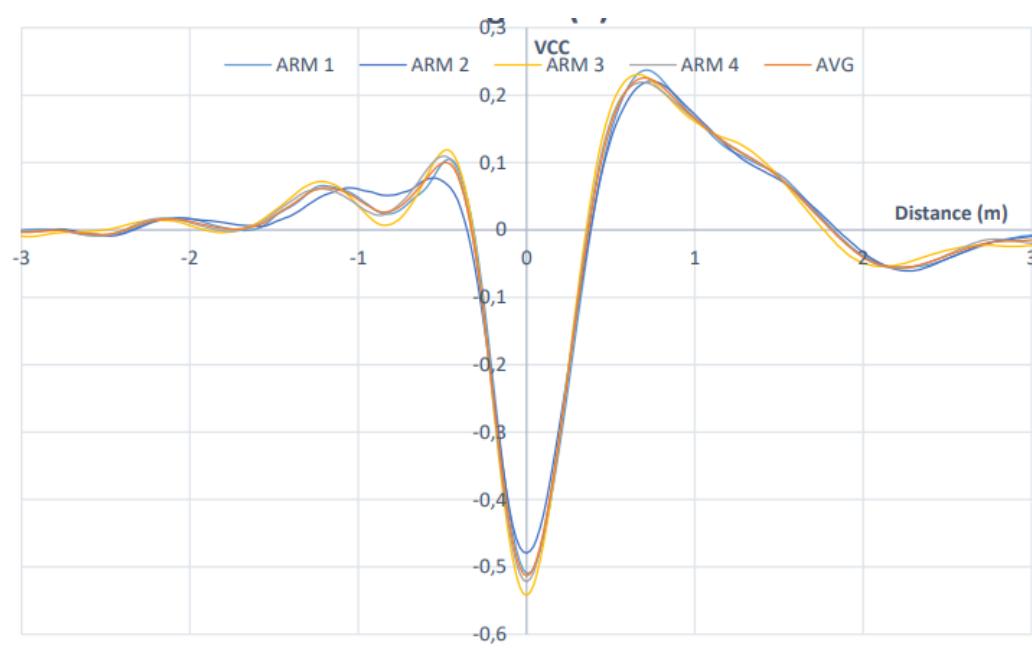
$$\approx 14.80 \approx 15 \text{ dB}$$

AWGN

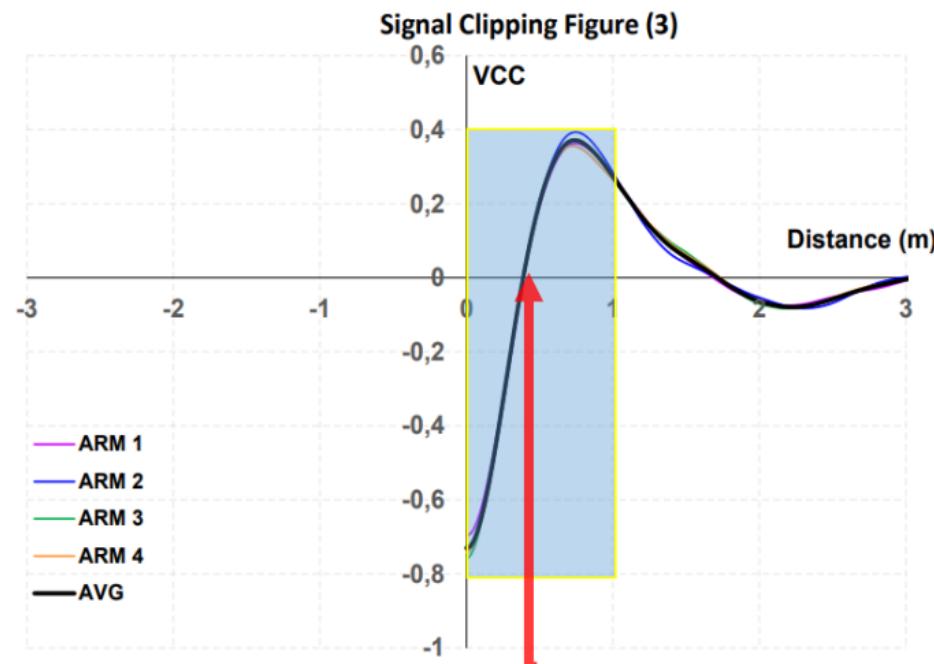




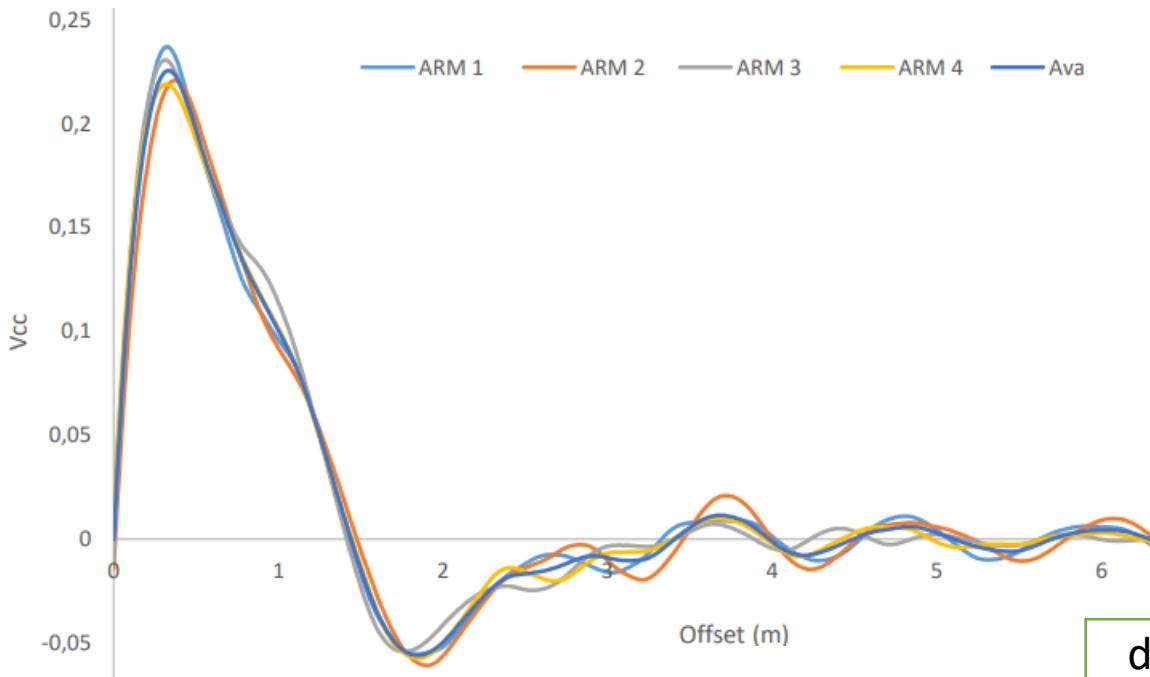
a)



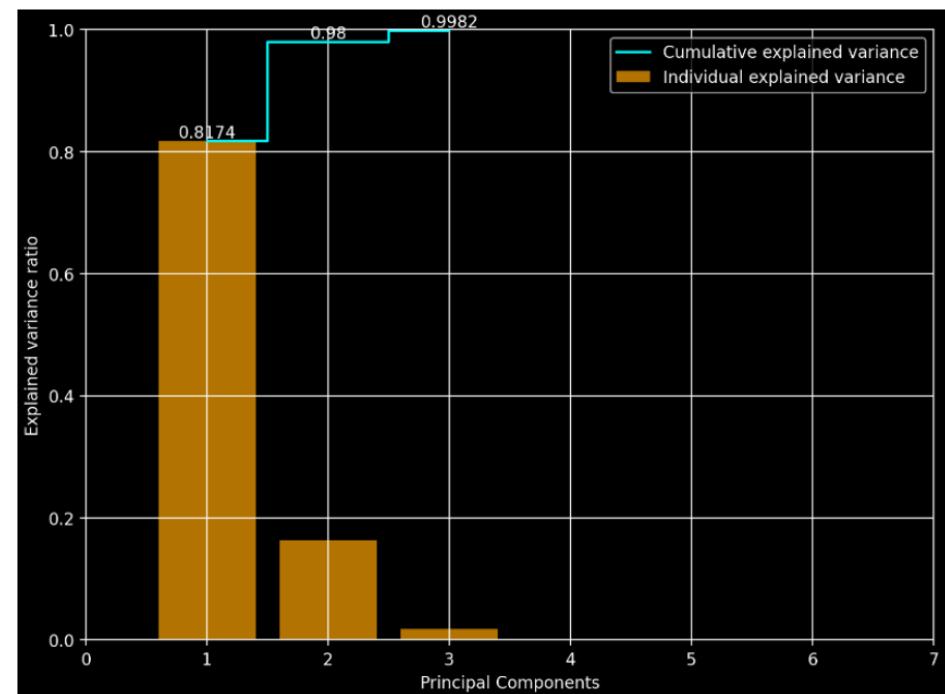
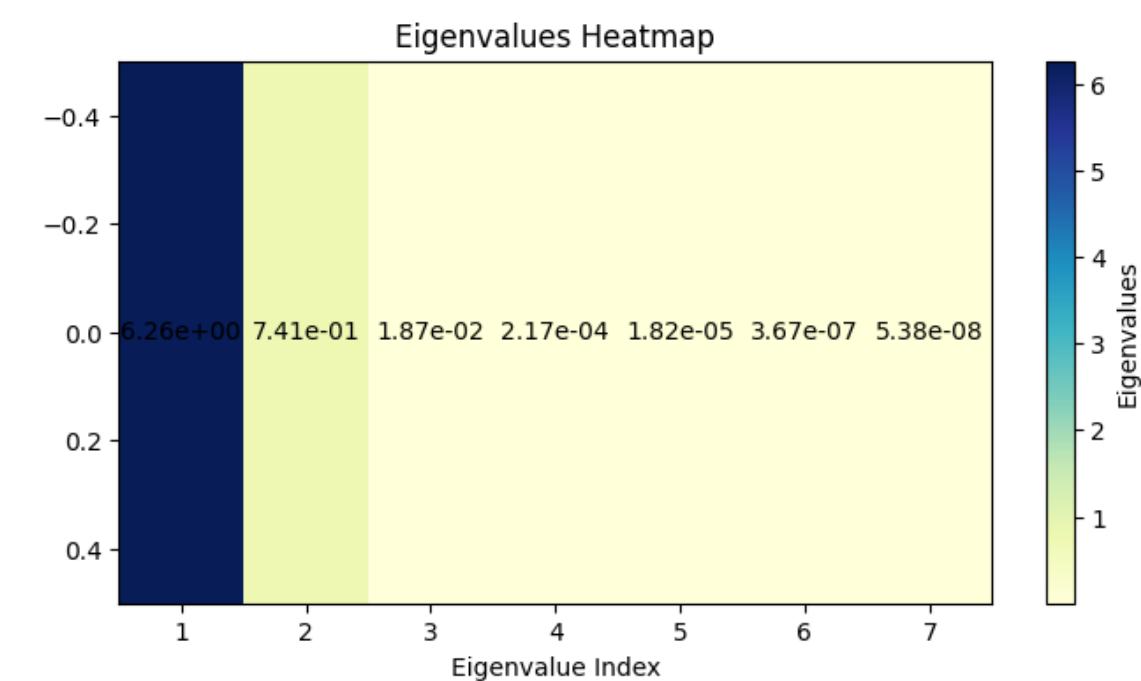
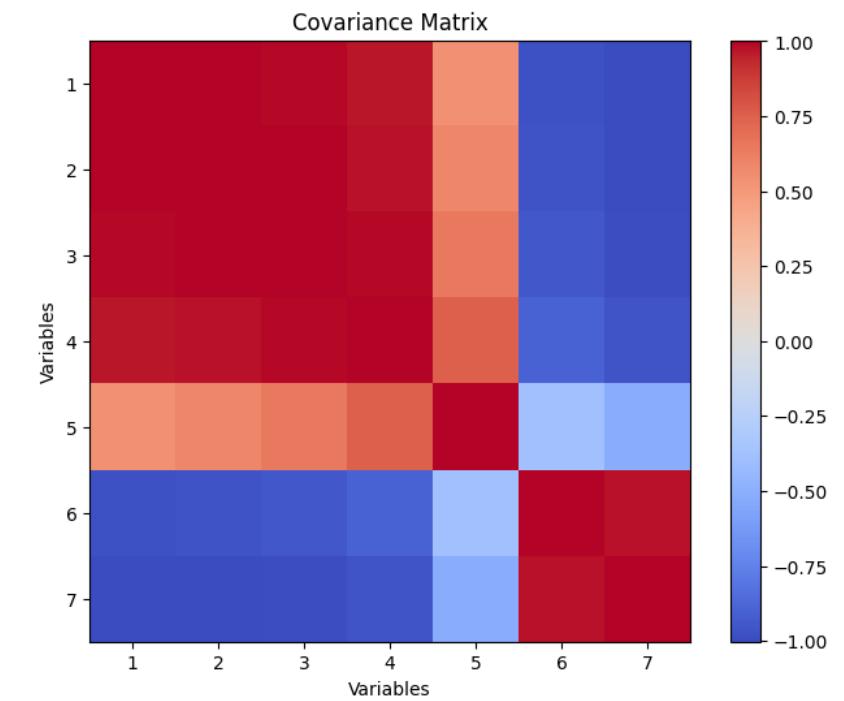
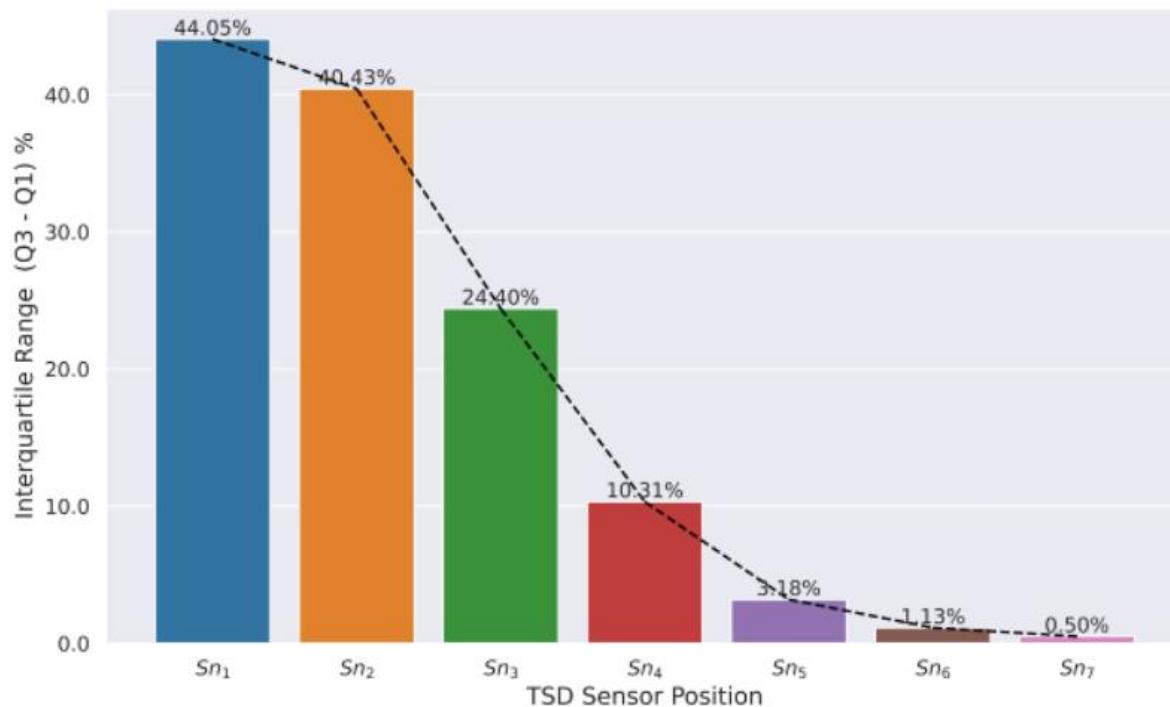
b)



c)



d)



# References

1. S. Fontul, Structural evaluation of flexible pavements using non-destructive tests. PhD thesis, Universidade de Coimbra (Portugal), 2004.
2. J-M. Simonin, V. Le Boursicaud, and P. Hornych, “Correction of deflection bowls measured by rolling devices using simple shape function,” *Transportation Engineering*, vol. 3, p. 100050, 2021.
3. S. Katicha, G. Flintsch, and B. Diefenderfer, “Ten years of traffic speed deflectometer research in the United States: A review,” *Transportation research record*, vol. 2676, no. 12, pp. 152–165, 2022.
4. F. Xiao, Q. Xiang, X. Hou, and S. N. Amirkhanian, “Utilization of traffic speed deflectometer for pavement structural evaluations,” *Measurement*, vol. 178, p. 109326, 2021.
5. Y. LeCun, Y. Bengio, and G. Hinton, “Deep learning,” *nature*, vol. 521, no. 7553, pp. 436–444, 2015.
6. J.-F. Corté and M.-T. Goux, “Design of pavement structures: the french technical guide,” *Transportation Research Record*, vol. 1539, no. 1, pp. 116–124, 1996.
7. Horak, Emile. (2008). Benchmarking the structural condition of flexible pavements with deflection bowl parameters. *Journal of the South African Institution of Civil Engineers*. 50.



Thank you  
for your attention