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|----------|--|--|--|---|--|
| 1 | Assessing decision tree, random forest, and XGBoost models for human capital readiness predictions in low-income areas | <ol style="list-style-type: none"> 1. รศ.ดร. พิทา จารุพูนผล 2. ผศ.ดร.วิภาวรรณ บัวทอง 3. ผศ.ดร.วิจิตรา สุจริต 4. ดร.สุทธาสินี คุปตะบุตร | <ol style="list-style-type: none"> 1. Interpretable Machine Learning Framework for Resilient Modulus Estimation Using LTPP Data for Pavements 2. Balancing accuracy, interpretability, and stability in machine-learning models: Live-weight prediction of Andean sheep from morphometric traits 3. A Dynamic Weighted Fusion-Based Hybrid XGBoost-LSTM Model for Financial Distress Prediction 4. Geospatial multi-criteria | <ol style="list-style-type: none"> 1. https://www.sciencedirect.com/science/article/pii/S221450952501201X 2. Balancing accuracy, interpretability, and stability in machine-learning models: Live-weight prediction of Andean sheep from morphometric traits Scientia Agropecuaria 3. https://dl.acm.org/doi/full/10.1145/3745238.3745276 4. Geospatial multi-criteria decision | <ol style="list-style-type: none"> 1. Scopus Q1 2. Scientia Agropecuaria (Scopus Q3) 3. ACM Digital Library |

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| | | | <p>decision analysis for vegetation health index mapping and landslide susceptibility assessment</p> <p>5. Synthetic index for the evaluation of territorial poverty in the municipalities of Querétaro, Mexico</p> <p>6. A Dynamic Weighted Fusion-Based Hybrid XGBoost-LSTM Model for Financial Distress Prediction</p> <p>7. Human Resource Demand Prediction Based on Particle Swarm Optimization and Convolutional Neural Network Model</p> <p>8. XGBoost and Mixed Effect Model: Can Zakat (Alms) Improve the Human</p> | <p>analysis for vegetation health index mapping and landslide susceptibility assessment Discover Geoscience Springer Nature Link</p> <p>5. https://www.proquest.com/openview/e8f3ad22f8a228df9612e11a748e21b9/1?pq-origsite=scholar&cbl=6593640</p> <p>6. https://dl.acm.org/doi/full/10.1145/3745238.3745276</p> <p>7. https://ieeexplore.ieee.org/abstract/document/11407089/references#references</p> <p>8. https://ejournal.uinsgd.ac.id/index.php/kjrt/article/view/1539</p> | |
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1. Introduction

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3. Results and discussion

4. Conclusions

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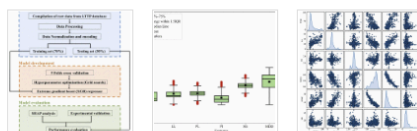
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Case Studies in Construction Materials


Volume 23, December 2025, e05403



Interpretable machine learning framework for resilient modulus estimation using LTPP data for pavements

Ishfaq Rashid Sheikh^{a b c}, Ming Zhang^{d e}, Xiaohui Sun^{a b c}  , Changqing Chen^{d f},
Xiangsheng Chen^{a b c}, Zijun Dong^{a b c}, Foci Chen^{d f}


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Highlights

- XGBoost SHAP Optuna framework predicts subgrade resilient modulus (M_r).
- Model achieves high accuracy ($R^2 = 0.91$ train, 0.82 test; RMSE = 1.77; MAE = 0.78).

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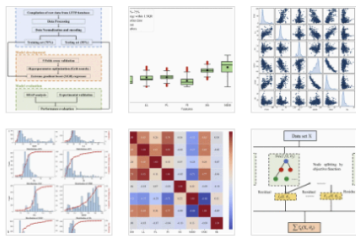
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