Smart Sensors: Safeguarding Everyday Infrastructure in Real Time

Adaptive Real-Time Systems Laboratory (ARTS-Lab) An Interdisciplinary Controls Lab at USC

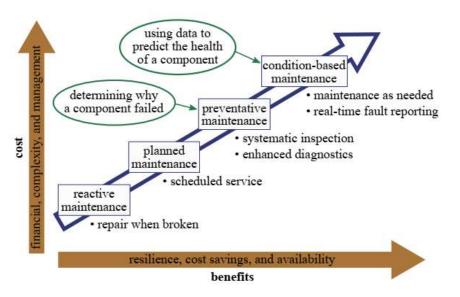
> Austin R.J. Downey Associate Professor Mechanical Engineering Civil and Environmental Engineering



Molinaroli College of Engineering and Computing

Introduction

- Part 1: Civil Infrastructure
- Part 2: Applications and Case Studies







Civil Infrastructure

Civil Infrastructure

- Civil infrastructure includes essential public systems and facilities:
 - Roads and highways
 - Bridges and tunnels
 - Water and sewage systems
 - Dams, levees, and flood control structures
 - Electrical grids, transportation networks, and more



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Infrastructure Maintenance and Upgrade

- Importance of Maintenance and Upgrades:
 - Critical for public safety and economic stability
 - Ensures infrastructure longevity and reliability
 - Necessary to handle increasing demand and urbanization
 - Key to reducing the risk of catastrophic failures, like bridge collapses or dam breaches



Donald Trung Quoc Don (Chữ Hán: 徵國單) - Wikimedia Commons - © CC BY-SA 4.0 International



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Infrastructure Challenges

Traditional Challenges:

- Aging infrastructure and limited budgets for repairs
- Manual inspection processes are labor-intensive and timeconsuming
- Delayed detection of structural issues leads to reactive maintenance
- Difficulty in predicting failures due to the complexity of infrastructure systems



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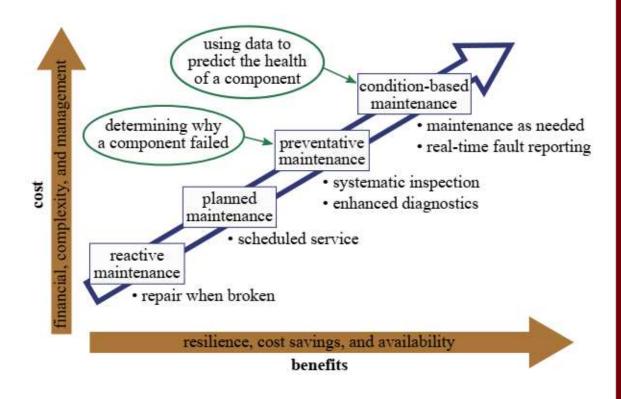


Arlington Memorial Bridge Repair & Reconstruction. National Park Service and Federal Lands Transportation Program. U.S. Department of the Interior. 2013, p. 2., Public domain, via Wikimedia Commons

Predictive Maintenance

AI Use Cases in Infrastructure:

- Continuous Sensing: Real-time strain, vibration, flow, and environmental data
- **On-Edge Analytics:** Automatic anomaly detection and remaining-useful-life (RUL) estimation
- Optimized Scheduling: Maintenance only when and where it's needed
- Improved Resilience: Fewer unplanned outages, lower lifecycle costs, higher availability



Austin Downey. Sensing skin for the structural health monitoring of mesoscale structures. Iowa State University, 2018, Iowa State University Graduate Theses and Dissertations. 16571

Goal: Self-Monitoring & Adaptive Infrastructure

Civil Engineering and Al include:

- Continuous Sensing: Embedded sensors track stress, strain, water level, vibration, etc., in real time
- Automated Actuation: Valves, gates, dampers or cables adjust themselves to changing loads or water pressures
- **Predictive Maintenance:** On-board analytics flag emerging issues— corrosion, fatigue, blockages
- Enhanced Resilience: Infrastructure that detects and reacts autonomously to floods, heavy traffic, or seismic events





Photos of the flood gate by Austin Downey in Örebro Sweden, Creative Commons BY-SA 4.0. 3D view taken from Google Maps in October 2024; copyright held by Google.

Applications and Case Studies

Application 1: Structural Health Monitoring

Structural Health Monitoring (SHM)

- Distributed Sensing: Strain, vibration, tilt and ultrasonic sensors embedded across the structure
- **Continuous Inspection:** Real-time data collection without bucket trucks
- **Damage Detection:** Automated identification of cracks, corrosion, loosened connections
- Actionable Alerts: Threshold-based alarms and dashboard reports for timely maintenance

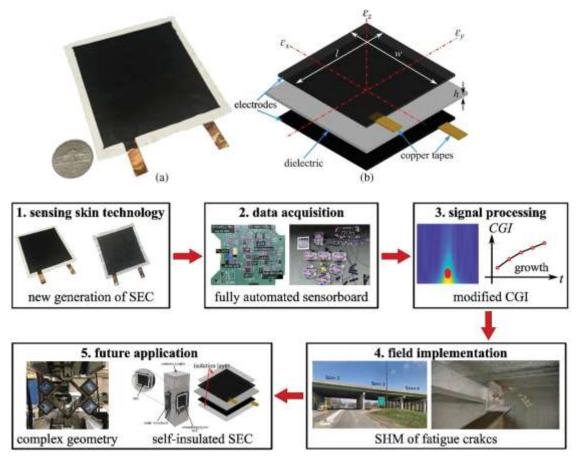


VTrans Rail Bridge Inspection Public Domain

Case Study 1.1: Bridge Monitoring

Key Benefits of AI in Bridge Monitoring

- Fatigue Crack Detection: Sensing skin technology (SEC) detects and monitors fatigue cracks in steel bridges.
- Large-Area Monitoring: Soft elastomeric capacitors (SECs) provide coverage over large bridge surfaces for crack detection.
- Field Validation: The system was successfully deployed on a highway bridge in Kansas, providing real-world validation for the SEC.



Han Liu, Simon Laflamme, Jian Li, Austin Downey, Caroline Bennett, William Collins, Paul Ziehl, Hongki Jo, and Michael Todsen. Sensing skin technology for fatigue crack monitoring of steel bridges: Laboratory development, field validation, and future directions. International Journal of Bridge Engineering, Management and Research, 1(1):21424002-1, 2024

Case Study 1.2: UAV Deployed Sensors

UAV Deployable Sensor Packages

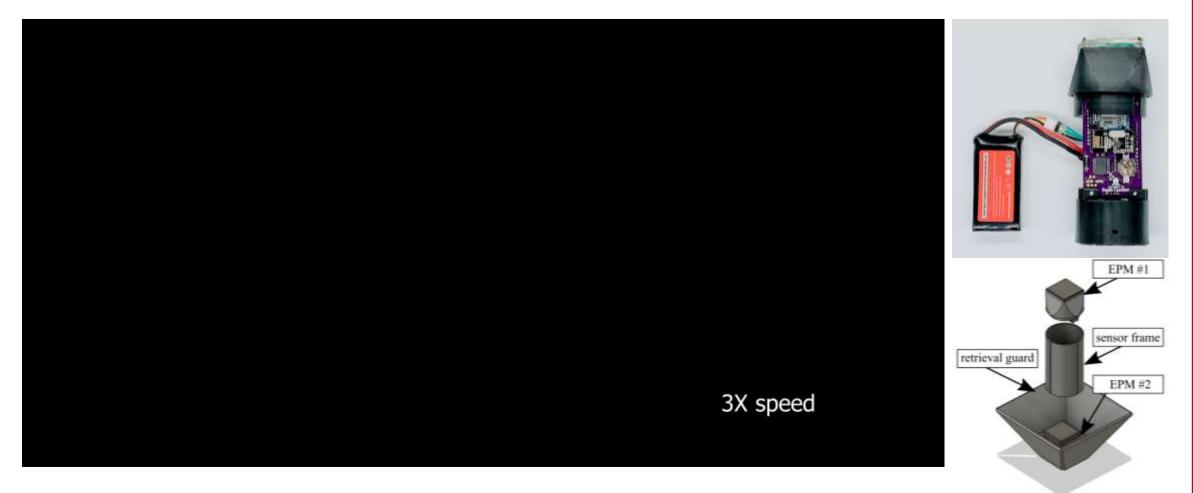
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- Autonomous Deployment: Fast, precise sensor placement.
- Real-Time Monitoring: Continuous data for proactive assessment.
- **Cost & Time Efficient:** Reduces manual inspections.
- Scalable Solution: Works for bridges, levees, and more.
- Enhanced Safety: Minimizes human exposure.



Joud N. Satme, Daniel Coble, Hung-Tien Huang, Austin R. J. Downey, and Jason D. Bakos. Non-linear vibration signal compensation technique for UAV-deployable sensor packages with edge computing. In Zhongqing Su, Maria Pina Limongelli, and Branko Glisic, editors, Sensors and Smart Structures Technologies for Civil, Mechanical, and Aerospace Systems 2023. SPIE, apr 2023. doi:10.1117/12.2658563

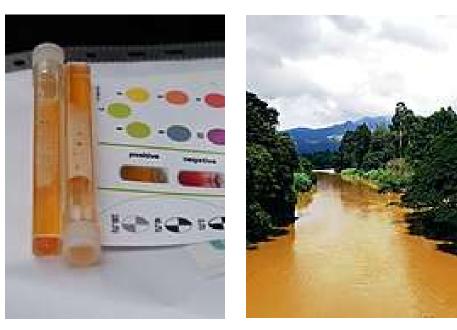
Case Study 1.2: UAV Deployed Sensors



Application 2: Water Quality Monitoring

Water Quality Monitoring

- **Real-Time Monitoring:** Continuous, real-time tracking of water quality parameters (e.g., pH, turbidity, contaminants).
- Early Detection of Contaminants: Machine learning models identify harmful substances in water early, improving response times.
- **Predictive Analysis:** Al predicts potential water quality issues based on environmental and historical data trends.
- **Cost Efficiency:** Reduces the need for manual sampling and testing, optimizing resource use and operational costs.



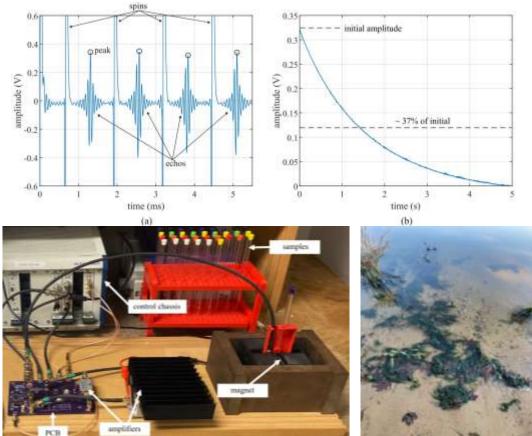
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Case Study 2.1: NMR-based Water Quality Monitoring

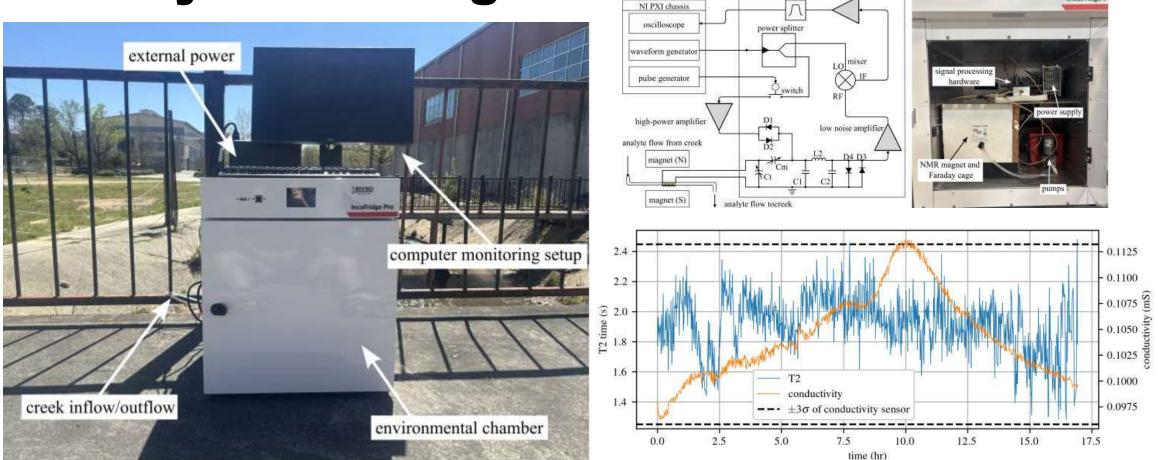
NMR-Based Water Quality Monitoring

- How NMR Works: Magnetic pulses align tiny "magnets" in water molecules, then we listen for their echo signals
- Echo Signals: The pattern and strength of those echoes tell us about what's in the water
- **Portable Design:** A small permanent magnet and custom electronics let us bring the lab to the field



Parker Huggins, Win Janvrin, Jake Martin, Ashley Womer, Austin R. J. Downey, John Ferry, Mohammed Baalousha, and Jin Yan. Assessing magnetic particle content in algae using compact time-domain nuclear magnetic resonance. In Weilin Hou and Linda J. Mullen, editors, Ocean Sensing and Monitoring XVI. SPIE, June 2024. doi:10.1117/12.3013987

Case Study 2.1: NMR-based Water Quality Monitoring



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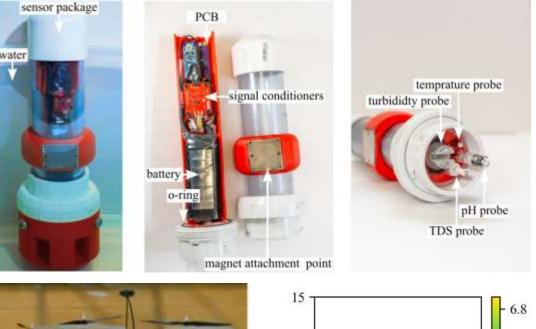
Daniel Hancock, David P. Wamai, Md Asifuzzaman Khan, Winford Janvrin, Austin R. J. Downey, Mohammed Baaloush, and Thomas M. Crawford. Continuous water quality monitoring using field deployable NMR and explainable AI. In Defense and Commercial Sensing. SPIE, May 2025

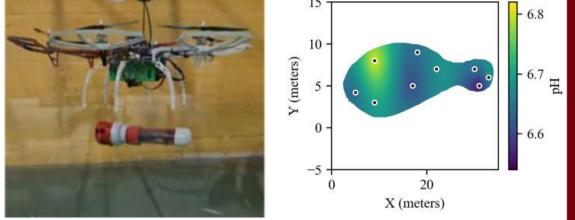
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Case Study 2.2: UAV-deployable in situ Water Quality Sensors

Key Benefits of in situ Sensors

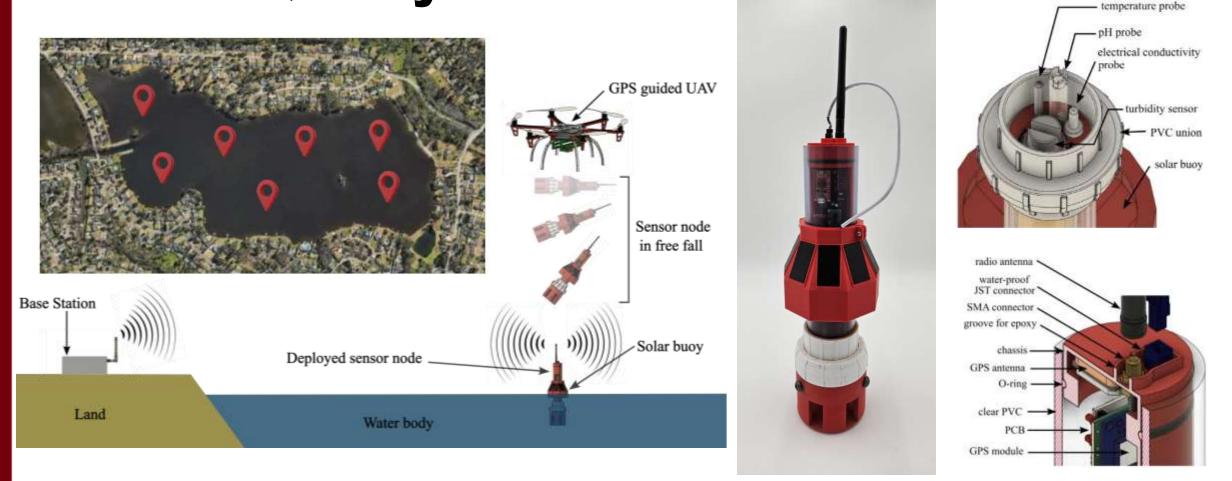
- Real-Time Monitoring: UAVs deploy water quality sensors to provide continuous, real-time data.
- **Rapid Deployment:** UAVs allow fast, efficient sensor deployment in remote or hazardous locations.
- Spatial and Temporal Analysis: Interpolation techniques map water quality over space time.
- **Cost-Effective Solution:** Affordable, open-source sensors provide reliable water quality data.





M. Burnett, M. Abdelwahab, J. N. Satme, A. R. J. Downey, A. Fonce, and I. Jasim, "Spatial and Temporal In-Situ Water Quality Monitoring and Mapping via UAV-Deployable Sensor Nodes," In Development, 2024.

Case Study 2.2: UAV-deployable in situ Water Quality Sensors

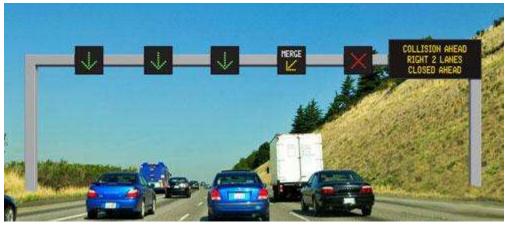


18 Md Asifuzzaman Khan, Matthew Burnett, Austin R. J. Downey, Joud N. Satme, and Jasim Imran. UAV deployable buoy-style sensor for in situ water quality monitoring. In Defense and Commercial Sensing. SPIE, May 2025

Application 3: Traffic and Transportation

Key Benefits of AI in Transportation

- **Traffic Flow Optimization:** Manage and optimize traffic flow in real-time, reducing congestion.
- **Predictive Traffic Management:** Forecast traffic patterns, allowing cities to adjust signals and infrastructure accordingly.
- Autonomous Vehicle Integration: Al plays a crucial role in the development and management of autonomous vehicles, enhancing safety and efficiency.
- Smart Public Transportation: Al enables efficient routing, scheduling, and capacity management for public transport systems.



Washington State Department of Transportation: Public Domain

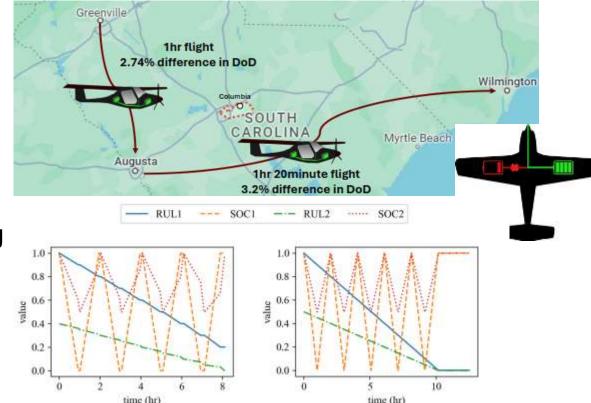


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Case Study 3.1: Electric Aircraft Optimization

Electric Aircraft Traffic Routing

- Route-Based Maintenance Scheduling: Use flight paths and traffic density data to predict optimal maintenance windows, minimizing unscheduled downtime.
- **Battery Life Forecasting:** Forecast remaining useful life (RUL) by accounting for route profiles.
- Traffic Corridor Optimization: Design and update airspace corridors to reduce delays and idle loitering—lowering overall energy.



George Anthony, Nathaniel Cooper, Jarrett Peskar, Austin R. Downey, and Kristen Booth. Extending battery life via load sharing in electric aircraft. In AIAA SCITECH 2024 Forum. American Institute of Aeronautics and Astronautics, January 2024. doi:10.2514/6.2024-2154

Korebami O. Adebajo, Nathaniel Cooper, and Austin R.J. Downey. Battery degradation prediction aided by multi-domain modeling. 2024 Battery Safety Workshop, August 2024

Application 4: Flood Modeling and Forecasting

Flood Modeling and Forecasting

- **Real-Time Data Analysis:** Process real-time sensor and weather data to predict flood risks.
- **Improved Forecast Accuracy**: Machine learning models enhance the accuracy of flood forecasts.
- Early Warning Systems: Al-driven models provide early flood warnings, improving disaster preparedness.
- **Risk Mapping:** Flood risk maps to identify vulnerable areas and inform urban planning.
- **Emergency Response:** Optimize resource allocation during flood events for efficient response.



U.S. Department of Agriculture, Public domain, via Wikimedia Commons

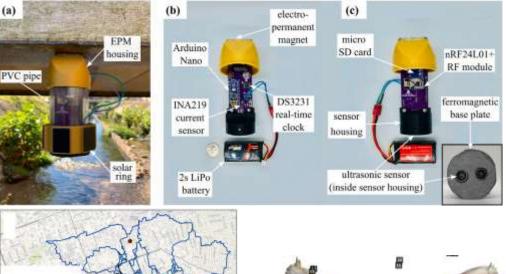


Staff Sgt. Roberto Di Giovine, Public domain, via Wikimedia Commons

Case Study 4.1: Low-cost Height Sensors

Flood Modeling and Forecasting

- Real-Time Monitoring: Sensors deployed by UAVs collect real-time water height data, providing real-time information during floods.
- **Optimization:** Optimize flood model parameters in real-time.
- **IoT Integration:** IoT-enabled sensors for seamless data transmission and faster model updates, improving flood response times.
- **Predictive Forecasting:** Enhance flood prediction by processing large data sets and optimizing forecasts for urban watersheds.

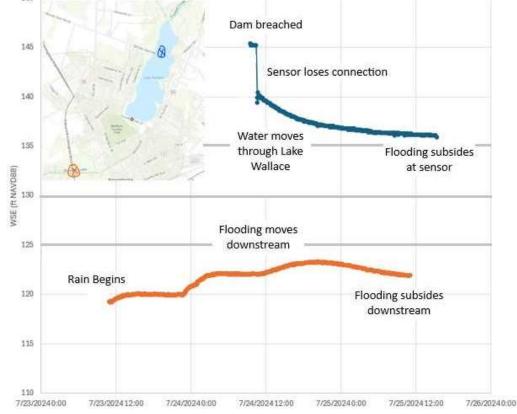




Ahad Hasan Tanim, Corinne Smith-Lewis, Austin R.J. Downey, Jasim Imran, and Erfan Goharian. Bayes_opt-swmm: A Gaussian process-based Bayesian optimization tool for real-time flood modeling with SWMM. Environmental Modelling & Software, page 106122, June 2024. doi:10.1016/j.envsoft.2024.106122 Corinne Smith, Joud Satme, Jacob Martin, Austin R.J. Downey, Nikolaos Vitzilaios, and Jasim Imran. UAV rapidly deployable stage sensor with electro-permanent magnet docking mechanism for flood monitoring in undersampled watersheds. HardwareX, 12:e00325, oct 2022. doi:10.1016/j.ohx.2022.e00325.

Case Study 4.1: Low-cost Height Sensors





Bennettsville dam failure blamed on internal erosion, state

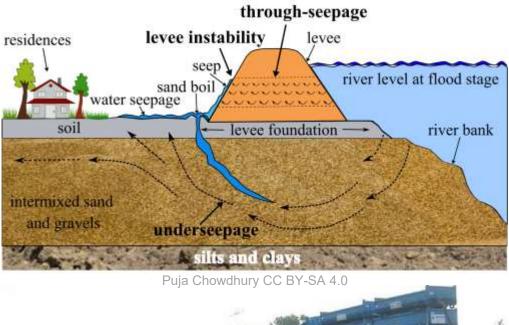


Internal erosion is to blame for a May 6 breach of the Lake Paul Wallace Dam in Marlboro's largest city - and the structure remains in need of ...

Application 5: Geotechnical Monitoring

Geotechnical Monitoring

- **Real-Time Monitoring:** Al processes sensor data to monitor soil movement and stability in real time.
- Early Detection of Failures: Detects early signs of slope instability, landslides, and foundation settlement.
- **Predictive Maintenance:** Forecast potential geotechnical issues, allowing for timely interventions.
- **Cost Reduction:** Monitoring reduces the need for frequent manual inspections and prevents costly failures.



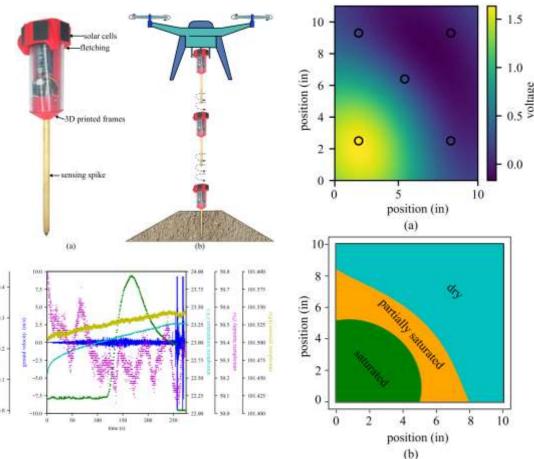


V. Bennett, T. Abdoun, M. Zeghal, A. Koelewijn, M. Barendse, R. Dobry, CC BY 4.0 https://creativecommons.org/licenses/by/4.0, via Wikimedia Commons

Case Study 5.1: UAV-Deployable Soil Saturation Sensors

Soil Saturation Monitoring

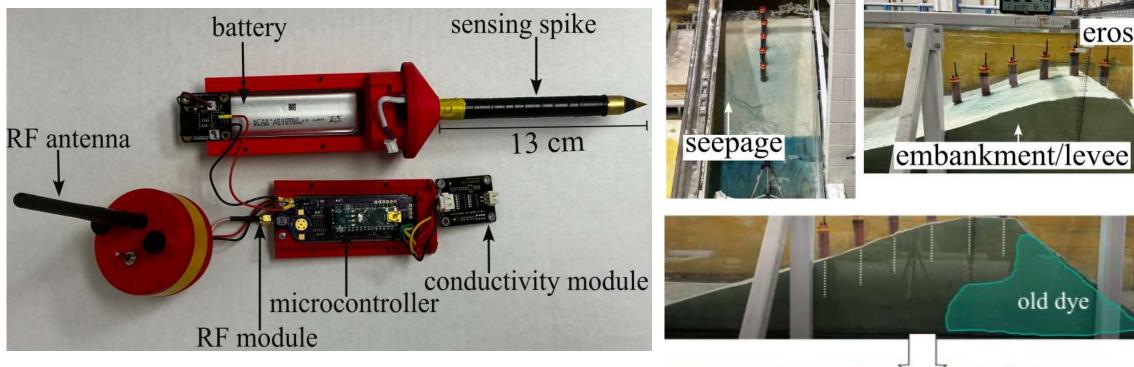
- **Real-Time Monitoring:** UAV-deployed smart sensing spikes provide continuous soil moisture monitoring across levees.
- **Data Expansion:** Gaussian process regression (kriging) to generate continuous moisture maps from discrete sensor data.
- Automated Classification: Categorize soil conditions into dry, partially saturated, and saturated zones using k-means clustering.
- Early Detection: The system predicts areas at risk of levee failure by monitoring soil saturation and detecting seepage.



Puja Chowdhury, Joud N. Satme, Ryan Yount, Austin R. J. Downey, Sadik Khan, Jasim Imran, and Laura Micheli. Classifying soil saturation levels using a network of UAV-deployed smart penetrometers. In ASME 2023 Conference on Smart Materials, Adaptive Structures and Intelligent Systems, SMASIS2023. American Society of Mechanical Engineers, September 2023. doi:10.1115/smasis2023-111009

Case Study 5.1: UAV-Deployable Soil Saturation Sensors

flume



Sydney Morris, Ayman Mokhtar Nemnem, Malichi Flemming, Austin R. J. Downey, Puja Chowdhury, Matthew Burnett, Jasim Imran, and Sadik Khan. Self-Contained Electrical Conductivity Sensing Spikes for Monitoring of Levee Wetting and Drying Cycles. In Proceedings of the International Design Engineering Technical Conferences & Computers and Information in Engineering Conference (IDETC-CIE), Anaheim, California, August 17 2025



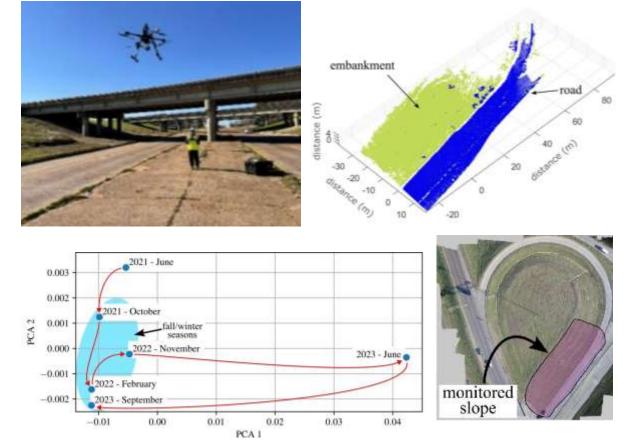
erosion

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Case Study 5.2: LIDAR-based Monitoring

LIDAR-Based Monitoring for Slope Stability

- **High-Resolution Data:** LiDAR captures 3D topography for precise monitoring of slopes.
- Seasonal Monitoring: Track moisture variations and their impact on soils.
- **Risk Assessment:** LiDAR scans identify potential failures, offering early warnings for slope instability.
- Efficient Processing: Advanced algorithms speed up data analysis for real-time monitoring.



AQM Zohuruzzaman, David P. Wamai, Weicong Feng, Sadik Khan, Austin R. J. Downey, Jie Wei, Erik Blasch, and Paul T. Schrader. Highway slope monitoring using 3D laser scanning at different seasons. In Kannappan Palaniappan and Gunasekaran Seetharaman, editors, Geospatial Informatics XIV. SPIE, June 2024. doi:10.1117/12.3016172

Challenges for Active & Smart Infrastructure

Challenges:

- Aging Assets: Many bridges, tunnels, and water-control structures exceed their design life, increasing collapse risk
- **Private Autonomous Systems:** Self-driving cars, delivery drones, and other on-demand vehicles introduce new, unpredictable loads and traffic patterns
- Delayed Maintenance: Scheduled service can lead to under- or over-maintenance, wasting resources



Volvo Cars Data Collection Video, Austin Downey CC BY-SA 2.0

Conclusion

Final Thoughts:

- Active Infrastructure: Sensors and actuators enable real-time self-monitoring.
- **Predictive Maintenance Pays Off:** Early fault detection cuts downtime.
- Interdisciplinary Teams Win: Collaboration drives resilient designs.
- Field-Proven Technologies: Case studies show practical impact.
- Your Turn to Build the Future: Pursue careers that protect and sustain.

No researchers were harmed during this endeavor!



Questions and Discussion



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