

INSPECTING AND PRESERVING INFRASTRUCTURE THROUGH ROBOTIC EXPLORATION

Robot-Assisted Bridge Inspection and Maintenance

TRB Webinar on

Robot-Enabled Sensing and Augmented Learning (RESEAL) for Bridge Inspection

Genda Chen, Ph.D., P.E., F.ASCE, F.SEI, F.ISHMII Director of INSPIRE University Transportation Center Missouri University of Science and Technology March 29, 2022, 1:00 – 2:30 pm



Outline of This Presentation

- Bridge Element Inspection in the U.S.
 - > 2019 Manual for Bridge Element Inspection
- Data-driven Bridge Management
 - > INSPIRE University Transportation Center Goal
 - Sensing and Nondestructive Evaluation Integration for Bridge Inspection and Maintenance
 - > Robotic Platforms
- Robot-assisted Bridge Maintenance and Inspection
 - Fatigue evaluation of steel structures from thermal imaging
 - > Scour evaluation from smart rocks as magnetic sensors
- Pooled-fund Study on 72 Highway Bridges
- Concluding Remarks

Bridge Element Inspection

• 2019 Manual for Bridge Element Inspection

- > All elements have 4 conditions of deterioration
- > Current visual inspection
 - ✓ Difficult to record all defects
 - ✓ Difficult to accurately estimate quantities

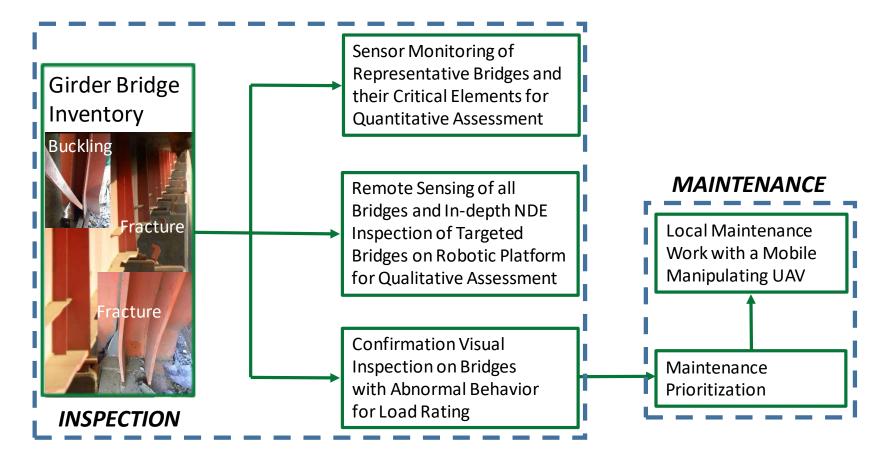
		Condition States				
Manual for Bridge		1	2	3	4	
Element Inspection	Defects	GOOD	FAIR	POOR	SEVERE	
Scord Editor, 2019	Delamination/Spall/ Patched Area (1080)	None.	Delaminated. Spall 1 in. or less deep or 6 in. or less in diameter. Patched area that is sound.	Spall greater than 1 in. deep or greater than 6 in. diameter. Patched area that is unsound or showing distress. Does not warrant structural review.	The condition warrants a structural review	
	Exposed Rebar (1090)	None.	Present without measurable section loss.	Present with measurable section loss but does not warrant structural review.	to determine the effect on strength or serviceability of the element of a bridge.	

Data-driven Bridge Management

- 5-Year Goal at the INSPIRE University Transportation Center
 - To transform the current labor-intensive, inconsistent, and expensive inspection and maintenance process into an efficient, safe, reliable, and costeffective management system for bridges
 - Make a paradigm shift from ad hoc local processes to a data-driven decision-making protocol
 - Involve basic, advanced, and applied research in sensing, nondestructive evaluation (NDE), data analytics, robotics, and workforce development

Data-driven Bridge Management

 Contact/Remote Sensing and NDE Integration for Bridge Inspection and Maintenance



Robotic Platforms

Unmanned Aerial Vehicle (UAV)

> Close-distance inspection



H. Zhang, Z. Li, G. Chen, A. Reven, B. Scharfenberg & J. Ou. UAV-based smart rock localization for bridge scour monitoring, *Journal of Civil Structural Health Monitoring*, https://doi.org/10.1007/s13349-020-00453-w.



Robotic Platforms

 Structural Crawler in Collaboration with Dr. Hung La from the University of Nevada, Reno
Near-surface inspection and testing



Robotic Platforms

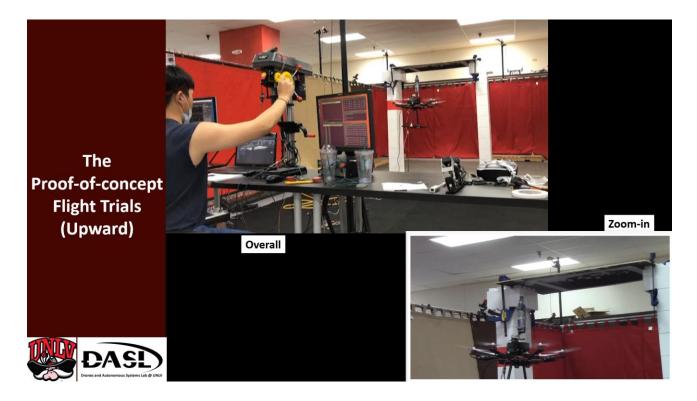
- Hybrid Unmanned Vehicle (BIRDS, proof of concept)
 - > Increased operation time
 - > Stable measurement platform
 - > Ease in navigation
 - > Accurate positioning

A. Reven, C. Fritsche & G. Chen (2019). Unmanned aerial and traversing robot as mobile platform for bridge inspections, *Proceedings of the International Conference on Structural Health Monitoring of Intelligent Infrastructure*, St. Louis.



Robot-assisted Bridge Maintenance

- UAV with Aerial Manipulator (proof of concept)
- Dr. Paul Oh's Laboratory at the University of Nevada, Las Vegas



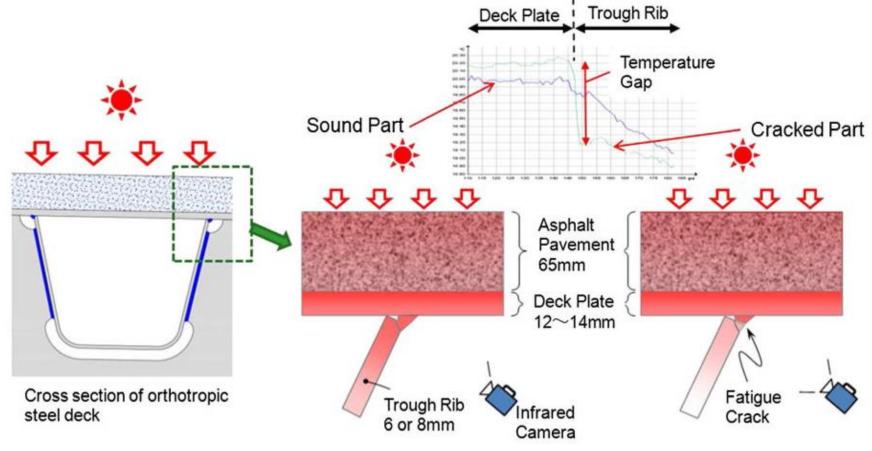
• Bridge No. 3128 – Steel and PC Girders, MO



Elios 2 Drone and Passive Thermal Image

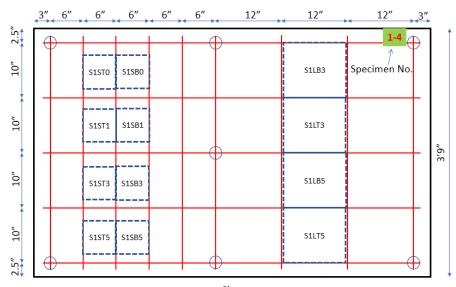


• Through-weld Fatigue Crack Detection from Passive Thermal Imaging in a Bridge



Sakagami, T. (2015) Remote nondestructive evaluation technique using infrared thermography for fatigue cracks in steel bridges. Fatigue & Fracture of Engineering Materials & Structures. doi: 10.1111/ffe.12302 (an invited review article).

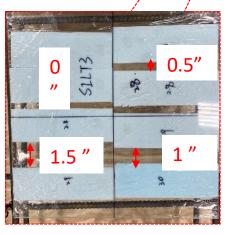
Concrete Delamination Detection from Active Thermal Imaging



6'		
Material	Conductivity (W/mK)	Specific heat (MJ/m³K)
Extruded Polystyrene (XPS)	0.0243	0.025
Acrylic board (Plastic holder)	0.213	1.274
Concrete	2.006	1.807
Steel reinforcement	45	3.5
Air	0.025	1.000

Not in final positions

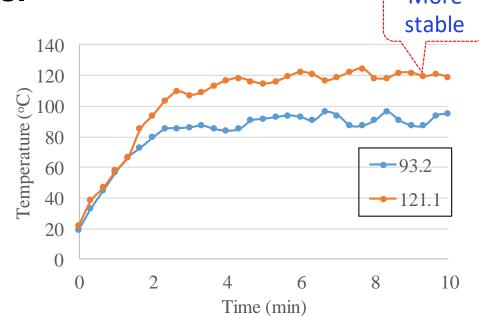




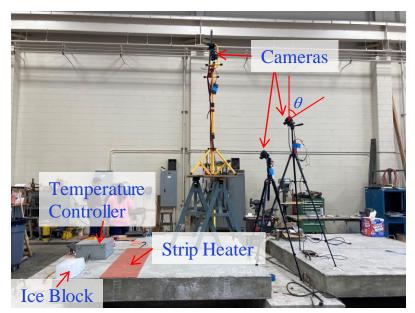
- 4 Slabs S1-S4
- 2 types of delamination: small and large
- 2 thicknesses: 3 and 5 mm in large foams
- 4 thicknesses: 0, 1, 3, and 5 mm in small foams
- Plastic holder: 2 mm

Test Procedure

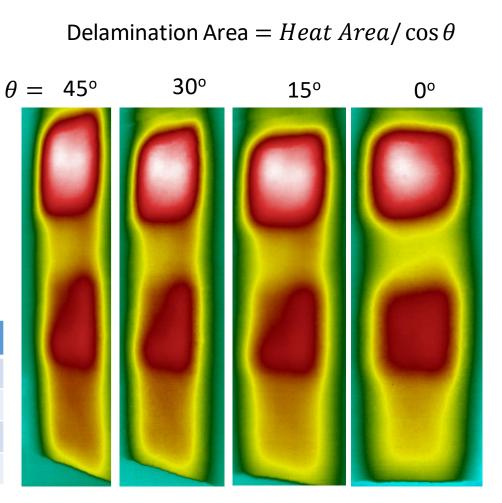
- > Heat each slab using a strip heater for 10 minutes to achieve the highest temperature of 121.1 °C.
- Remove the strip heater and then take thermal images every 1 minute during the cooling process.



• Effect of the Angle of Camera View

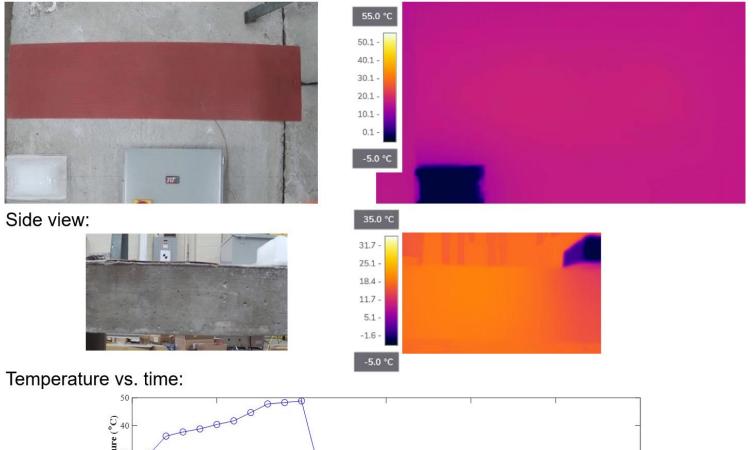


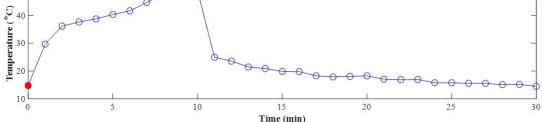
Angle of view $ heta$ (°)	Delamination Area (mm ²)	Error
0	63826	3.0%
15	66485	6.8%
30	67013	8.2%
45	67880	9.6%



• Temperature on Top and Side Surfaces of the Heater

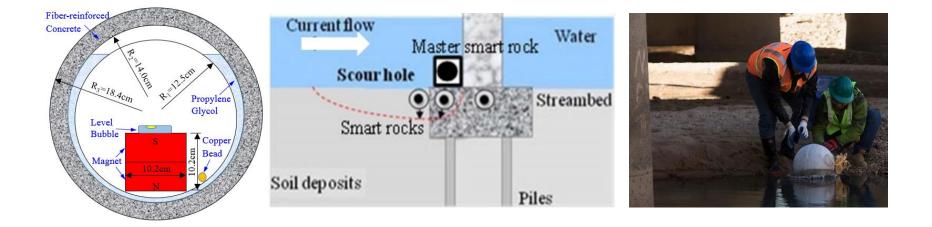
Top view:



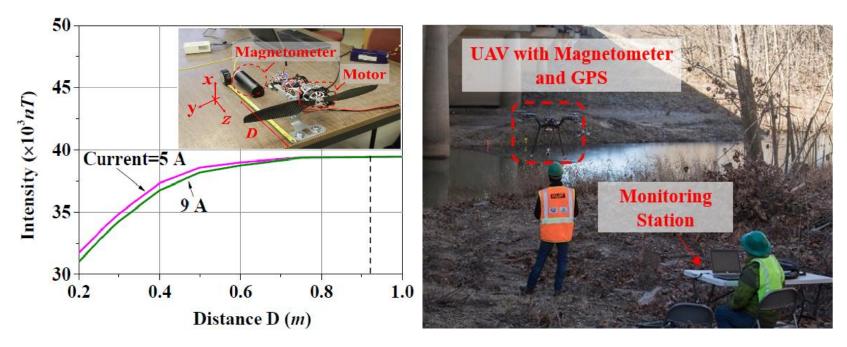


Scour Evaluation from Smart Rocks

- > A smart rock in ball shape
 - Easy to roll to the bottom of a scour hole when formed at an unknown location and depth as deposits around the hole are washed away.
 - Designed based on water flow velocity and soil shear resistance.



- A Magnetometer
 - > Set 0.92 m away from the motors to eliminate magnetic interference
 - > Wired and plugged into the WIFI router with wireless communication with the nearby ground station



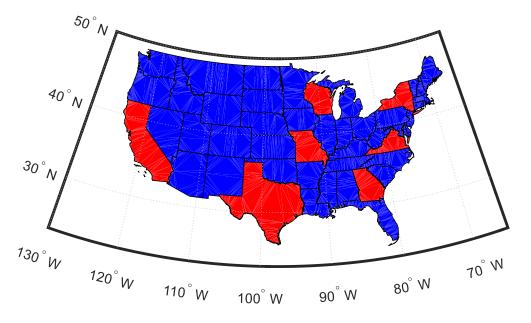
Haibin Zhang, Zhaochao Li, Genda Chen, Alec Reven, Buddy Scharfenberg and Jinping Ou. "UAV-based Smart Rock Localization for Bridge Scour Monitoring," Journal of Civil Structural Health Monitoring, January 19, 2021. <u>https://doi.org/10.1007/s13349-</u>

• Drone- vs. Crane-based Measurements

Monitoring		Predicted coordinate			Measured coordinate			Error
method	Date	Х	Y	Ζ	Х	Y	Z	(m)
CRANE (1st)	11/06/2015	0.06	23.49	-3.03	0.09	23.24	-3.04	0.26
CRANE (2 nd)	04/14/2016	0.55	24.38	-3.21	0.37	24.60	-3.38	0.33
CRANE (3rd)	10/20/2016	0.00	22.73	-2.59	0.00	22.63	-2.87	0.30
UAV (4 th)	01/24/2018	0.02	23.50	-2.89	0.25	23.77	-2.93	0.36
UAV (5 th)	05/10/2018	0.49	25.00	-2.81	0.45	24.78	-3.01	0.30
UAV (6 th)	10/08/2018	0.43	25.07	-2.76	0.41	24.84	-2.98	0.32
UAV (7 th)	02/25/2019	0.37	25.60	-3.16	0.35	25.50	-3.41	0.28
UAV (8 th)	05/17/2019	0.43	24.00	-3.02	0.26	23.80	-3.17	0.30
UAV (9 th)	08/27/2019	0.41	23.32	-3.12	0.23	23.53	-3.22	0.29

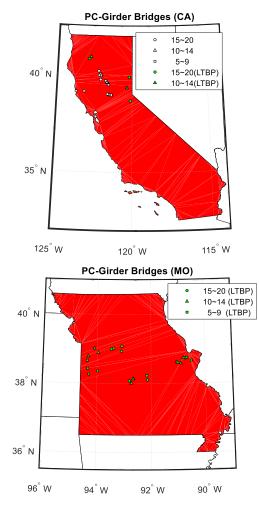
Goals

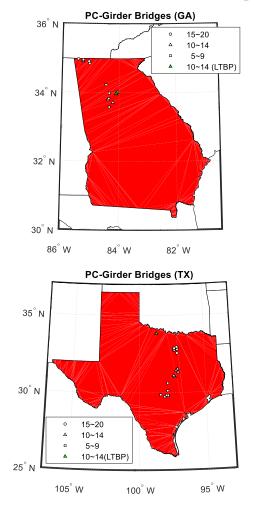
- > To engage closely with 7 state Departments of Transportation (DOTs).
- > To leverage the center resources to develop case studies, protocols, and guidelines that can be adopted by state DOTs for bridge inspection without adversely impacting traffic flow.



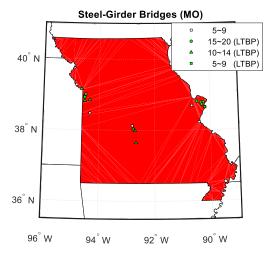
- Bridge Selection
 - > National Bridge Inventory
 - > Long Term Bridge Performance Program
- Bridge Selection Criteria
 - State owned
 - Not over a railroad
 - > Max span length between 10 and 50 m
 - > Maximum of four lanes on bridge
 - > Average daily traffic (ADT) less than 50,000
 - > Built after 1970
 - > Three Age Groups
 - ✓ 15-20 Years
 - ✓ 25-30 Years
 - ✓ 35-40 Years

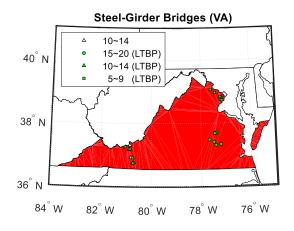
• 36 Prestressed Concrete Girder Bridges

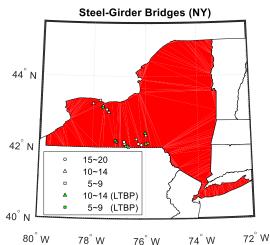


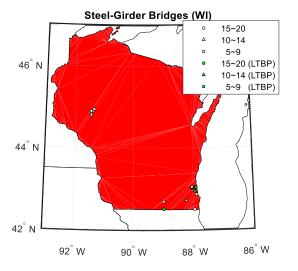


36 Steel Girder Bridges





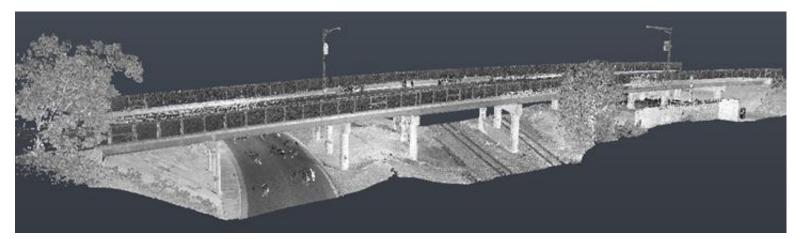


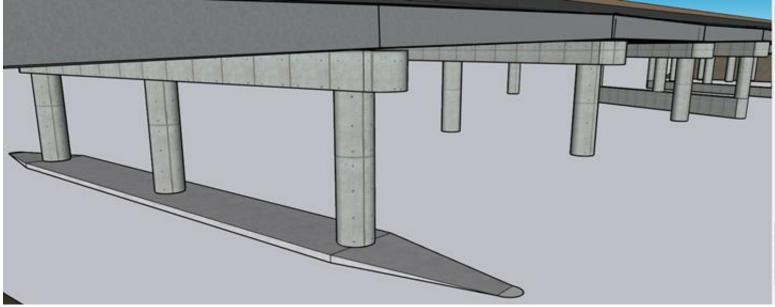


- Steel-girder Bridge No. 3080, Rolla, MO
- Bridge Scanning Using P40 Laser Scanner



• 3D Reconstruction of Bridge No. 3080





Concluding Remarks

- An automated inspection platform can help inspect and maintain bridges faster, safer, cheaper, and more consistent.
- Advanced technologies to support the automated inspection platform are being developed in the INSPIRE UTC.
- The pooled-fund initiative can help develop case studies, protocols, and guidelines that can potentially be adopted by state DOTs for bridge inspection and maintenance.

Acknowledgement

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- The advanced technologies presented were originally developed with financial support from the U.S. Department of Transportation, Office of the Assistant Secretary for Research and Technology (USDOT/OST-R) under Grant No.
 69A3551747126 through INSPIRE University Transportation Center (<u>http://inspire-utc.mst.edu</u>) at Missouri University of Science and Technology.
- The views, opinions, findings and conclusions reflected in this publication are solely those of the authors and do not represent the official policy or position of the USDOT/OST-R, or any State or other entity.



INSPECTING AND PRESERVING INFRASTRUCTURE THROUGH ROBOTIC EXPLORATION

Climbing Robots for Steel Bridge Inspections

TRB Webinar on

Robot-Enabled Sensing and Augmented Learning (RESEAL) for Bridge Inspection

Hung (Jim) La, Ph.D., M.ASCE, F.IEEE

Associate Director of INSPIRE University Transportation Center

University of Nevada, Reno

March 29, 2022, 1:00 – 2:30 pm



Talk Outline

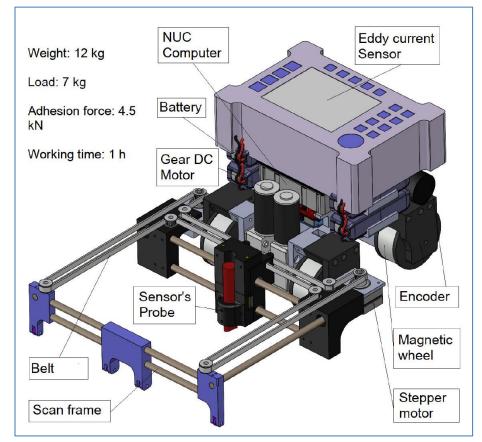
- Overview of recent development of steel climbing robots
- Robotic developments for steel bridge inspection from the Advanced Robotic and Automation (ARA) Lab, University of Nevada, Reno
- Defect (crack, rust) detection with the climbing robot

Overview of steel climbing robot developments

Robot	Type of Locomotion	Detail	Climbing Ability		Adhesion Method	
Mecanum-wheel Robot	4-mecanum wheels	681 × 559 × 323 mm	Flat Concave	x	Permanent Magnet	
[Kamdar2015]		34 kg	Convex Cylinder	x	(untouched)	
-		0.64 m/min				
Tank-like Robot Versatrax100-Inuktun]	2 roller chains	376 x 220 x 115 mm	Flat Concave	x x	Permanent Magnet	
Versatiax100-maxtunj		4.5 kg	Convex	A	(untouched)	
		0.15 m/s	,			
4-Wheels robot	4 magnetic wheels	352 × 215 × 155 mm	Flat Concave	x x	Permanent	
[Wang_ICA2014]	(flexible frame)	3 kg	Convex Cylinder	x	Magnet	
		0.32 m/s				
2-Wheels robot [Eich_MED2015]	2 magnetic wheels	380 x 280 x 150 mm	Flat Concave	x x	Permanent	
[Elch_MED2013]		0,67 kg	Convex	x	Magnet	
6.3		0.5 m/s	Cymaei			
Inch-worm robot [Ward2016]	7-DOF-Inch worm	220 x 240 x 150 mm	Flat Concave	x	Permanent Magnet	
		18 kg	Convex Cylinder		Wagnet	
		0.2 m/s	oyimadi			
Spider robot [Genki Sato2017]	6 limbs-spider	600 x 600 x 250 mm	Flat Concave	x	Electromagne	
		12 kg	Convex Cylinder			
PART		0.2 m/min				

- Most existing designs are developed for particular applications with limited functions.
- Most robots provide visual inspection only. Some robots use untouched magnets making them too heavy.
- Most existing designs have a fixed distance between the magnet and surface, and may not work on different types of surface contours. They might be difficult to apply on complicated structures of real bridges that require adaptable, light, and effectively data-collecting robots.
- Drones still get limited with energy issues and can only perform visual/shallow inspections.

- Specifications:
 - ✓ Length 465.5 mm
 - ✓ Width 312 mm
 - ✓ Height 217.3 mm
 - ✓ Weight 12 kg
 - ✓ Load 7kg
 - ✓ Drive:
 - 4 motorized wheels
 - 3 motorized Eddy scan
- Work well on a flat surface



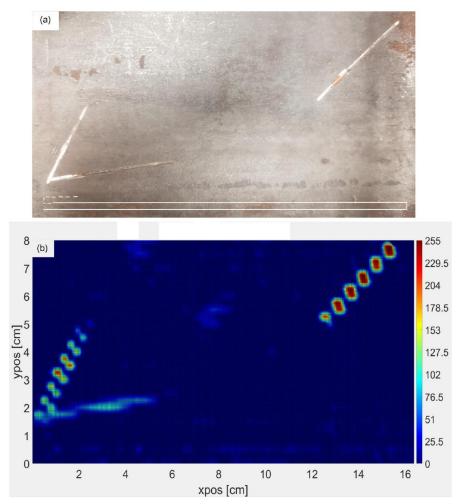
Adhesion Test



Adhesion test: a woman (weight of 64kg) hanging from robot adhered to steel I beam. The robot's adhesion force is significantly stronger than the force that multiple people can exert combined is.

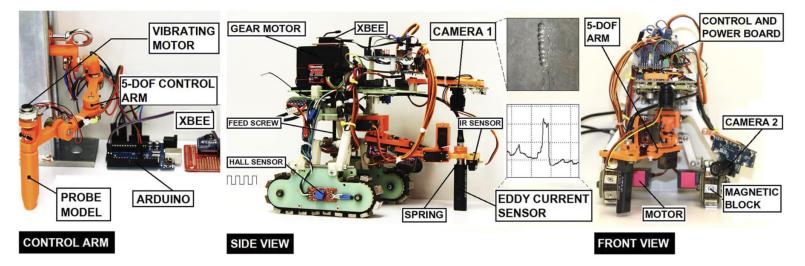


Eddy Current-based Defect Map

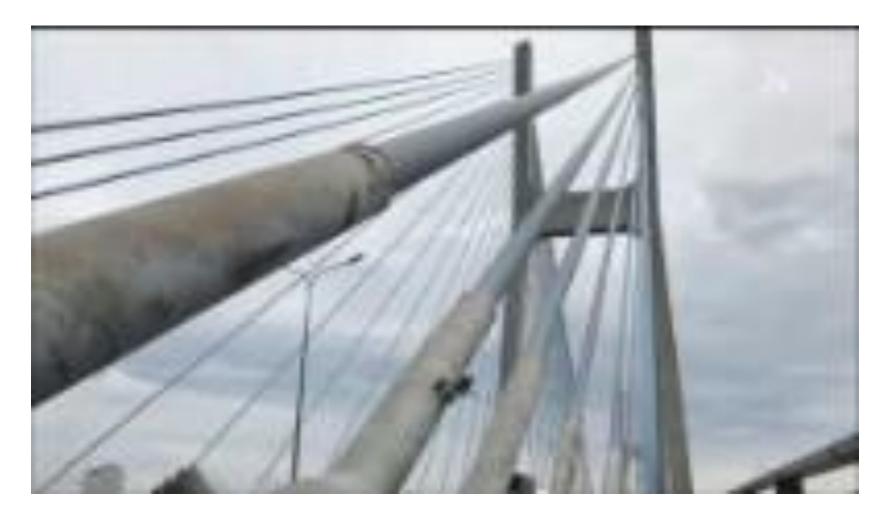


Prototype 2: Tank-Liked Mobile Climbing Robot

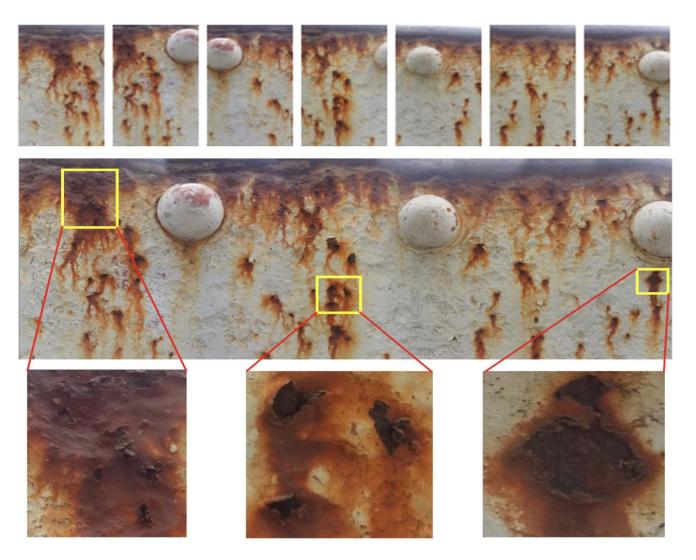
- Specifications:
 - ✓ Length 163 mm
 - ✓ Width 145 mm
 - ✓ Height 198 mm
 - ✓ Weight 3 kg
 - ✓ Drive: 2 motorized roller-chains and 1 motorized transformation
- Work well on both flat and curving surfaces



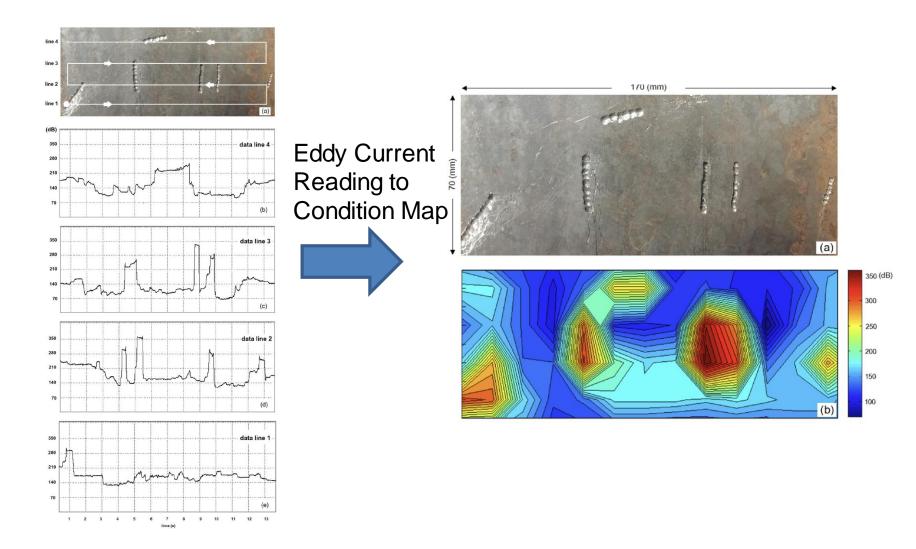
Prototype 2: Tank-Liked Mobile Climbing Robot



Prototype 2: Tank-Liked Mobile Climbing Robot

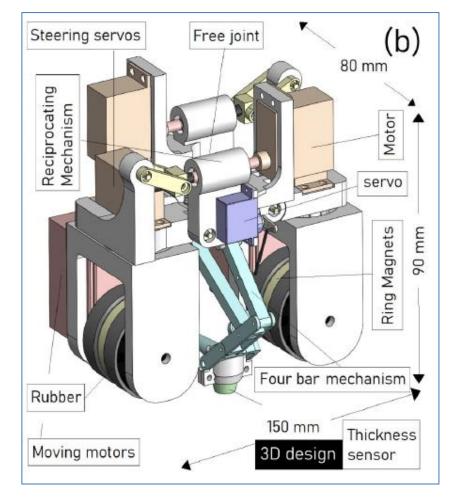


Prototype 2: Tank-Liked Mobile Climbing Robot

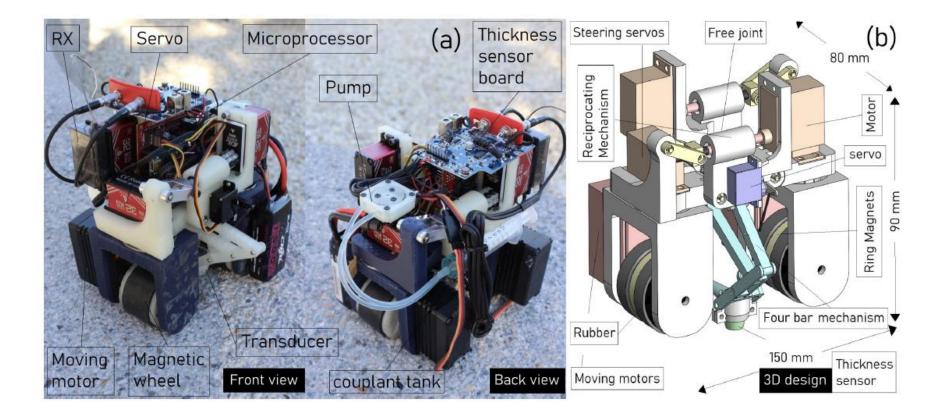


Prototype 3: Bicycle-Like Climbing Robot

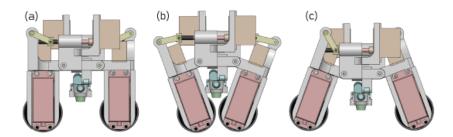
- Specifications:
 - > Length 150 mm
 - > Width 80 mm
 - > Height 90 mm
 - > Weight 1.2 kg
 - > Drive:
 - ✓ 2 motorized wheels
 - ✓ 1 motorized steering
 - 2 motorized reciprocating mechanism
- Work well on both flat and curving surfaces



Prototype 3: Bicycle-Like Climbing Robot



Prototype 3: Bicycle-Like Climbing Robot



The robot's shape when applying reciprocating mechanisms. a) in normal conditions, b) when passing thin edges, c) when passing acute internal corners.

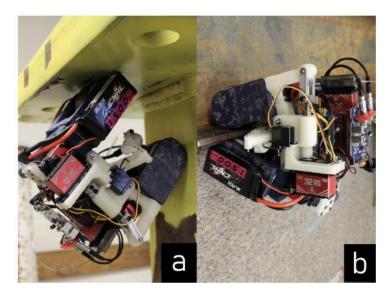


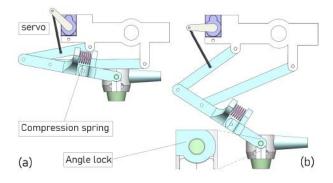
TABLE I

SPECIFICATIONS OF OUR TESTING CONDITIONS.

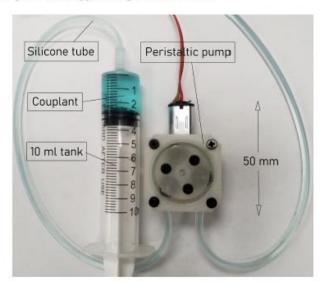
Structural parameters	Dimension (mm)
Thinnest steel surface	2
Smallest steel cylinder diameter	100
Thickest coated paint	3
Highest nut or bolt area	4

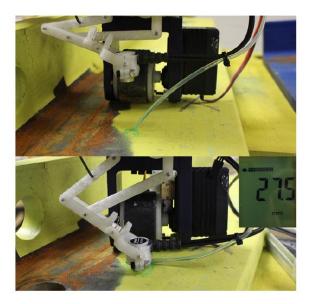
The robot is passing convex and concave surfaces: a) It makes a turn of 90 degrees on internal corners. b) It transforms the wheels to pass a thin edge on a U-shaped beam.

Prototype 3: Sensor Deployment Mechanism



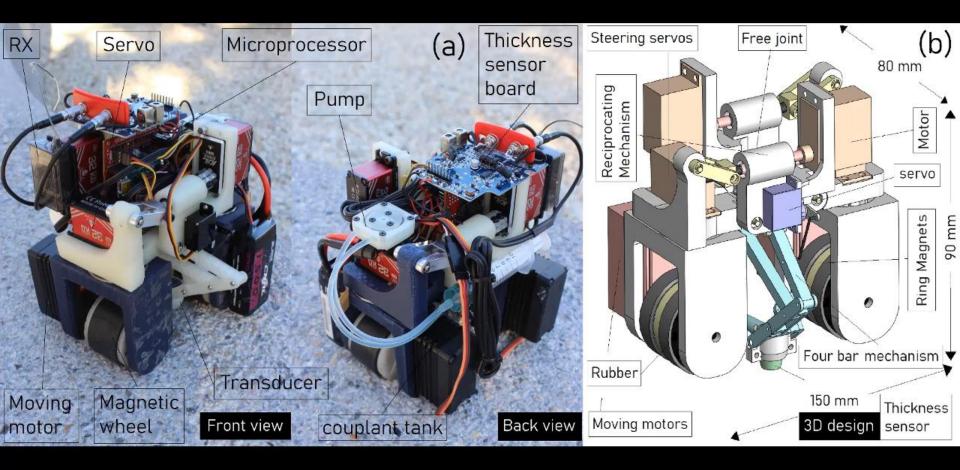
The four-bar mechanism. A compression spring acts as a soft contact with the surface. An angle lock is added to create a free movement of the probe when approaching uneven surfaces.





A demonstration of measuring the thickness of a steel surface. The transducer is well contacted to the surface thanks to the compression spring and angle lock. The final result is averaged over three times.

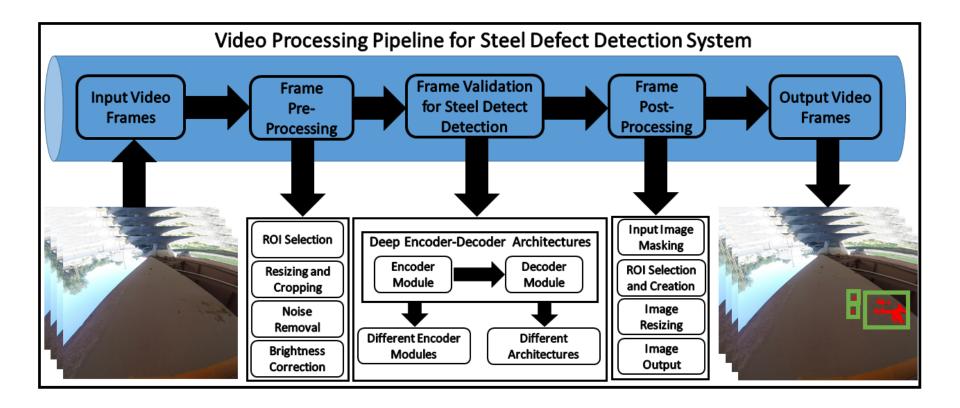
Prototype 3: Robot Climbing Testing



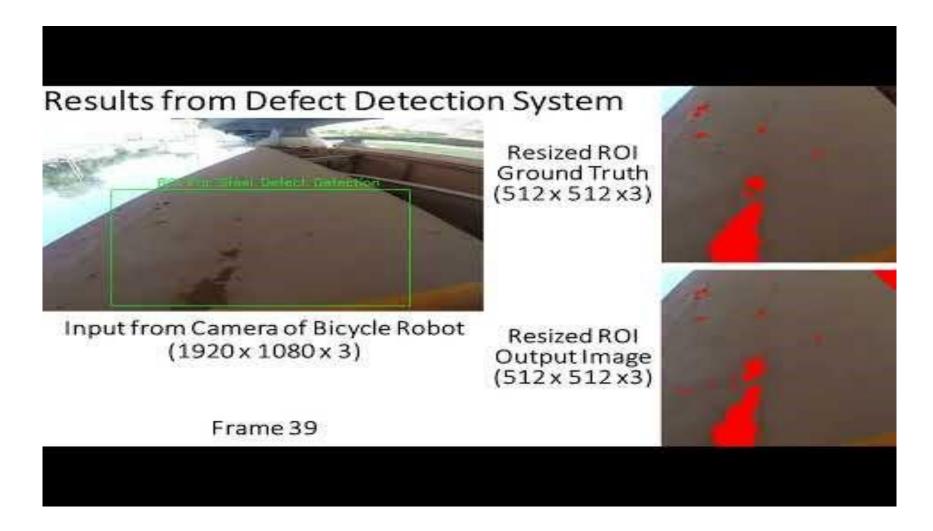
Prototype 3: Bicycle robot with defect detection test



Prototype 3: Bicycle Robot with Steel Defect Detection



Real-time Defect/Rust Detection



Steel Defect Detection

TABLE III

UNet [37] Architecture						
Encoder		Dice	mIOU	Precision	Recall	
		Loss				
ResNet-18 [40]	Max.	31.80	91.86	<u>99.92</u>	91.59	
	Min.	4.37	54.87	99.54	54.86	
	Avg.	12.59	80.88	99.73	81.02	
ResNet-34 [40]	Max.	28.11	96.40	99.83	96.57	
	Min.	1.96	59.40	99.56	59.43	
	Avg.	11.11	83.47	99.72	82.13	
RegNet-X-2 [42]	Max.	18.81	97.13	99.78	99.35	
	Min.	1.59	71.56	99.55	71.71	
	Avg.	7.26	88.01	99.65	87.06	
Efficient-b0 [41]	Max.	32.17	97.33	99.80	97.53	
	Min.	1.41	55.85	99.53	55.92	
	Avg.	11.44	83.26	99.61	83.46	
Efficient-b2 [41]	Max.	47.25	96.06	99.75	96.36	
	Min.	2.18	43.56	99.56	43.60	
	Avg.	14.39	69.84	99.65	81.87	

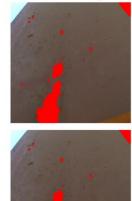
TABLE IV

DeepLab [39] Architecture						
Encoder		Dice	mIOU	Precision	Recall	
		Loss				
ResNet-18 [40]	Max.	26.46	95.50	99.8 0	95.58	
	Min.	2.52	62.24	99.55	62.02	
	Avg.	9.05	86.26	99.68	86.26	
ResNet-34 [40]	Max.	26.46	93.76	99.82	93.82	
	Min.	3.36	61.19	99.55	61.06	
	Avg.	10.45	84.30	99.68	64.14	
RegNet-X-2 [42]	Max.	15.36	97.58	99.78	97.85	
	Min.	1.30	75.79	99.56	75.94	
	Avg.	6.71	89.41	99.69	90.02	
Efficient-b0 [41]	Max.	22.89	96.21	99.59	96.56	
	Min.	1.99	65.52	99.55	65.57	
	Avg.	9.40	85.13	99.55	85.39	
Efficient-b2 [41]	Max.	40.91	90.06	99.85	90.24	
	Min.	5.76	48.12	99.46	48.16	
	Avg.	17.12	75.38	77.17	55.55	

Input Image



Final Output



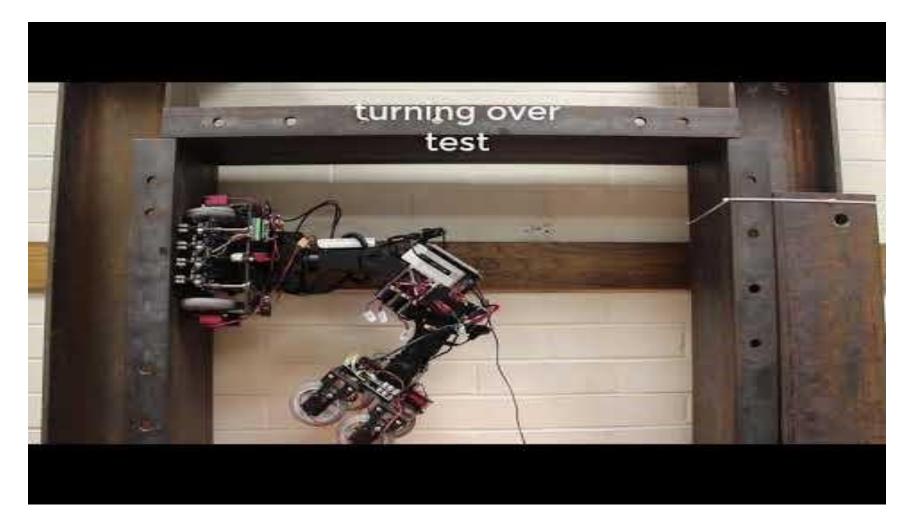




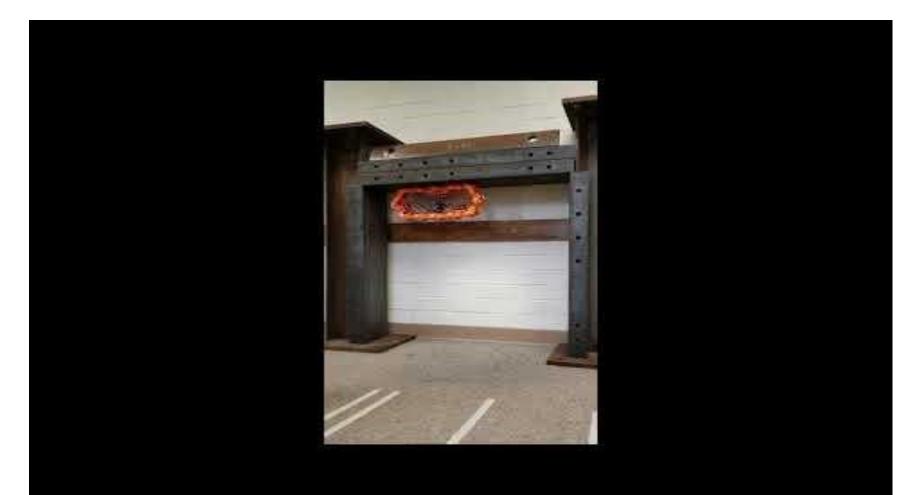


- The first column: original images.
- The second column: ground truths annotated with red color.
- The third column: final outputs of LinkNet ResNet-18, Unet ResNet-18, and DeepLab ResNet-18.

Other climbing prototype



Other climbing prototype



Acknowledgment

- This project is supported by the U.S. Department of Transportation, Office of the Assistant Secretary for Research and Technology (USDOT/OST-R) under Grant No. 69A3551747126 through INSPIRE University Transportation Center (http://inspireutc.mst.edu) at Missouri University of Science and Technology.
- This work is also supported by the National Science Foundation under Hung La's NSF CAREER #1846513 and NSF-PFI # 1919127awards.

TRB Webinar: Robot-Enabled Sensing and Augmented Learning for Bridge Inspection

Tuesday March 29th 2022



Augmented learning through augmented reality and artificial intelligence

Fernando Moreu^{1,2,3,4}

¹Department of Civil, Construction & Environmental Engineering (CCEE) ²Department of Electrical & Computer Engineering (CEC) ³Department of Mechanical Engineering (ME) ³Department of Computer Science (CS) University of New Mexico





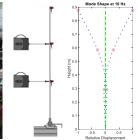






Outdoor Field Monitoring Railroads, Bridges, and Tramways Structural Health Monitoring **Crack Sensing**





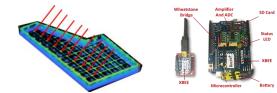


Highlights



Human-Infrastructure Interfaces Human Factors in Engineering





Autonomous Sensing Wireless Smart Sensors **LEWIS Sensors Rockets Sensors**



Cyber Physical Systems Cybersecurity



AI AR for automatic crack measurement

Non-linear Dynamics Earthquakes & Disasters **Random Vibrations**





Human-Infrastructure Interfaces



Human-Machine Interfaces **Emergency Rescue**

Outline

AR Overview

Eye Gazing

Steel Fatigue Crack Finder (SFCF)

Concrete Crack Characterization

Vibration Monitoring in AR

Robot Control Application for sensor placement with AR

Conclusions



AR Overview



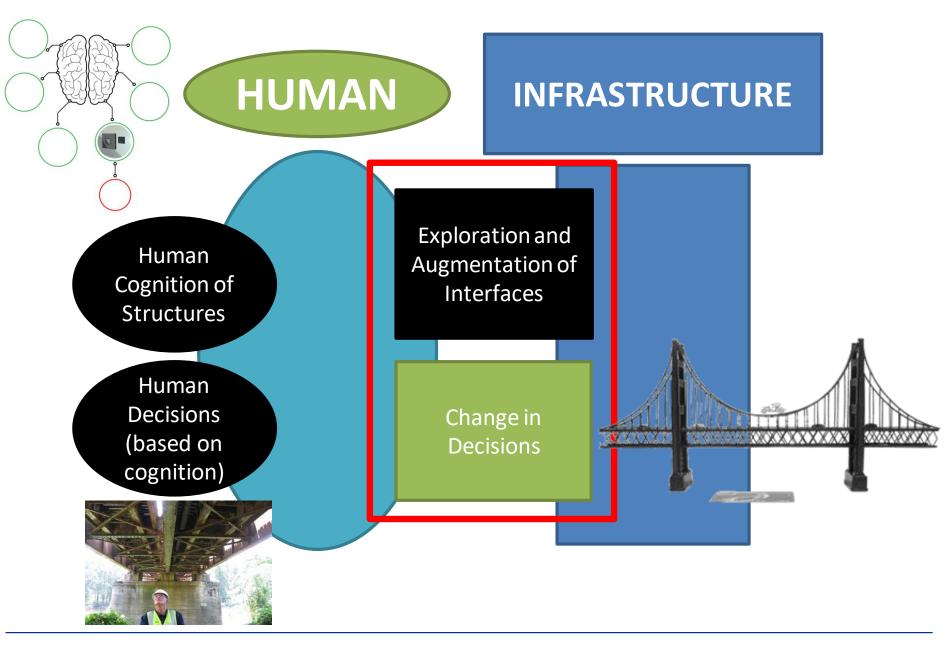
Overview of using AR: what the inspector sees



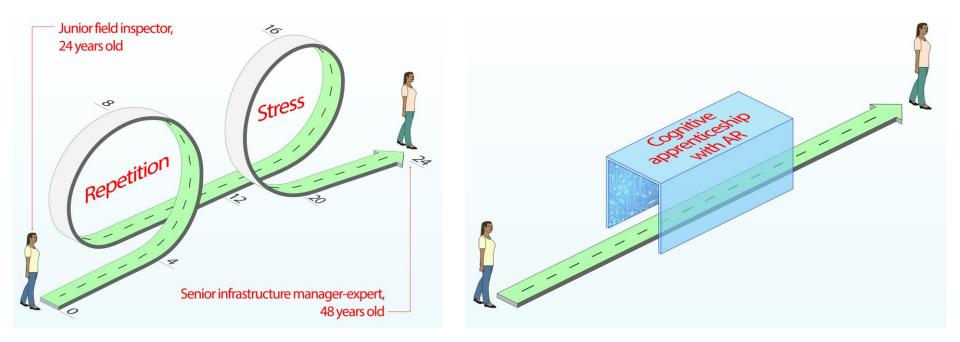
Overview of using AR: rail detection



Slide 57



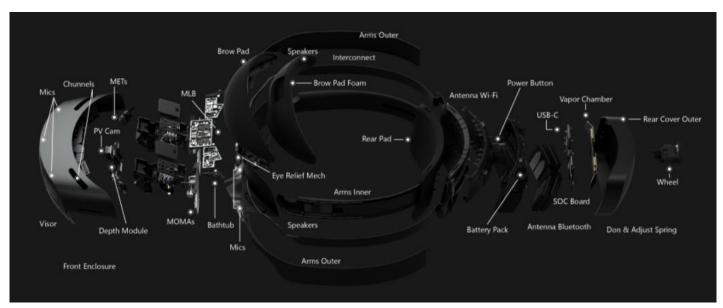
New Learning in Engineering



Augmented Reality – Head Mounted Device (HMD)



Microsoft HoloLens 2



Microsoft HoloLens 2 features

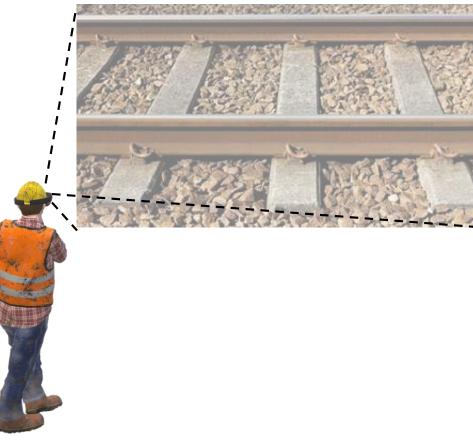
Image courtesy of Microsoft https://docs.microsoft.com/en-us/hololens/hololens2-hardware



Eye Gazing for inspections

Visual Inspection Quality

 Important to ensure comprehensive visual coverage of the rail track.



EyeRR Software for inspection accuracy quantification

 Quantifies the inspection accuracy and records inspector's visual coverage during the inspection.



EyeRR's application in rail track inspection

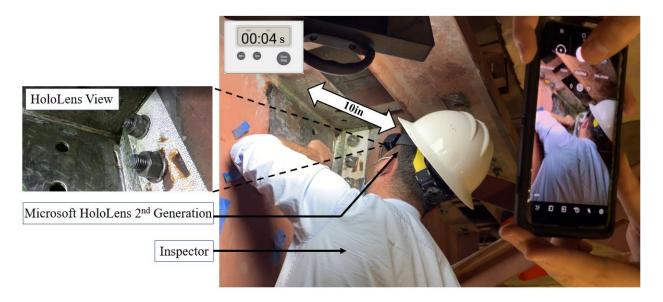


Steel Fatigue Crack Finder (SFCF)

Fatigue Cracks

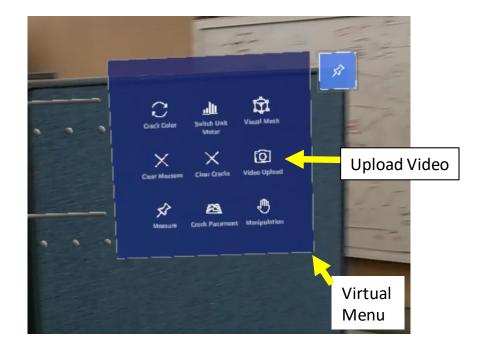
Those are Important because they are:

- Hard to find
- Could propagate and contribute to bigger problems.



Steel Fatigue Crack Finder (SFCF) software

Improves the inspector's perception during the inspection to find tiny fatigue cracks



SFCF's indoor test



SFCF's test on a steel structure



Preprocessed

Post-Processed

Concrete Crack Characterization

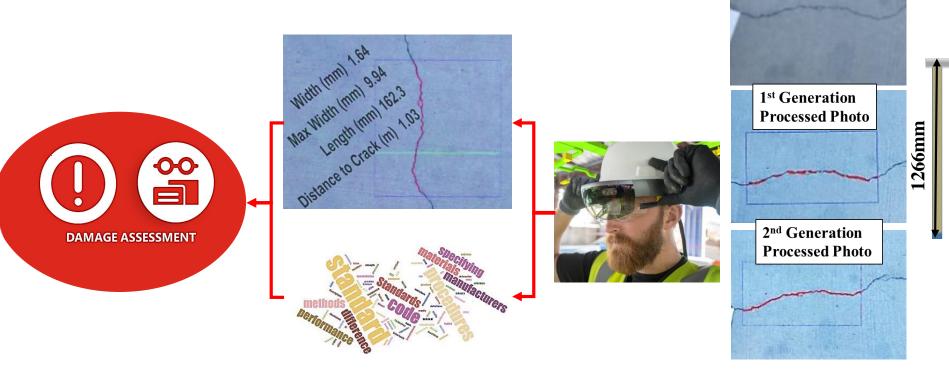
Objective

This project seeks to provide the inspectors with :

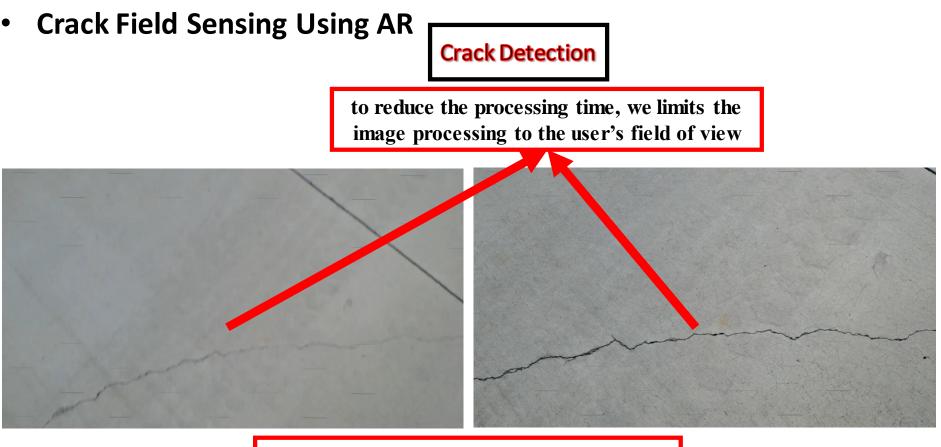
- A crack-detection assistant tool for concrete structure inspection
- A faster and more accurate substitute for traditional crack measurement

Unprocessed Photo

• An estimate damage assessment and prognosis tool



Results

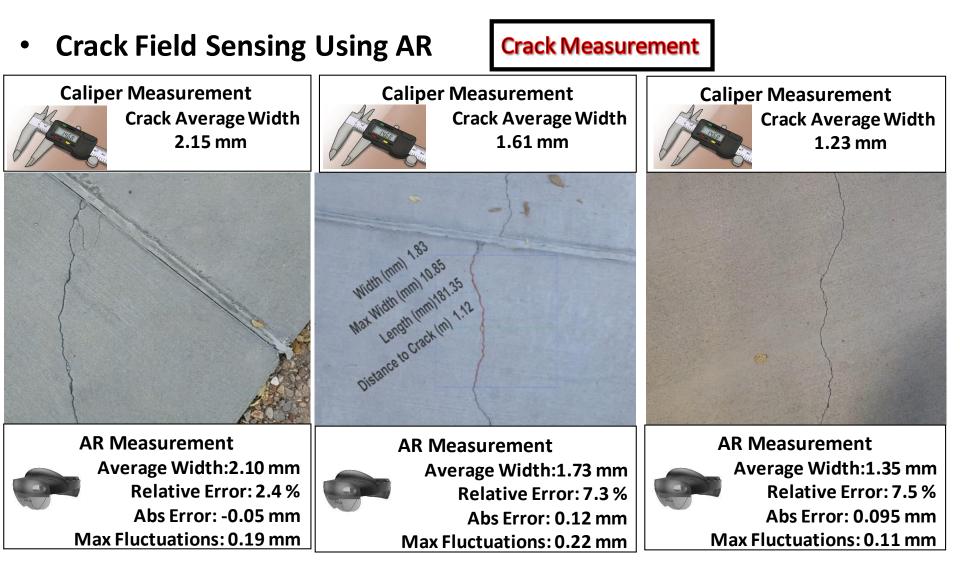


processing time $\cong 1.5$ s

by applying a simplification Kernel to the algorithm, the processing time is less than 0.8 in the newer versions

processing time $\cong 1.5s$

Results



Usability Evaluation

• Evaluation with bridge engineers from NMDOT

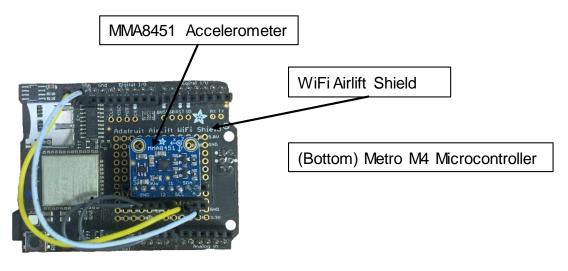




Vibration Monitoring in AR

Measuring Vibrations and AR

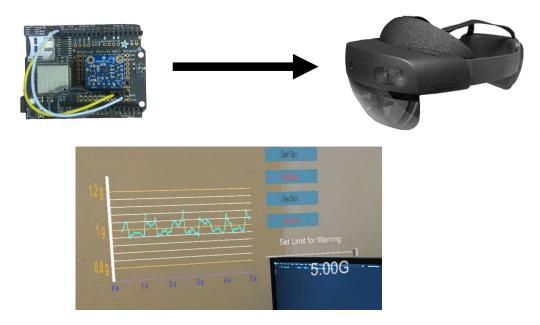
- The LEWIS5 (Low-cost Efficient Wireless Intelligent Sensor) is built with a Metro M4 microcontroller programmed in Arduino.
- Sensor is equipped with an accelerometer to read acceleration data and a WiFi shield for wireless capabilities, which allows it to connect to the HoloLens. Requires a power source connected via micro-USB.



LEWIS5 Sensor

Measuring Vibrations and AR

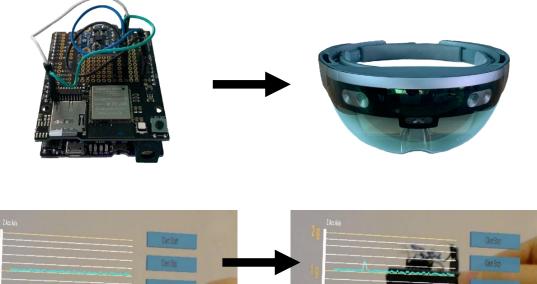
- AR can be used as an interface for sensor feedback to enable a higher level of understanding of dynamic events and experiments.
- An application is developed to send acceleration measured by the LEWIS5 to the HoloLens over WiFi and the data is plotted in the AR interface.

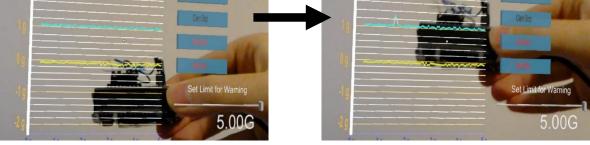


LEWIS5 Sensor sends data to HoloLens and data is plotted in AR

Summary - Overview

- Augmented Reality (AR) can provide a mode of amplifying human cognition of structural response.
- Create AR application to send and graph sensor data in AR headset close to real-time.
- Problem statement: Time delay is a barrier to implementing the app.
- Quantify and compare time delay over a local WiFi network versus a mobile WiFi hotspot.





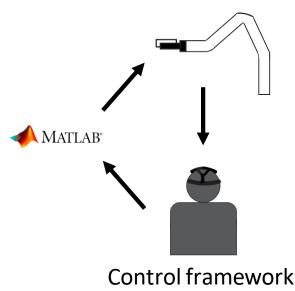
Time delay in sensor movement and response in graph

Robot Control Application for sensor placement with AR

- Human intuition often allows humans to solve various tasks faster than robots.
- When these human capabilities are coupled with the repeatability and endurance of robots, full potential can be realized.
- A 7-degree-of-freedom robotic arm (kinova gen3) can be controlled in AR through a pipeline of Hololens and MATLAB code.

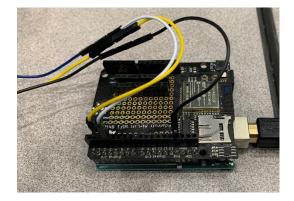


Kinova gen3



 A preliminary AR application is developed to define a set path for pick-and-place commands for an older model of robot (Cyton Alpha).

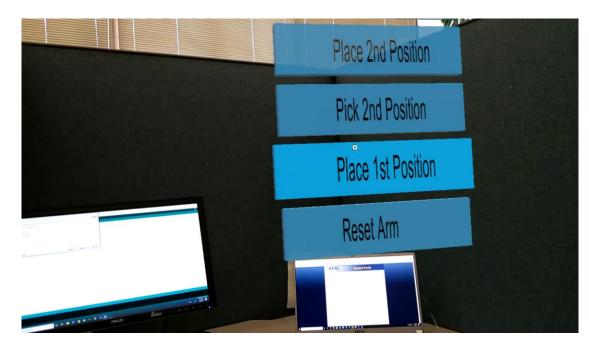




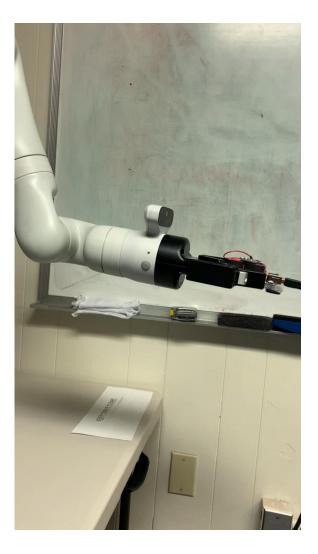


Cyton Alpha Arduino board with WiFi shield Power supply and control box

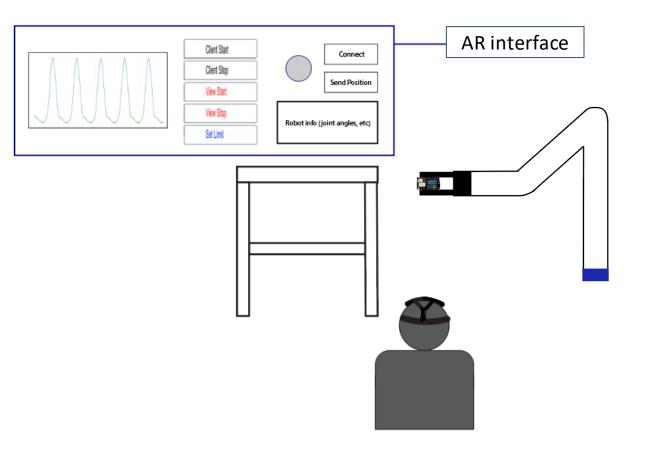
• In the AR interface the user has the option of selecting the position and pick or place. The robot reacts to command, which is sent over WiFi from Hololens to the board.



 The robot arm is capable of moving and placing sensors both securely and safely.



• Combining the kinova gen3 consistency and durability with sensors and human intuition enables high-level control for application to manufacturing, sensing, inspection, and other tasks related to railroads.





Conclusions

New advances in technology can contribute to objectively inform humans of their surroundings

The future bridge engineer, manager, inspector will be collecting and analyzing data to inform decisions enhanced by human-infrastructure new interfaces

Our future inspector will see/measure/collect data faster and change decisions about their surroundings

Research opportunities for multidisciplinary engineers interested in AR and Human-Infrastructure Interfaces