



Al-Enabled Robots for Automated Nondestructive Evaluation and Repair of Power Plant Boilers

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Background

- Boiler inspection and repair is one of the main focuses during power plant overhauls.
- Boiler failures can:
 - cause loss of life and safety issues
 - cost hundreds of thousands of dollars in equipment repairs, property damage, and production losses
 - drive up the cost of electric power
- Failures can be prevented by performing regular maintenance.

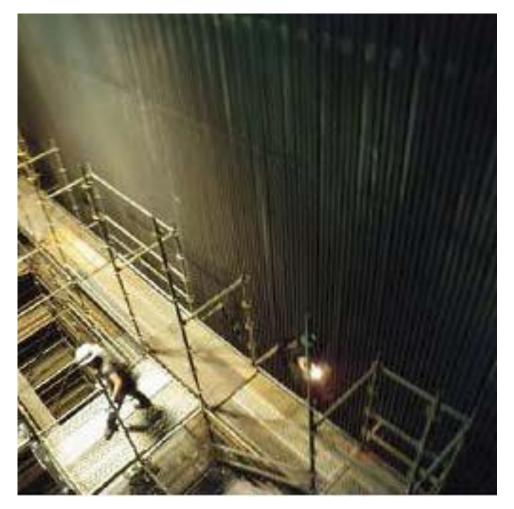


The coal-fired Hayden generating station in Colorado

Introduction

Challenges

- Boiler maintenance is challenging and dangerous on scaffolding in the confined space inside a boiler.
- The operation is time-consuming due to the large area of vertical structures and tremendous effort needed for scaffolding.
- E.g.: maintaining 50,000 ft2 of furnace walls (water walls) during a single overhaul takes two weeks to complete, while scaffolding itself costs \$100,000.



Hayden Unit 1 scaffolding for manual inspection and repair



Purpose of Project

• The project goal is to develop integrated autonomous robots that are equipped with nondestructive evaluation (NDE) sensors to perform real-time inspection of boiler furnace walls, operate repair devices to achieve live repair, and use artificial intelligence (AI) to enable smart data analysis and autonomy.

| Current Practice | Limits |
|-----------------------------|---|
| Manual maintenance | Safety risks caused by environment hazards and climbing up scaffolding, time- consuming to mount/dismount scaffolding, often slow and inaccurate inspection. |
| UAV-based inspection | Limited payload and operation time, requirement of human control (auto-pilot not feasible yet), dependence on wireless communication, typically only visual inspection (nondestructive inspection and repair not feasible yet). |
| Robotic inspection crawlers | Typically incapable of repair, requirement of constant remote control (no full autonomy yet), no AI for smart autonomy and predictive analysis. |



Strategic Alignment to FE Objectives

• This project well aligns with the Fossil Energy Objective 2.2:

Advance technologies to improve the efficiency, reliability, emissions, and performance of existing fossilbased power generation,

by developing "... advanced sensors and controls to help increase coal plant efficiency, reduce forced outages, and avoid downtime related to equipment failures..."

Using Artificial Intelligence in Fossil Energy R&D

Office of Fossil Energy

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When you hear about **artificial intelligence (AI)**, the first technologies that probably come to mind are voice-activated assistants like Alexa or Google Home, self-driving cars, and drone delivery programs. If you think these innovations are exciting, then we can't wait to share how **we're using AI here at the Office of Fossil Energy** (FE).

Heralded as the next technological revolution, AI makes it possible for machines to learn from experience and perform tasks commonly performed by human beings. The integration of AI into our research and development (R&D) efforts is transforming FE's ability to analyze massive datasets and solve complex problems.

We have over 60 AI-enabled projects underway. **Check out these 5 examples** that show how AI R&D is utilizing and protecting our Nation's vast fossil energy resources:

1. Smart Robots that Inspect and Repair Power Plant Boilers

Power plant boilers are the most important part of a power plant, but they are difficult and timeconsuming for human operators to inspect and repair. To reduce risks and shorten maintenance and unplanned outages, FE is developing Al-enabled robots that can perform real-time, nondestructive inspection of boiler furnace walls. If they find a crack, they can operate repair devices to make an immediate repair, while using Al to enable smart data analysis and autonomy.

Project Objective Overview

- NDE: Designing nondestructive evaluation sensors for crack detection.
- Repair: Designing repair tool and control protocols for crack repair.
- Robotic platform: Integrating NDE sensors and repair tools on a mobile robot.
- Artificial intelligence: Designing AI methods for smart data analysis and autonomy.

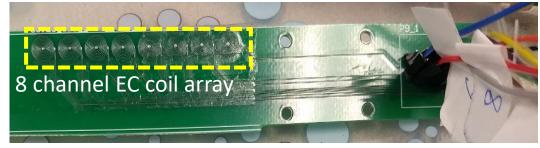
| Objectives | Lead | 2018 2019 | | | | 2020 | | | | | 2021 | | |
|-------------------------|-------------------|-------------------------|---------|---------|-----|-----------------------|----------|----|----|--------------------|------|----|----|
| | | Q4 | Q1 | Q2 | Q3 | Q4 | Q1 | Q2 | Q3 | Q4 | Q1 | Q2 | Q3 |
| NDE Sensing and | Dong & Udna (MSU) | NDE probe design | | | | Signal processing | | | | | | I | |
| Assessment | Deng & Udpa (MSU) | | | | | | | | | | | | |
| Repair Device Design | Vu & Lin (CSM) | Integrated repair sys. | | | | Repair protocol | | | | | | | |
| and Control | | | | | | | | | | | | | |
| Robotic Platform | Petruska (CSM) | Robot retrofit for Clea | | | | aning NDE integration | | | | Repair integration | | | |
| Design | × / | vertical motion | | | des | sign | on robot | | | on robot | | | |
| Artificial Intelligence | Zhang (CSM) | 3D mapping and | | | | Spatiotemporal | | | | Predictive damage | | | |
| Development | | int | formati | on fusi | on | damage tracking | | | | analysis on robot | | | |



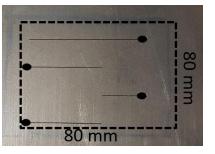
Project Achievements: NDE Sensor

NDE Probe Design (Task 1.1)

• An eddy current (EC)-based sensor array was designed which showed high scanning speed and detection accuracy.

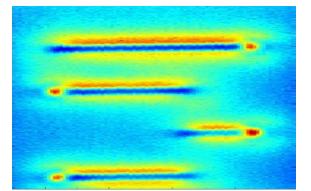


Diameter of the coil is 0.5 (unit: cm)



Steel Sample (A108): The width of the fabricated cracks is 200 µm and 400 µm

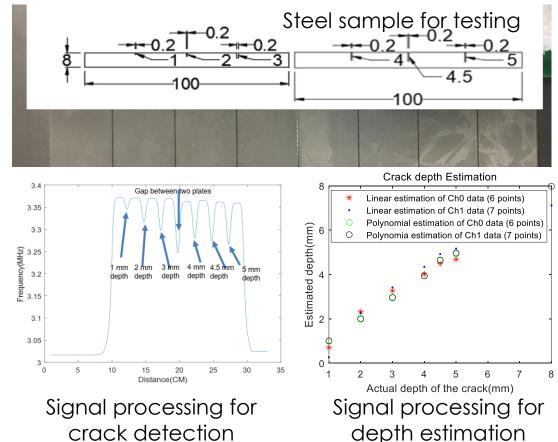
robotics lab



Sensor array scanning result

NDE Signal Processing (Task 1.2)

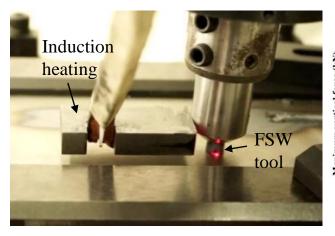
• Signal processing methods that extract crack parameters for localization and size estimation were designed.



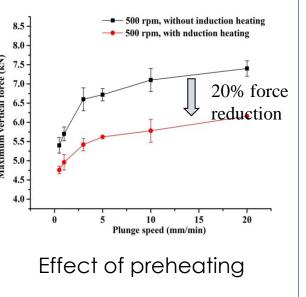
Project Achievements: Repair Tool

IHA-FSW Repair Tool (Task 2.1)

- An induction heating (IH) assisted friction stir welding (FSW) repair tool was designed, implemented, and tested in collaboration with EnergynTech Inc.
- IH serves as a preheating source that reduces the vertical force during FSW process by about 20%.

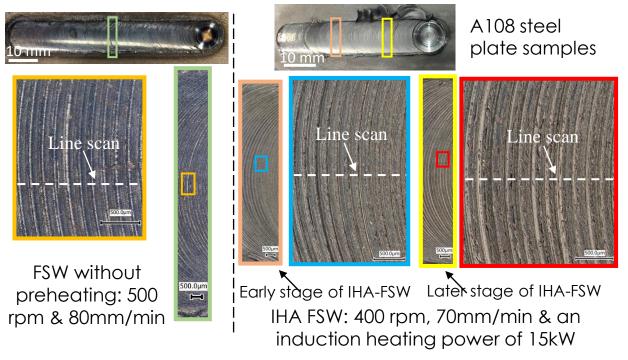


Side View of the IH-FSW repair tool



Repair Protocol (Task 2.2)

• Testing repair conditions were determined through surface characterization, that are <u>500</u> <u>rpm and 80 mm/min without preheating</u>, and 400 rpm and 70 mm/min with 15kW induction heating power.



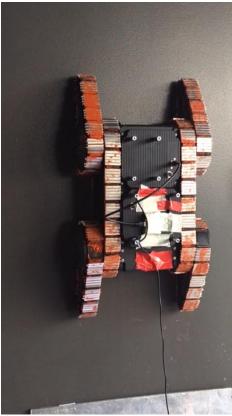


Project Achievements: Robotic Platform

Robot Retrofit (Task 3.1)

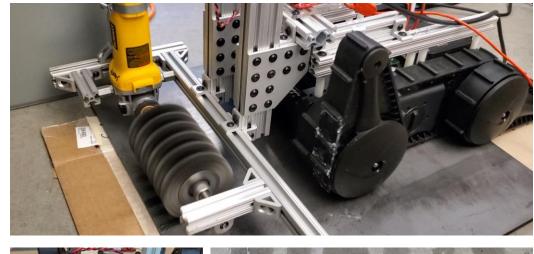
We modified a tracked mobile robot to include magnets on track surfaces to attach to a vertical metal surface.





Cleaning Tool Design (Task 3.2)

• We developed a brush cleaning mechanism for metal surface cleaning by modifying offthe-shelf cleaning tools.





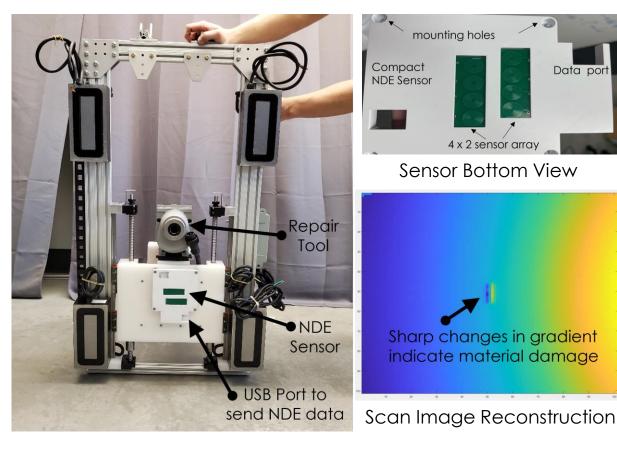




Project Achievements: Robotic Platform

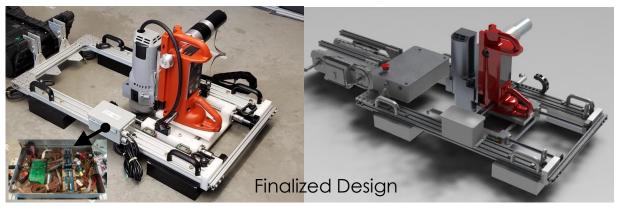
NDE Sensor Integration (Task 3.3)

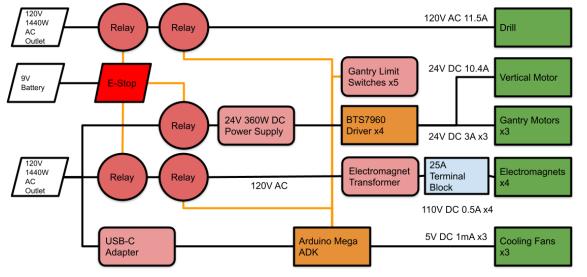
• An 8-channel NDE sensor is integrated to the bottom of gantry system and scans along two dimensionalities at a sub-millimeter precision.



Repair Tool Integration (Task 3.4)

• A repair tool and a 2-dimensional tool controller are integrated with the gantry system.



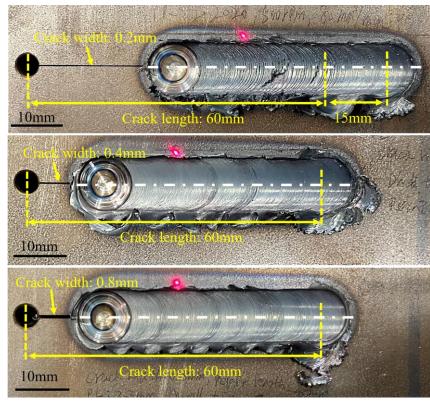


Repair Tool Schematic and Wiring Design

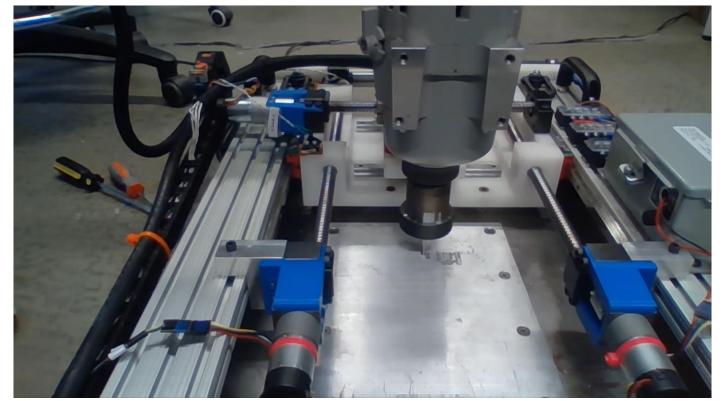
Project Achievements: Robotic Platform

Onboard Demonstration and Offboard Repair Evaluation

• Using the tool with a diameter of 10mm, <u>the repair depth is up to around1.6mm, and the</u> width from 0.2mm up to 0.8mm on the A108 steel plate.



Off-board Evaluation



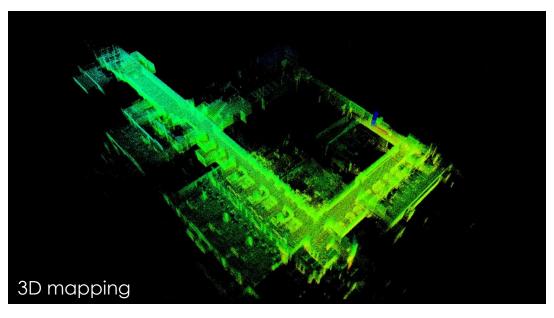
Onboard Demonstration



Project Achievements: Artificial Intelligence (AI)

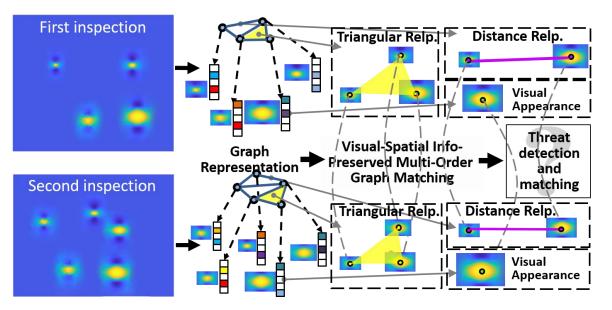
3D Mapping & Info Fusion (Task 4.1)

- Iterative closest point (ICP) algorithms build multi-layer maps by matching and overlaying adjacent frames of 3D point clouds from LiDAR.
- The method has been evaluated initially in dark, confined environments with sparse features.



Spatiotemporal Tracking (Task 4.2)

- We developed a voxel-based representation learning method for robots to recognize places, correct mapping errors, and perform localization during execution.
- We are developing methods to represent landmarks as a graph to fuse visual-spatial information for crack tracking across inspections.

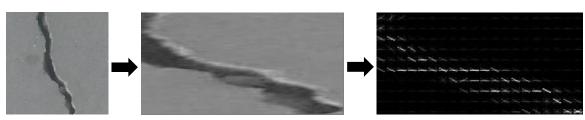




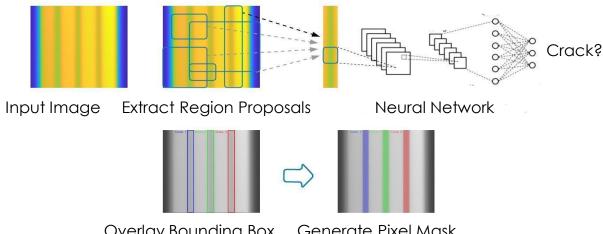
Project Achievements: Artificial Intelligence (AI)

Machine Learning for Crack Recognition (Task 4.3)

- We developed feature extraction and machine learning methods, e.g., Convolutional Neural Network (CNNs), to recognize simulated cracks.
- The methods obtained > 95% precision and recall.

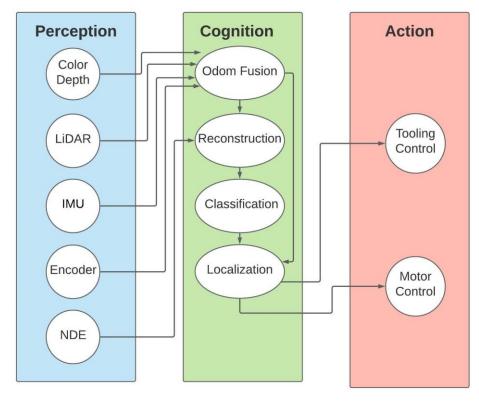


Shape features: Histogram of Oriented Gradients (HOG)



Overlay Bounding Box Generate Pixel Mask

Software System Architecture



- Perception: Sensing and data acquisition.
- Cognition: AI, robot localization and damage tracking.
- Action: Control of robot motors and tooling machinery.

Educational Impacts

- This project provided opportunities to faculty, graduate students, and undergraduate students for interdisciplinary research in fossil energy fields.
- Research topics from this project were also used as motivating examples in the PIs' courses (e.g., as a topic for final course projects).
- 2020-Dr. Zhenzhen Yu and Dr. Hao Zhang were promoted to Associate Professor with tenure at the Colorado School of Mines.
- Publication: Related research topics from this project was published at multiple conferences on NDE and robotics, including ENDE, ISEM, ICRA and IROS, and by the journal of Materials Evaluation.
- 2019-Best Poster Award: Xiaodong Shi, Zachary Nahman, Fares Alharbi, Ciaron Hamilton, Yiming Deng, and Hao Zhang, "Al-enabled Robotic NDE for Structural Damage Diagnosis and Mapping", Int'l Symposium on Applied Electromagnetics and Mechanics, 2019.



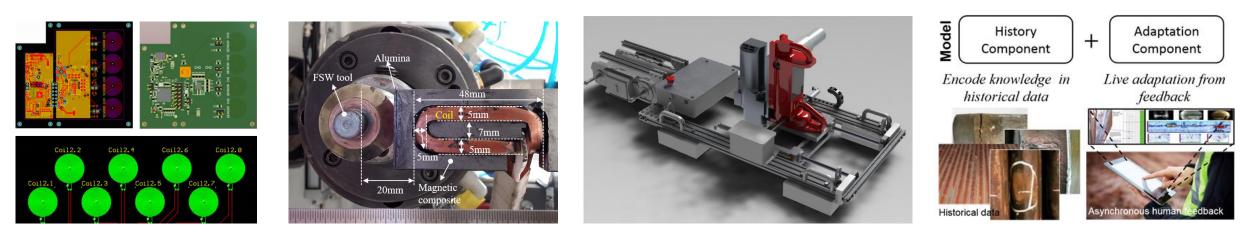
Insights and Future Work

- Designing a portable repair tool and integrating it with the autonomous robot is the biggest challenge. Further development is still required to mature the tool to address the complex geometry of furnace walls.
- Data, data, more data!: Machine learning models, especially deep learning methods, require big data for training. Obtaining real-world large-scale data of boiler anomalies (e.g., cracks) is a big challenge.
- Many future works must be done to continue this basic research in order to move the developed robotic system closer to real-world deployment, and to address many additional challenges, such as:
 - Multi-sensory fusion to integrate NDE data from multiple sources to improve assessment accuracy.
 - Design of robot motions to coordinate the sensor and the repair tool for real-time integrated and iterative assessment and repair.
 - Navigation planning for the robot to intelligently cover all furnace walls in a boiler.
 - Customer discover to understand the project's impacts, such as time and cost savings, reduced risk for humans, and increased boiler reliability, usability, and efficiency.



Concluding Remarks

 This project developed an autonomous robot that is equipped with NDE sensors to perform real-time inspection, operate repair devices to achieve live repair, and use artificial intelligence to enable smart data analysis and autonomy, in order to address cracks on boiler furnace walls.



NDE Sensing and Assessment Repair Device Design and Control Robot Platform Integration

Artificial Intelligence

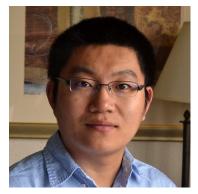


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Thank you!



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