#### UAV-Based Sensor Deployment and Edge Computing for Rapid Infrastructure Assessment

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# The ARTS-Lab at USC

#### foundational science We use



to develop essential tools



to solve real-world problems



Day School **Dan Thompson Dan Thompson Dan Thompson public domain** 

#### **We are Engineers (mostly)**

#### Sensing **Constitution** Data Assimilation dielectric  $(SEBS + TiO<sub>2</sub>)$ copper tape DAQ<sup>'</sup> electrode (SEBS + CB) data synthesis and you host computer misclassifier ignorable defec  $3B$ ₫  $2B$ real-time 18 target Month los machine-learned e(y,v) trainin point AI/ML AI/ML AI/ML AI/ML AI/ML Embedded Systems









In Situ Monitoring of AM Water Quality Sensors Geo Technical Sensors











#### Flexible Electronics Nuclear Magnetic Resonance Vibration Sensors

signal roading & amplific











# Data Assimilation















 $\frac{-1}{\alpha} = \sum_{r=1}^m \frac{v_r^2}{\omega_r^2 - \Omega_r^2}$ 



Civil Structures **High-Rate Systems** Battery Systems



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# Embedded Systems





7 Microcontroller/ microprocessor Real-Time OS FPGA













#### **Outline for Today's Talk:**

- 1. Driving Challenges in UAVdeployed Sensors
- 2. UAV and Sensor Hardware
- 3. Edge Processing of Sensor **Signal**
- 4. Networks of UAV-deployed Sensors
- 5. Future Directions and Preliminary Results
- 6. Conclusion



# Driving Challenges in UAV-deployed Sensors

## **Challenges and Innovations in Structural Health Monitoring**

- **Current Limitations:** Traditional SHM relies on specialized equipment and skilled personnel, limiting speed and flexibility.
- **Deployment Challenges:** Remote, hazardous locations, or decaying structures add time, cost, and safety risks to manual sensor deployment.
- **Need for Rapid SHM Solutions**: Realtime, data-driven insights, autonomous sensor deployment, and effective wireless communication are critical for safe, efficient monitoring.



## **Our Solution – UAV-Deployable Sensor Package**

- **Rapid Aerial Sensor Deployment:**  Designed for quick, efficient sensor placement in SHM scenarios
- **Enhanced Spatial Awareness:** Multiple camera views for precise navigation, docking, and sensor deployment
- **Electropermanent Magnetic Docking:**  Secure attachment with a recovery cone for guided docking
- **Built-in Redundancy:** Safety and reliability features to ensure successful deployments





#### **Understanding the Ceiling Effect in UAVs**

- **Definition:** The ceiling effect occurs when a propeller operates near a barrier, like a ceiling, altering the airflow and making lift more efficient.
- **Cause:** Impeded airflow above the propeller leads to a pressure drop, creating an increase in lift.
- **Impact on Control: The UAV** operator may notice sudden, unexpected lift or reduced control near the ceiling.









# **Challenges in Human-Operated Flight for Sensor Deployment in SHM**

- **Ceiling Effect Variability:** Sudden lift changes near ceilings
- **Pilot-Induced Instability:** Oscillations from manual control
- **Signal Interference:** Issues near metal structures
- **Line of Sight Limitations:** Restricted visibility impacts precision

No researchers were harmed during this endeavor!



# UAV and Sensor Hardware

#### **Sensor Package**

- **Deployment System:** Uses a 3D-printed recovery cone for guided docking
- **Integrated Streaming: Provides multiple** camera views for precise navigation
- **Electropermanent Magnets:** Secure sensor placement and retrieval
- **Error Compensation: Redundancy** measures for safe, reliable operation in complex environments





#### **Deployment and Retrieval System**



#### **Deployment and retrieval system**

- **Electromagnetic Activation:** Pulseactivated magnetic polarity control
- **Energy-Efficient: Holds magnetic** state without continuous power
- **Versatile Applications:** Ideal for clamping, lifting, and sensor deployment
- **Stable Magnetic Configuration:** Maintains position securely using South-South or South-North fields



#### **Sensor hardware and onboard systems**

- **Robust Design:** Aerially deployable with noninvasive EPM docking
- **Reliable Operation:** Power management, nonvolatile memory, and wireless communication
- **Sensing:** Accelerometer up to 28 kS/s; frame minimizes transmissibility loss





#### **Sensor Package System Architecture**

- **Core Processing:** Teensy 4.0 microcontroller (ARM Cortex M7) with SD card for data storage
- **Communication:** Highsensitivity accelerometer and RF module for real-time data and commands



# Edge Processing of Sensor Signal

## **LSTM-Based Signal Compensation Process**

- Model training procedure
- Supervised learning method
- Assumptions:
	- Sampling rates were set equal (400 S/s)
	- Zero phase between the two sensors
	- Bandwidth of interest to be < 10 Hz
- Model chosen is a single-layer 50-unit LSTM
- Backpropagation is done online every 400 datapoints (1 second)

$$
f_t = \sigma(W_f x_t + U_f h_{t-1} + b_f)
$$
  
\n
$$
i_t = \sigma(W_i x_t + U_i h_{t-1} + b_i)
$$
  
\n
$$
o_t = \sigma(W_o x_t + U_o h_{t-1} + b_o)
$$
  
\n
$$
\tilde{c}_t = \tanh(W_c x_t + U_c h_{t-1} + b_c)
$$
  
\n
$$
c_t = f_t \circ c_{t-1} + i_t \circ \tilde{c}_t
$$
  
\n
$$
h_t = o_t \circ \tanh(c_t)
$$
  
\n
$$
y_t = W_d^T h_t + b_d
$$



update

state

update

state

# **Signal Conditioning and Error Compensation**

- Chirp excitation is fed into the electromagnetic shaker using an analog output module
- A data acquisition is used to record reference acceleration
- A digital trigger is set to synchronize both the reference accelerometer and sensor package
- Various dynamic ranges were used to expand the training range of the LSTM model

$$
x(t) = \sin\left(2\pi \left(\frac{f_{\text{end}} - f_{\text{start}}}{2(\text{test time})}t^2 + f_{\text{start}}t\right)\right)
$$

$$
\text{SNR}_{\text{dB}} = 10\log_{10}\left(\frac{\sum_{i=1}^{\text{data length}}(\text{signal}(i))^2}{\sum_{i=1}^{\text{data length}}(\text{noise}(i))^2}\right)
$$

RMSE = 
$$
\sqrt{\frac{\sum_{i=1}^{\text{data length}} (\text{truth}(i) - \text{prediction}(i))^2}{\text{data length}}}
$$



#### **LSTM compensator performance**

- For testing a chirp excitation in 0-5 Hz is used
- SNR<sub>dB</sub> enhancement of 9.34%
- RMSE reduction of 19.66%
- Usable bandwidth  $( $\pm 2\%$ )$  is shown to increase form 2.78 Hz to 1.34 Hz
- An overall increase in gain below 0.9 Hz due to training bias





# Networks of UAV-deployed Sensors

# **Sensing network for SHM**

#### **Wireless system:**

- protocol: Enhanced **ShockBust**
- bandwidth: 2.4 GHz
- data rate: 2 Mbps
- RF links: 6 channels





# **Case Study – Modal Detection**

- Active modal detection using UAVdeployable sensing network
- data acquisition and real-time synchronization
- signal processing and state estimation





response

## **USC Waking Bridge**

- **Objective:** Track and analyze bridge health using UAV-deployed sensors
- **Finite Element Analysis:** Model properties, natural frequencies, and mode shapes









Mode 2 - 22.84 Hz

## **Time Response of a Pedestrian Bridge**

- **Test Setup:** Mass shaker and UAV-deployed sensors positioned at nodes A0, A1, and A2
- **Objective:** Capture acceleration data across key nodes for modal analysis
- **Results:** Real-time acceleration data reveals dynamic response patterns across nodes





## **Frequency Response of a Pedestrian Bridge**

- **Setup:** Data collected from three sensor nodes deployed on bridge
- **Frequency Sweep:** 0 to 15 Hz to capture modal responses
- **Peak Detection:** Observed resonance at ~11 Hz indicating first flexural mode



# **Sensing system experimental challenges**

- **Latency Threshold: Set at 10** µs to maintain signal alignment across sensors
- **Latency Results:** 85% of instances fell below the threshold, improving synchronization
- **Low-Frequency Detection:**  Limited by buffer size; mitigation options include extended sampling and data combination



Future Directions and Preliminary Results

#### **Camera-Assisted Navigation and Alignment**

- **Multi-Camera Setup:** Provides real-time spatial awareness for precise navigation.
- **Target Identification:** Assists in locating the sensor package with visual feedback.
- **Accurate Alignment:** Guides the UAV to align the recovery cone with the sensor package.
- **Foundation for Autonomy:** Key step towards a fully autonomous UAV system.



## **Camera-Assisted Navigation and Alignment**



## **Experimental Insights and Preliminary Success**

- **Camera-Aided Navigation:** Enhances spatial awareness during approach, deployment, and retreat.
- **Efficiency Gains:** Camera-assisted deployment reduces time across all stages.
- **Improved Precision:** Minimizes failed approaches and unintended contact.







## **Future Work**

- **Automate Delivery:** Enable precise, autonomous sensor deployment and retrieval with minimal human input.
- **Improve Streaming: Boost video** quality for better navigation.







# Conclusion

# **Conclusion**

- **Key Achievements:**  UAV-deployable sensor with autonomous alignment
- **Enhanced Data Quality:**  Real-time LSTM signal compensation
- **Future Goals:** Full autonomy and robust field deployment



# Questions and Discussion

#### **Key GitHub Repositories**

- **Sensor Package:** [https://github.com/ARTS-Laboratory/Drone-](https://github.com/ARTS-Laboratory/Drone-Delivered-Vibration-Sensor)[Delivered-Vibration-Sensor](https://github.com/ARTS-Laboratory/Drone-Delivered-Vibration-Sensor)
- **Docking System:** [https://github.com/ARTS-Laboratory/UAV-](https://github.com/ARTS-Laboratory/UAV-Package-Delivery-System)[Package-Delivery-System](https://github.com/ARTS-Laboratory/UAV-Package-Delivery-System)
- **Bridge Data:** [https://github.com/ARTS-Laboratory/USC-walking](https://github.com/ARTS-Laboratory/USC-walking-bridges)[bridges](https://github.com/ARTS-Laboratory/USC-walking-bridges)

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