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

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## **The ARTS-Lab at USC**

#### We use foundational science

Day School

to develop essential tools



to solve real-world problems



public domain

Dan Thompson

#### We are Engineers (mostly)



-4 1.0

## **Data Assimilation**









# **Embedded Systems**





#### **Flexible Electronics**





In Situ Monitoring of AM







#### Nuclear Magnetic Resonance









Water Quality Sensors







#### **Vibration Sensors**



# **Data Assimilation**







**Civil Structures** 





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



**High-Rate Systems** 







**Battery Systems** 



# **Embedded Systems**





Microcontroller/7 microprocessor







**Real-Time OS** 







### **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.





propeller under ceiling effect





# 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)

$$f_t = \sigma(W_f x_t + U_f h_{t-1} + b_f)$$
  

$$i_t = \sigma(W_i x_t + U_i h_{t-1} + b_i)$$
  

$$o_t = \sigma(W_o x_t + U_o h_{t-1} + b_o)$$
  

$$\tilde{c}_t = \tanh(W_c x_t + U_c h_{t-1} + b_c)$$
  

$$c_t = f_t \circ c_{t-1} + i_t \circ \tilde{c}_t$$
  

$$h_t = o_t \circ \tanh(c_t)$$
  

$$y_t = W_d^T h_t + b_d$$



LSTM

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

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

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

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



# **LSTM Performance**

#### LSTM compensator performance

- For testing a chirp excitation in 0-5 Hz is used
- SNRdB enhancement of 9.34%
- RMSE reduction of 19.66%
- Usable bandwidth (< ±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

testing	$\mathrm{SNR}_{\mathrm{dB}}$	RMSE
sensor package	17.26  dB	$1.795 \times 10^{-3}$
LSTM compensator	18.88 dB	$1.442 \times 10^{-3}$
% improvement	9.34%	19.66%



# **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: <u>https://github.com/ARTS-Laboratory/Drone-</u> Delivered-Vibration-Sensor
- Docking System: <u>https://github.com/ARTS-Laboratory/UAV-</u> <u>Package-Delivery-System</u>
- Bridge Data: <u>https://github.com/ARTS-Laboratory/USC-walking-bridges</u>

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