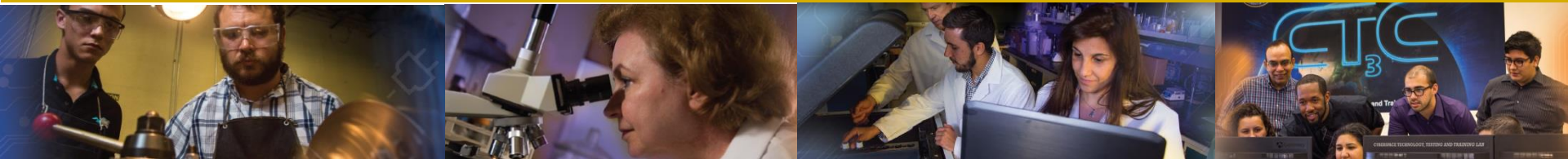




# Anomaly Detection and Task Planning via Neural Networks and Hierarchical Task Networks

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DOE-FIU Science and Technology Workforce Development Program  
Applied Research Center  
Florida International University





# Summer 2017 Internship at SNL



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Sandia National Laboratories



# Project Description/Background



Challenge: How to extract valuable insights from the vast quantities of diverse data that are generated across the DOE complex?

- Current systems record large amounts of information for environmental monitoring, tank and pipe inspections, etc:
  - Video
  - Sensor data (temperature, radiation, etc.)
  - Records/documents
- Currently rely on human operators for analysis and decision-making
  - Too much information to parse
  - Limited reaction time
  - Not easily scalable



# Scope/Objective

- Develop a system for automatically detecting and handling anomalies in multimodal, spatial and temporal data
  - Determine a sequence of actions to search for anomalies
  - Able to interpret sensor data and extract meaning



# Method / Approach



- Inspection Tool/Process Selection
  - Hierarchical Task Network (HTN) to make decisions
  - Goal-oriented
  - Adaptive (incorporates current world-state)
  - Extensible (new actions are easily incorporated)
- Detection
  - Shared representations for multimodal data
  - Long Short Term Memory (LSTM) Neural Networks to detect anomalies



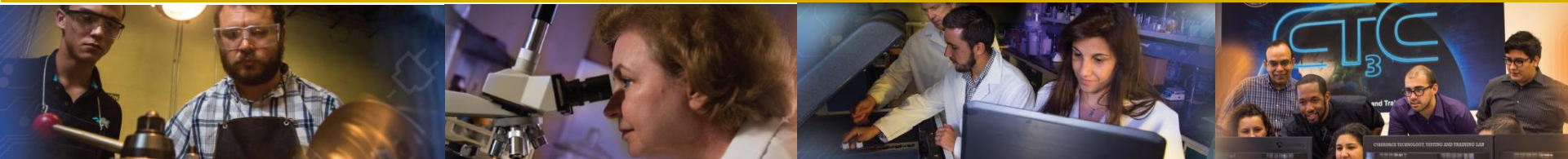
**FIU**  
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solution driven

Automated Task Planning

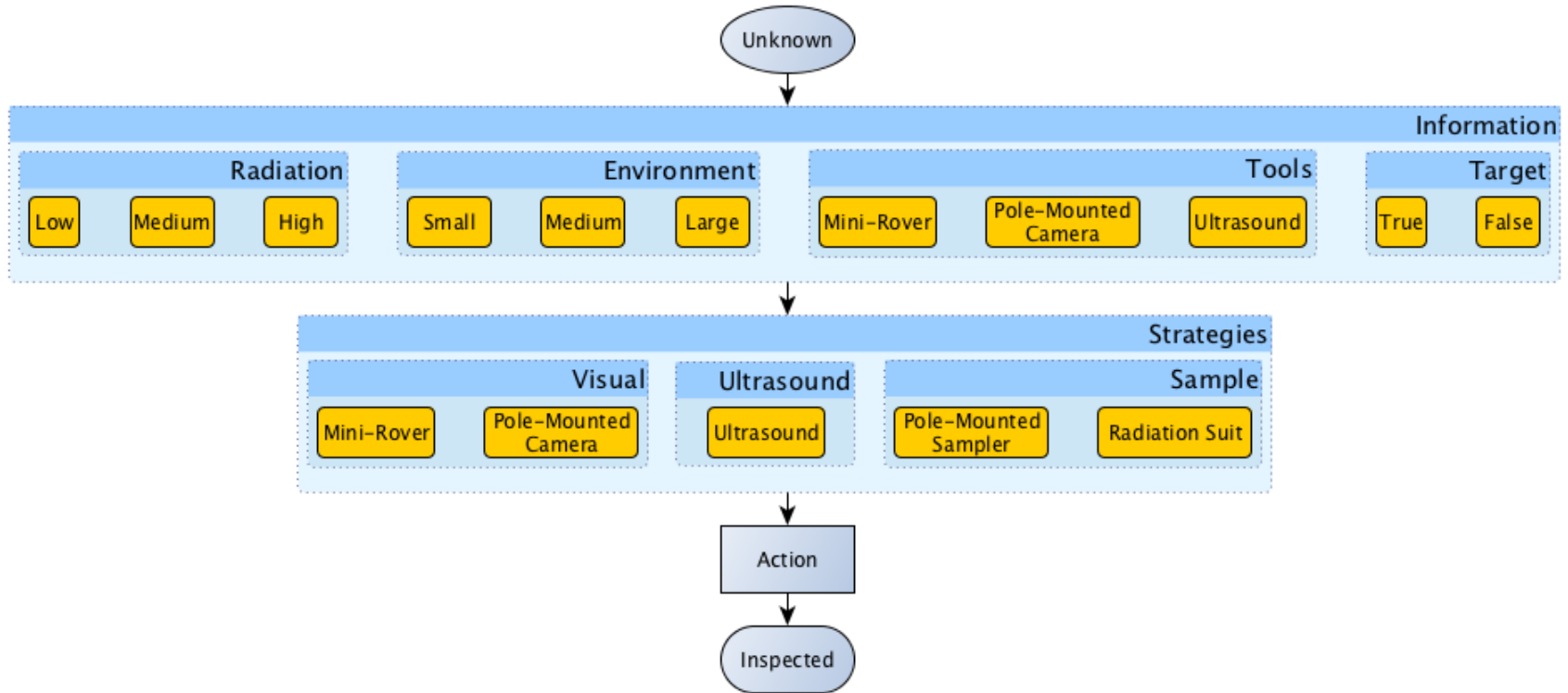
# PRELIMINARY RESULTS/DISCUSSION

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# Automated Inspection Planning





# Automated Inspection Planning



## World State

- status: unknown
- tools = mini-rover, pole-mounted camera, ultrasound
- environment: large
- radiation: low
- target: **true**

## Response

- status: inspected
- tool = **ultrasound**
- strategy = **ultrasound**





# Automated Inspection Planning



## World State

- status: unknown
- tools = mini-rover, pole-mounted camera, ultrasound
- environment: **large**
- radiation: low
- target: **false**

## Response

- status: inspected
- tool = **pole-mounted camera**
- strategy = **visual**



# Automated Inspection Planning



## World State

- status: unknown
- tools = mini-rover, pole-mounted camera, ultrasound
- environment: **small**
- radiation: low
- target: false

## Response

- status: inspected
- tool = **mini-rover**
- strategy = **visual**



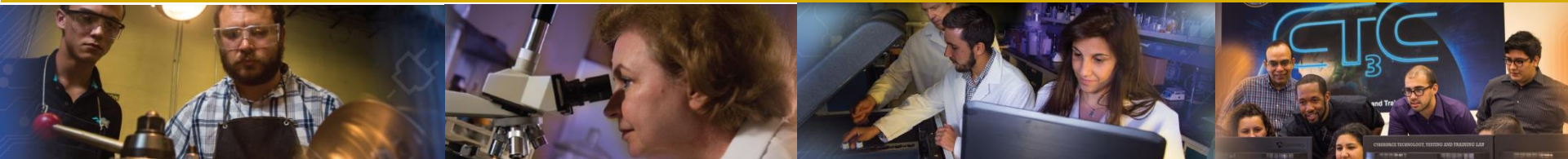
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Image Captioning

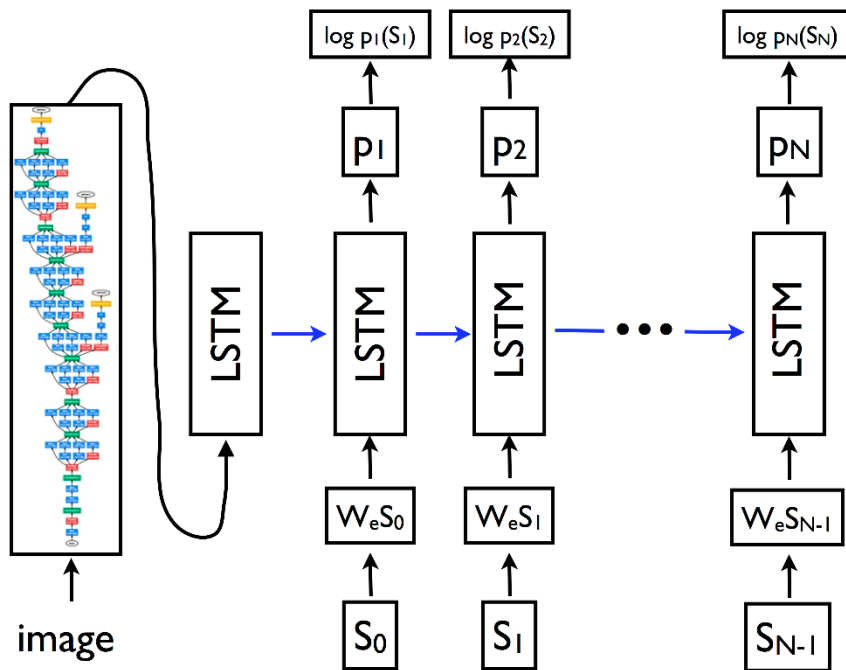
# PRELIMINARY RESULTS/DISCUSSION

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# Machine Learning using LSTM



- Learning a prediction model using LSTMs and ConvNets
- Input : time series multimodal, spatially correlated data
- Output : predicted error vectors



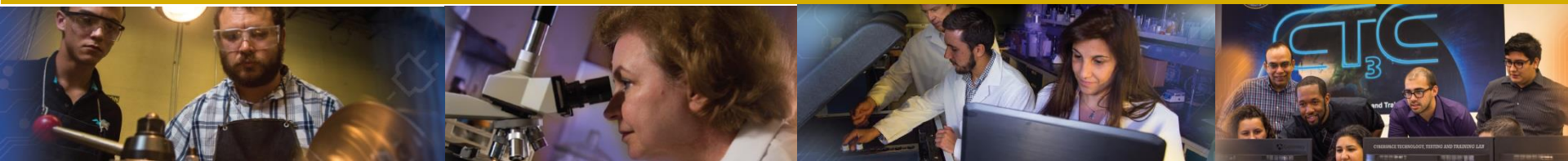
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Anomaly Detection

# PRELIMINARY RESULTS/DISCUSSION

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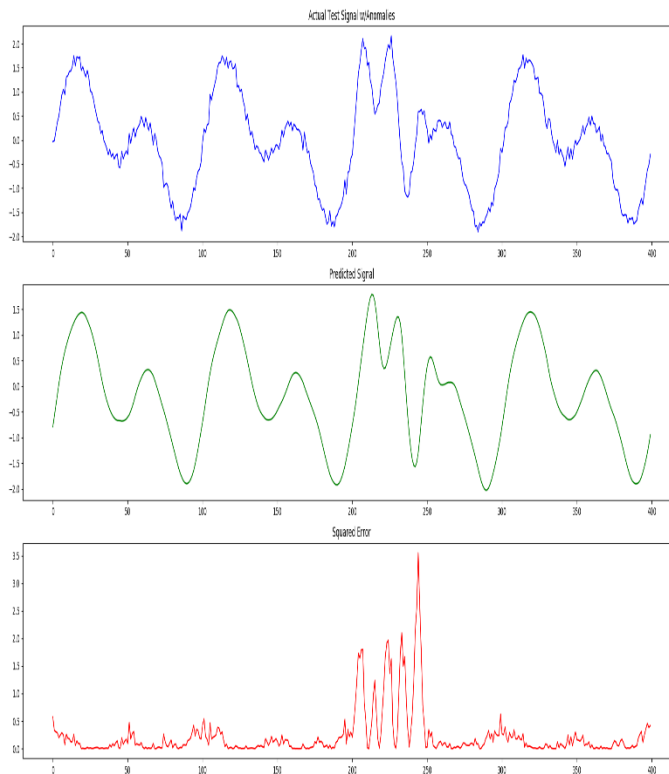


# Anomaly Detection



Example :

- Train on sine wave
- Expect a repeating sine-like wave
- Inject an anomaly
- Error when measured value does not match predicted value





# Preliminary Results/Discussion



- Elements in place for a comprehensive inspection system
  - Select inspection method and tool
  - Interpret data
  - Determine if data is anomalous
  - Respond to environment



# Conclusions

- Neural networks have been studied extensively for image captioning
- Some work has been done for anomaly detection
  - Little work on multimodal data fusion via shared representations
- Preliminary results indicate that the system presented is capable of addressing pressing inspection and monitoring concerns
  - Easily extensible towards other areas





# Future Work

- Collect multimodal, spatio-temporal correlated training data
- Fuse sensor data into a shared representation
- Expand LSTM to work with multimodal data
- Modify LSTM to work with continuous data stream vs sliding window
  - Tune and train LSTM
- LSTM research is expected to continue after returning to FIU from Sandia



# Acknowledgements



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