Scalable Modeling and Adaptive Real-time Trust-based Communication (SMARTc) System for Roadway Inundations in Flood-Prone Communities (()) UNIVERSITY VIRGINIA Khan Iftekharuddin, Old Dominion University **OLD DOMINION** Award Type: SCC-IRG Track 2, #1951745 **UNIVERSITY**

Project Challange

Recurrent flooding is expected to get worse due to sea lever rise, storm surge, and heavy rain. There is no scalable and effective system to monitor and forecast road inundations due to flooding.





Project Goals

- Develop a Scalable Modeling and Adaptive Realtime Trust-based communication (SMARTc) system for roadway inundation detection and monitoring
- Evaluate the system for a flood-prone region in the City of Norfolk, VA using data from the City's cameras, tide gauges, and existing and new overland water level sensors in the field



Community Partners

- City of Norfolk, VA
- RISE nonprofit organization focused on community resiliency

Test and **Evaluation:**

i) Online learning performance in H-DSRN for RNF classification; (ii) Traffic behavior analysis using combined road inundation classification obtained from H-DSRN and FPM models; (iii) Efficacy of RNF risk communication

Major Outcomes/Progress



Tested various Deep Learning (DL) models for segmenting images for detecting floodwater on roadways Created an ML pipeline to predict floodwater levels in relation to partially submerged vehicle images

A hydrodynamic model was developed in Delft3D to simulate storm surge and provide boundary data for a hydrological model for inland flooding.



Using data from a physics-based model, convolutional LSTM (ConvLSTM) networks are trained to create surrogate models that require low runtime in comparison to physics-based models to forecast street-scale flooding



To examine the effects of flood warning type on perceived risk, and the actions taken by drivers 72 participants were hired to drive through different flood scenarios created in a driving simulator.





Participants' responses when faced with an 8-inch flood warning were equivalent to responses when no depth was specified in the flood warning.

Intellectual Merit

- surveillance camera images.
- image data.
- and extent with driver behavior.
- uncertainty of flood information.

Broader Impacts

- real-time.
- region.
- projects.

Future Goals

- depth and extent from image data





Novel machine learning (ML) algorithms for detecting floodwater extent and depth in real-time based on

A coupled hydrologic-stormwater-coastal model to predict flood levels at city network level and real-time update of these predictions based on sensor and

Prediction of roadway capacities in real-time under partial inundations and correlation of floodwater depth

Effective communication of flood risk and road inundation to the public, leveraging granularity and

New solutions for predicting flooding on city roads in

Safer roads since drivers can use the information to avoid driving through flooded roads and emergency vehicles can reroute around inundated roads. A strong partnership with the City of Norfolk and RISE for evaluating the SMARTc system for a flood-prone

Integration of research outcomes into undergraduate and graduate classes, hands on activities for visiting high school students, and interdisciplinary capstone

Further enhance the ML-pipelines for predicting flood

Couple the flood inundation forecasting models with the ML-based storm surge model to improve the accuracy of forecasting compound floods.

Calibrate car following models to simulate how drivers navigate through partially flooded road segments