

National Aeronautics and Space Administration



Ellicott City Disasters II

Enhancing a Statistical Flood Risk Model to Continue Improving Early Warning Systems and Public Safety in Ellicott City, Maryland

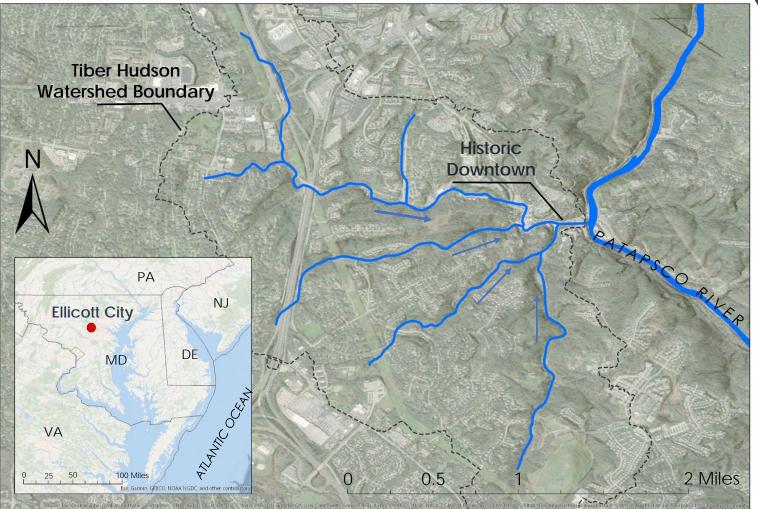
Alina Schulz, Scott Cunningham, Jonathan Donesky, & Matthew Pruett

Maryland - Goddard | Spring 2020



ELLICOTT CITY, MARYLAND

- Located on the Patapsco River in Howard County, MD
- City population ~65,000 in 2010
- 1,000 residents in Historic Downtown



Source: DEVELOP Team

COMMUNITY CONCERNS





Severe flooding over the past decade Effective use of data for **proactive** emergency management



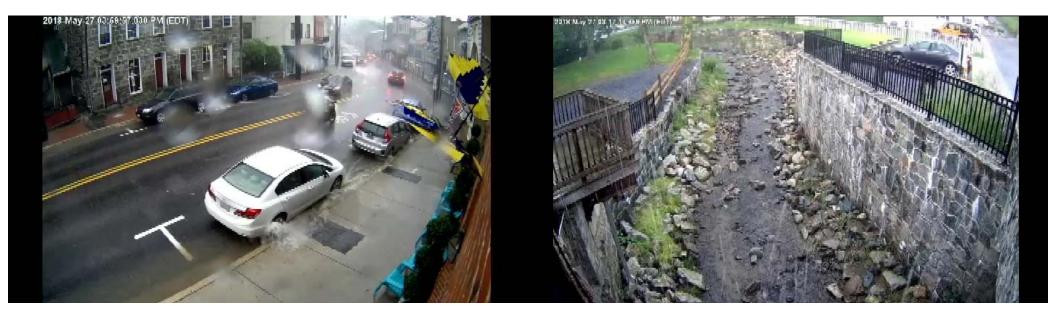


Increasing warning time before a flood can save lives and money

Source: Howard County Gov't , DEVELOP Team

DOWNTOWN IN DANGER





May 2018 Flood 2 hour time lapse

Source: Ron Peters

PARTNERS







Howard County Government Office of Emergency Management Stormwater Management Division

NOAA – NWS Baltimore-Washington Weather Forecast Service

Source: DEVELOP Team, NOAA, NWS

A THREE-TERM PROJECT

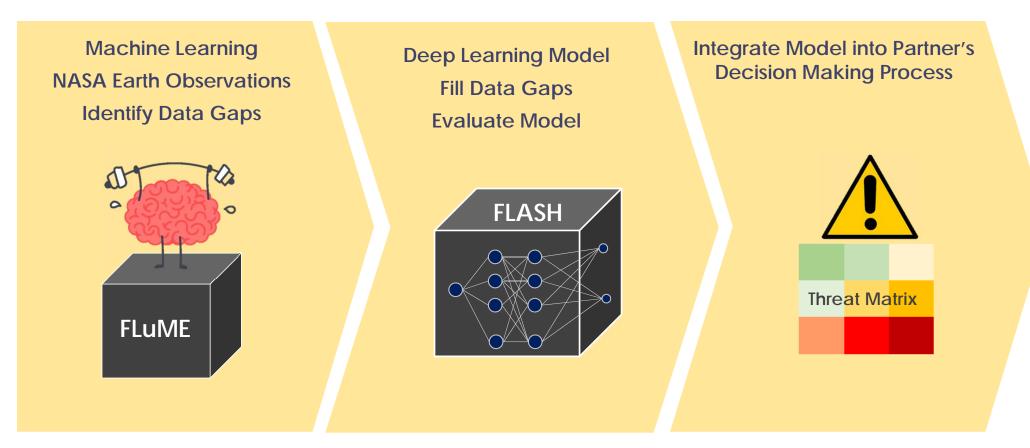


Image credit: DEVELOP Team, NextAvenue, Max Pixel

OBJECTIVES

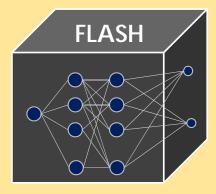


Improve FLASH with Long Short-Term Memory (LSTM) framework

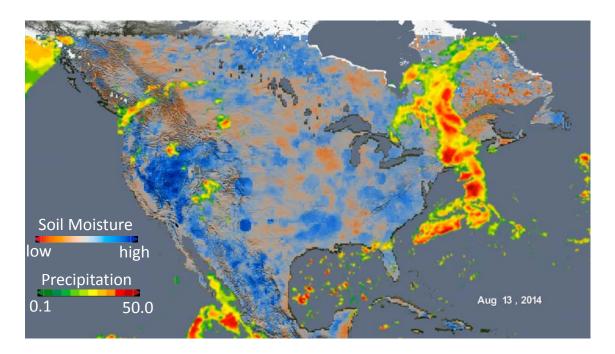
Utilize new datasets to improve FLASH accuracy

Determine the importance of each input variable for best performance

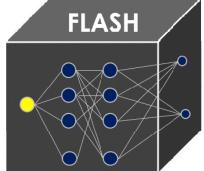
Improved Model Performance



THE DEEP LEARNING APROACH



A Landscape with Memory: Soil Moisture and Precipitation



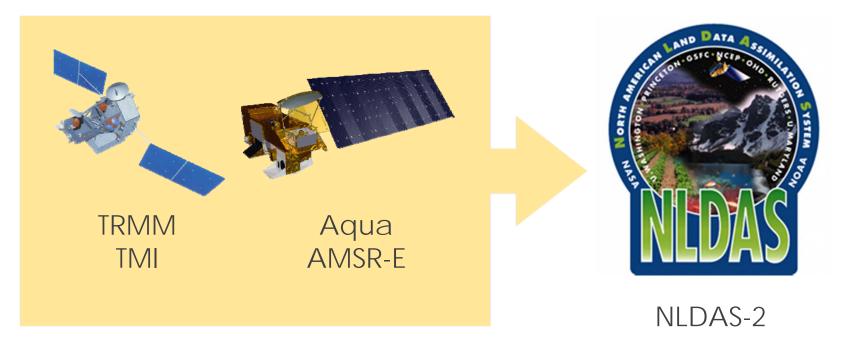
Long Short-Term Memory Deep Learning Architecture

Video Credit: NASA's Science Visualization Studio

RESOURCES USED NASA Earth Observations

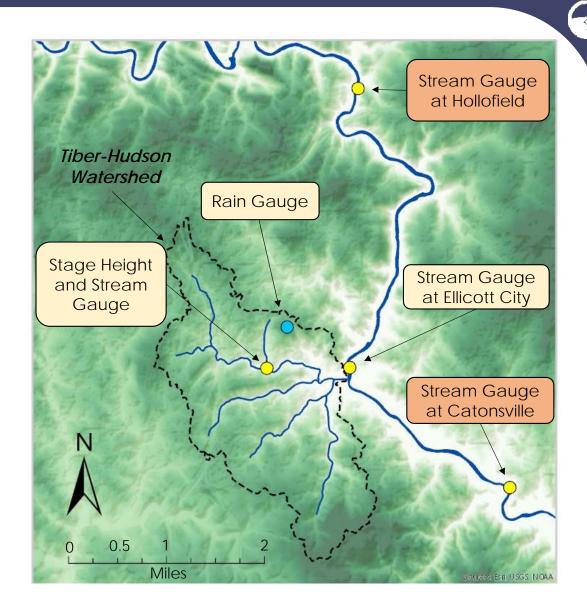
Satellites





RESOURCES USED Gauge Data

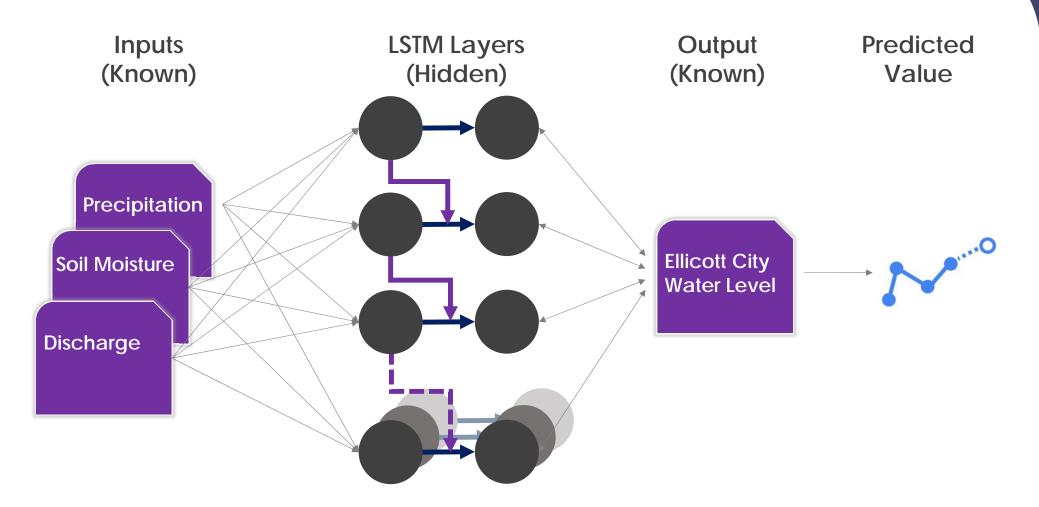
- ► Howard County Gauges (3)
 - Precipitation
 - Stream height in City
 - Discharge
- USGS Gauges (2)
 - Discharge



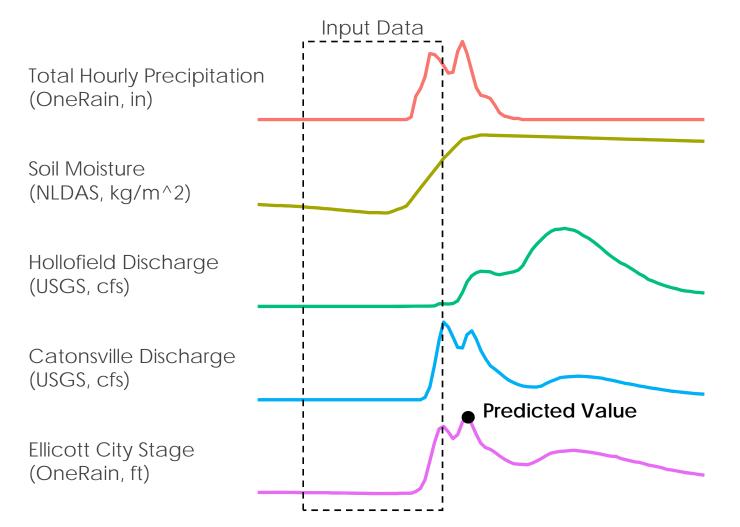
METHODOLOGY – Framework

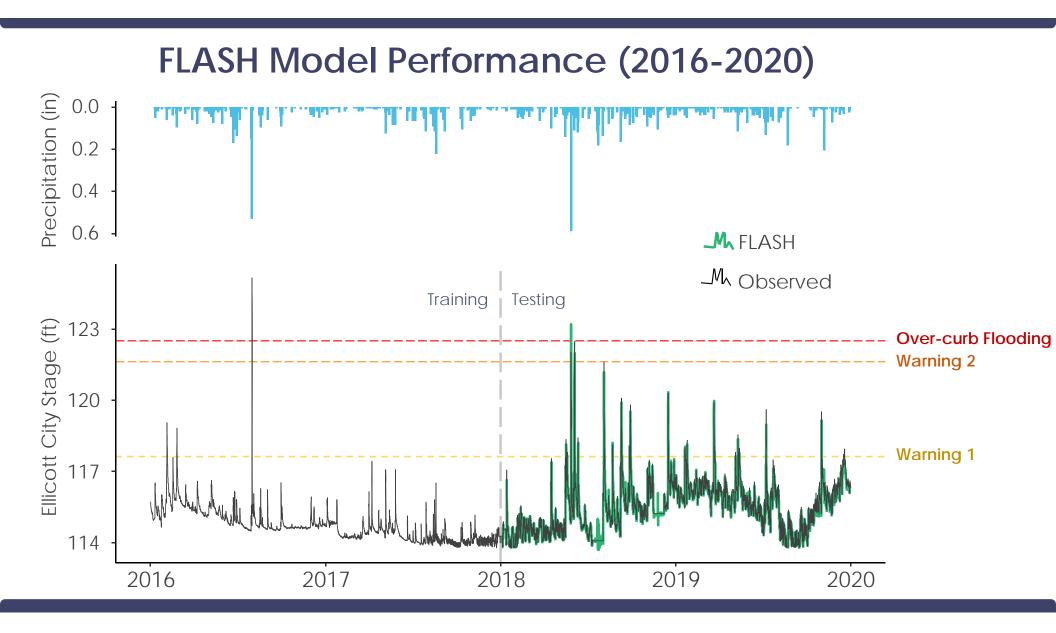
Source: DEVELOP Team, Behance

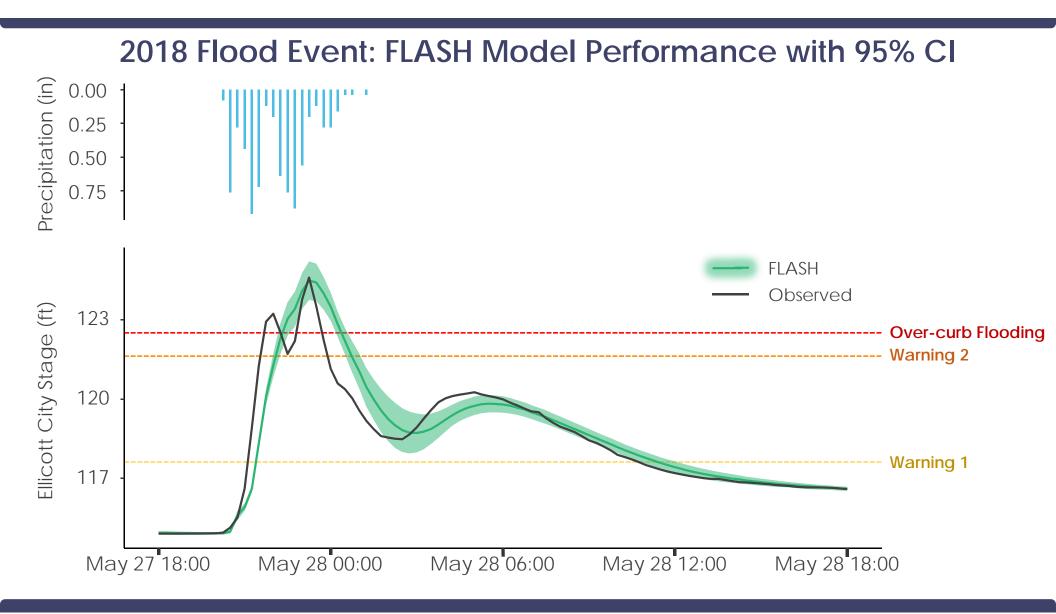
METHODOLOGY – Deep Learning



MAKING PREDICTIONS







CONSIDERATIONS

- > The model's predictions are only as good as the data it is provided.
- Accuracy evaluations are only relevant at stream gauges used for calibration.
- Currently, weights are randomly initialized; therefore, model output quality can vary.

CONCLUSIONS

The **LSTM** deep learning framework improved the model's performance.

The model's predictive capability was developed. NASA Earth observations coupled with well documented ground data bolstered model performance.

NEXT STEPS

- Enhance model's prediction accuracy using data from other flood-prone watersheds
- Integrate socio-economic data to represent the impacts of predicted flood levels on different parts of the community
- Create an simple, intuitive graphical user interface for end users



Source: DEVELOP Team

ACKNOWLEDGEMENTS

- Our Advisors:
 - Dr. John Bolten, NASA GSFC Lead Science Advisor
 - Dr. Sujay Kumar, NASA GSFC
 - Perry Oddo, NASA GSFC
- Darcy Gray, GSFC Center Lead
- Ellicott City Disasters I Team, NASA GSFC
 - Terra Edenhart-Pepe, Julio Peredo, Caroline Resor, and Callum Wayman

- Howard County Government
 - Brian Cleary, Mike Hinson, Chris Meyer, Christopher Strong, Shaina Hernandez, and Calvin Ball
- Christopher Strong, Baltimore-Washington Weather Forecast Office of NOAA-NWS Warning Coordination Meteorologist

This material is based upon work supported by NASA through contract NNL16AA05C. Any mention of a commercial product, service, or activity in this material does not constitute NASA endorsement. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Aeronautics and Space Administration and partner organizations.