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# An AI Classifier for Post-Disaster Aerial Imagery

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*ITU/WMO/UNEP Workshop on "Resilience to Natural Hazards through AI Solutions"*

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# Outline

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  - **Results**
  - **Proposed transition architecture**
- **Demo**



# Motivation

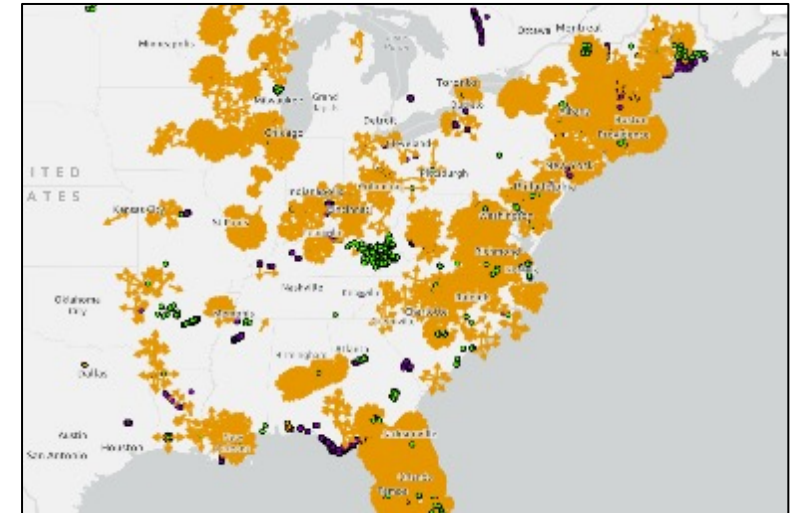
**Civil Air Patrol (CAP) collects thousands of aerial images after disasters!**



**Images are used to assess damage to homes and infrastructure**



**But it can be hard to find the most useful ones.**



**We need a way to filter the images to find the most useful ones**



# MIT LL previously trained an artificial intelligence model to identify the contents of images



**MIT LL developed a labeled CAP dataset, the Low Altitude Disaster Imagery (LADI) dataset, as part of a NIST-funded effort and released it open source in 2020.**

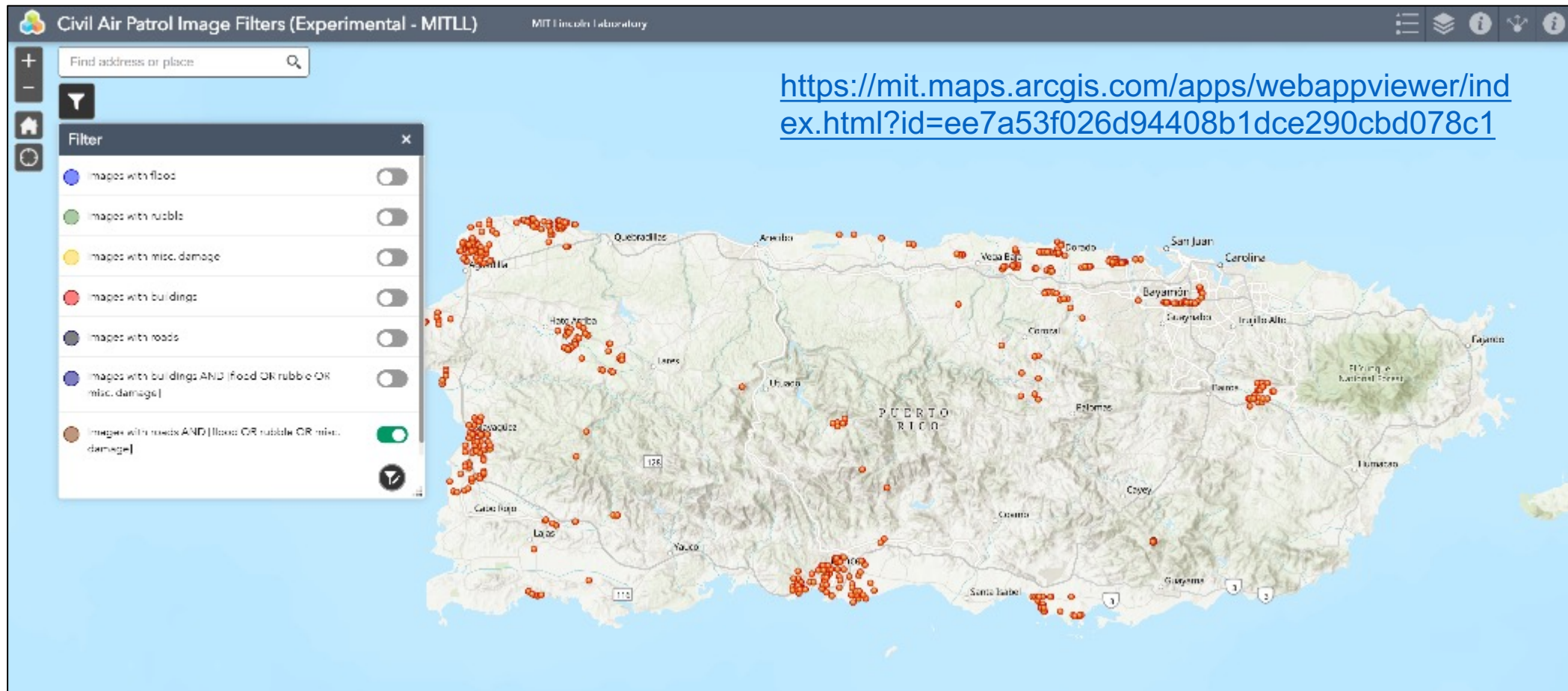
## Demonstrations

- 2021: Hurricane Ida, Kentucky Tornadoes
- 2022: Hurricanes Fiona, Ian





# New capability to filter points to find ones that show damage or other features of interest



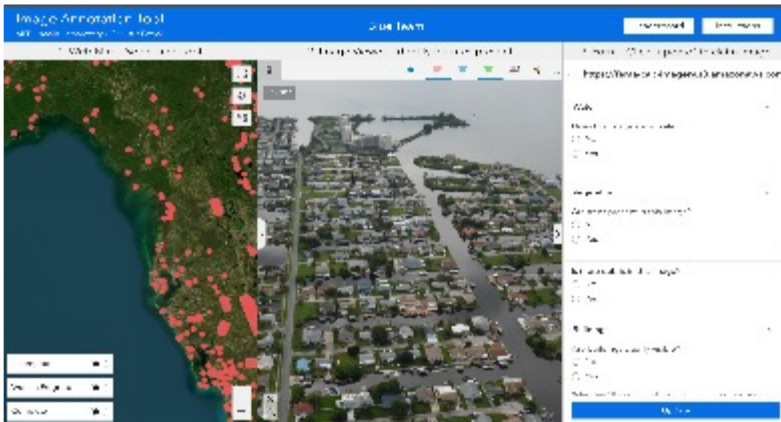
**The classifier had not been validated and had middling performance on new data.**



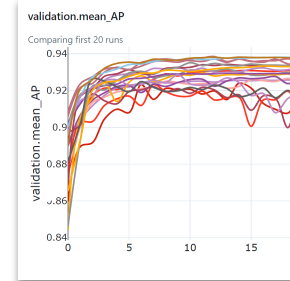
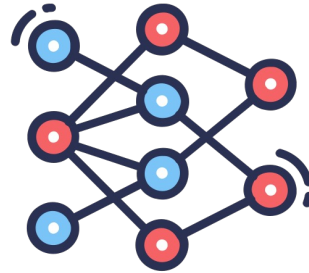
# Project overview

**Goal: All new CAP images will receive labels when they are uploaded through the FEMA CAP image uploader, so that we can filter the images when doing damage assessments.**

## 1. Create new training data



## 2. Retrain the AI model



## 3. Release LADI v2 dataset and model open source



open data



Hugging Face



GitHub



# Crowdsourcing campaign overview

- **Photos from all federal disasters in 2023 and numerous historic incidents**
  - 10,086 images in two rounds (4035, 6051)
  - Hurricanes, floods, tornadoes, earthquakes, fires
  - Contiguous US, Alaska, Hawaii, and some territories
- **Triplicate reviews**
  - 46 volunteers from CAP FEMA damage assessment team participated
  - Three teams: Yellow, Green, Blue
  - Over 250 estimated man-hours
- **16 questions per image**
  - Water, trees, debris, buildings, roads, bridges
  - Presence of features and damage to features
  - Open text field
- **Campaign dates: December 19-30, 2023**





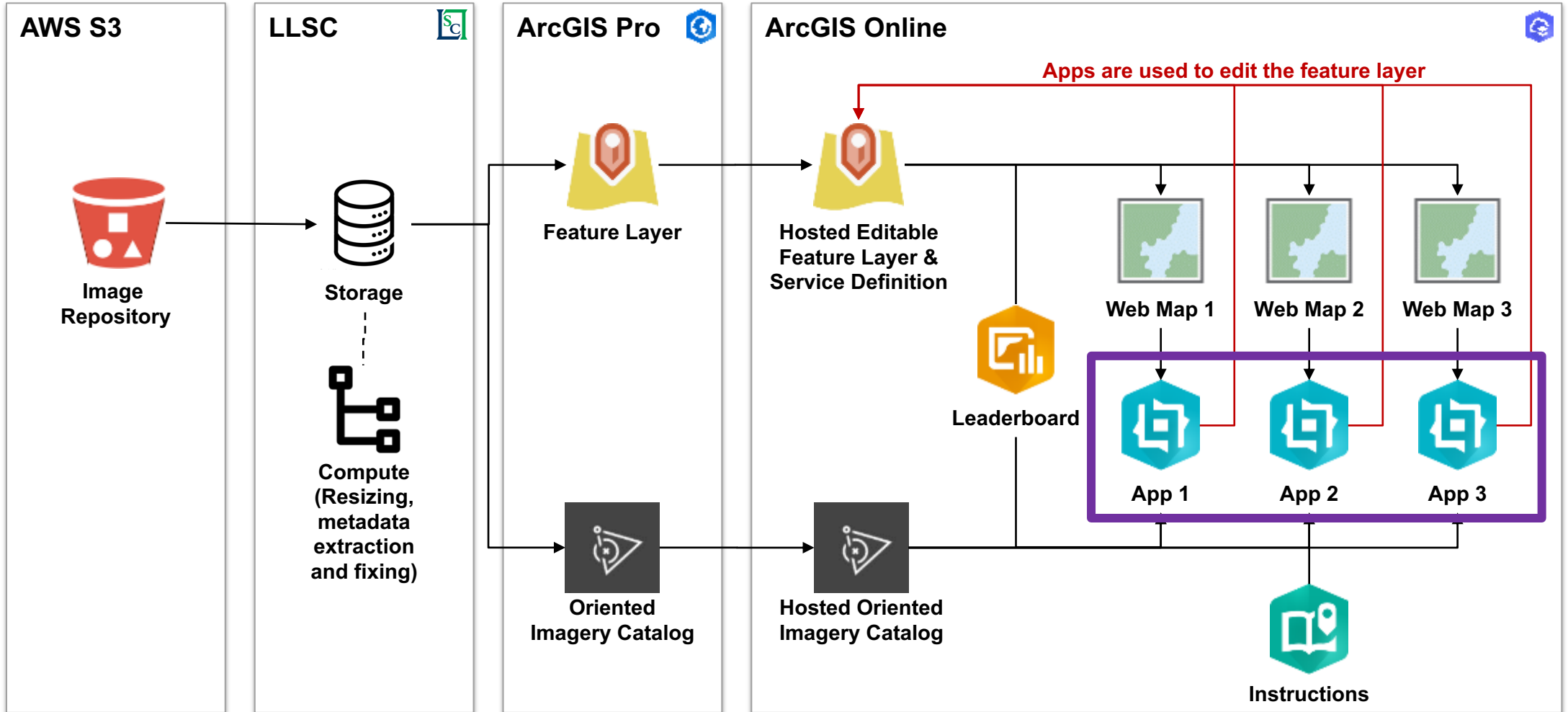
# Questions

- Does this image show **water**?
  - Is there clear evidence of flooding (any water that is not normally there?)
    - Is there clear evidence of flooding around structures?
- Are **trees** present in this image?
  - Are any trees damaged? (uprooted, downed, flooded, burned)
- Is there **debris** in this image?
- Are **buildings** clearly visible?
  - Does this image clearly show any affected buildings?
  - Does this image clearly show any buildings with minor damage?
  - Does this image clearly show any buildings with major damage?
- Does this image clearly show any destroyed buildings? (wind, fire)
- Are any **roads** clearly visible?
  - Does this image clearly show damaged roads? (flood, pavement damage/washout, debris)
- Are any **bridges** clearly visible?
  - Does this image clearly show damaged bridges? (flood, pavement damage/washout, debris)
- Please describe any other useful information this image provides.





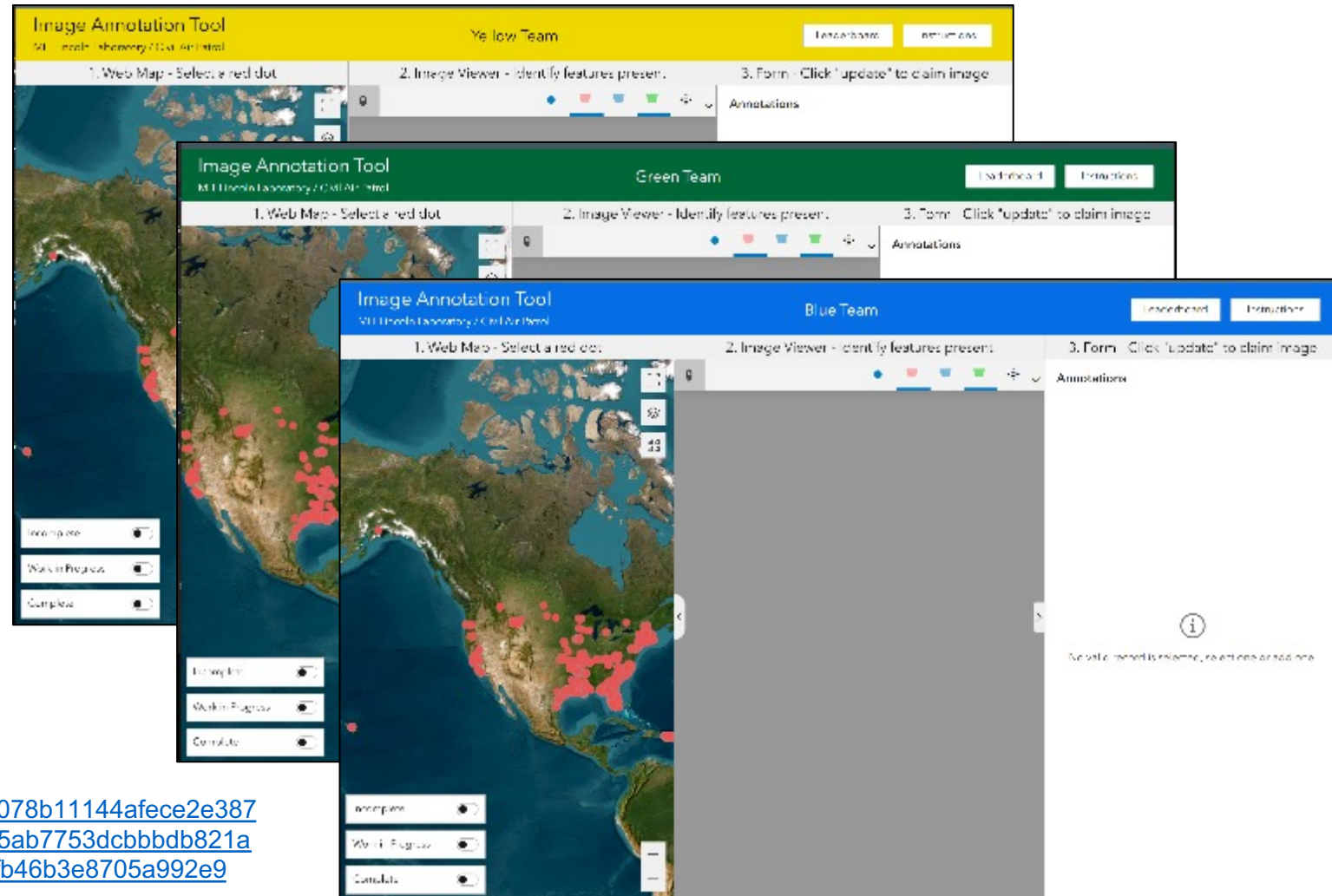
# Crowdsourcing app architecture





# Three apps, one for each team

- The apps are the same except for the color-coded header.
- Access controlled through ArcGIS Online in collaboration with FEMA.



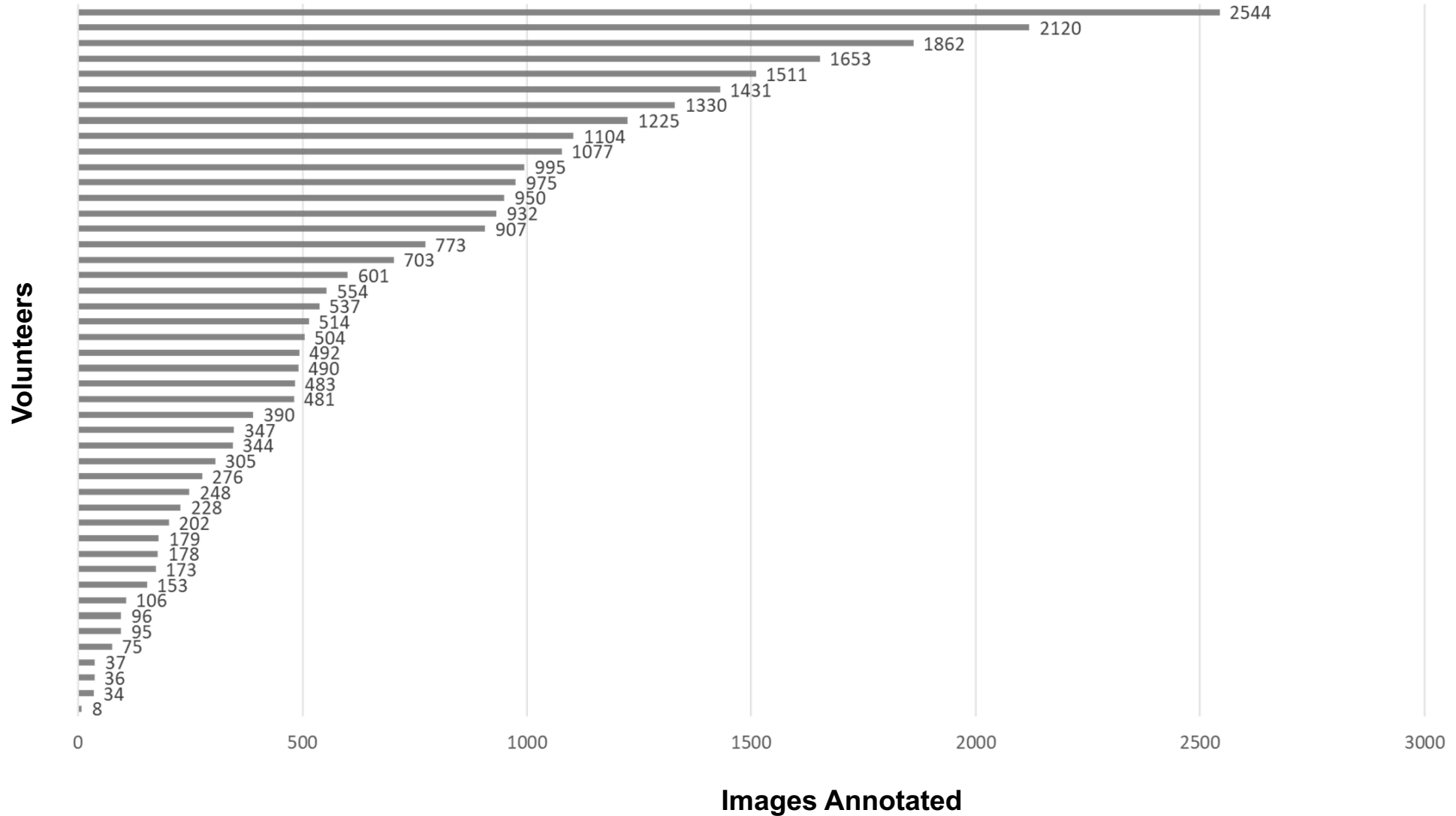
Yellow: <https://experience.arcgis.com/experience/3f627ea99ab44078b11144afece2e387>

Green: <https://experience.arcgis.com/experience/f1575e74fcd5425ab7753dcbdbdb821a>

Blue: <https://experience.arcgis.com/experience/12fa91ed79234f0fb46b3e8705a992e9>



# Leaderboard





# Results







# Training approach

## Data Preparation

- **Majority vote for annotation**
- **Image Augmentations**
  - **Resize, Crop, Color Jitter, Normalize**
- **Train/Validation/Test splits**
  - **Train: 8k (2015-2022)**
  - **Validation: 1k (2015-2022)**
  - **Test: 1k (2023)**

## Training

- **Evaluate various open-source architectures/base models**
- **Select top 5 base models, run hyperparameter tuning**
  - **Learning rate**
  - **Optimizer algorithm**
- **Select top candidates**
  - **Considering performance, model size**



# AI base models evaluated

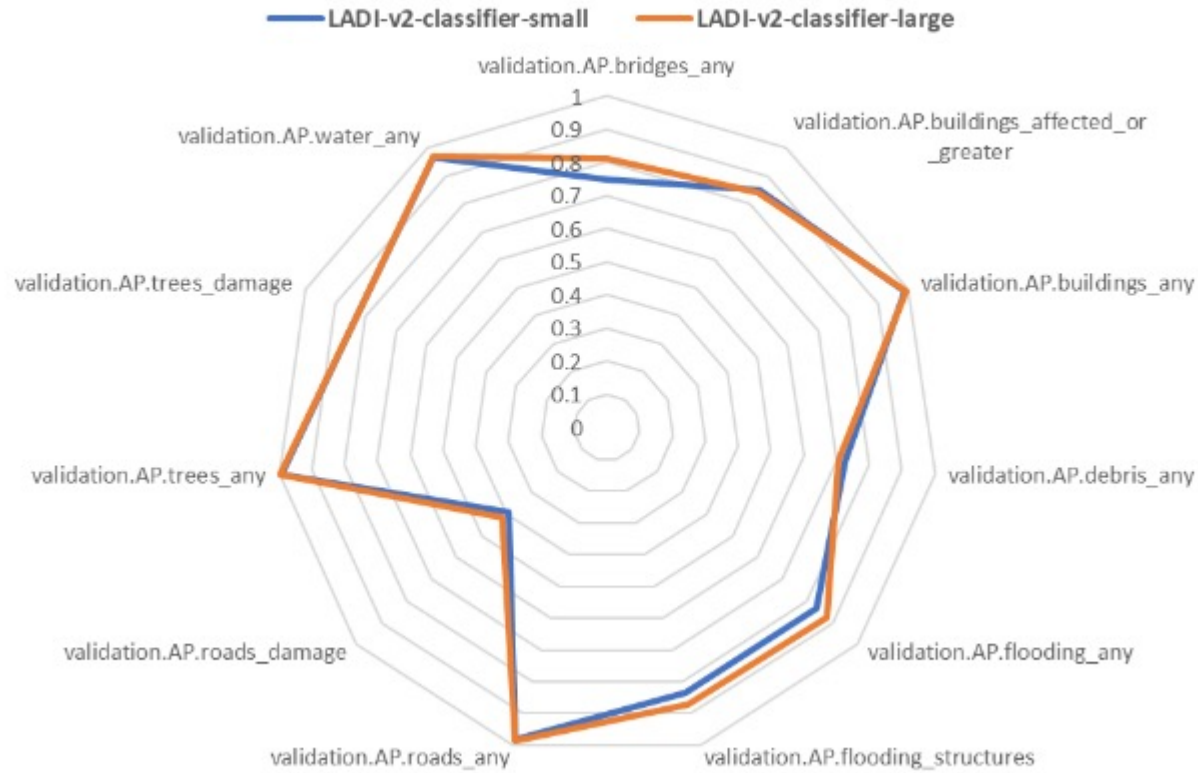
- Sourced 25 pre-trained image classification models from 🤗 Hugging Face repositories
  - Fine-tuned on crowdsourced data using the MIT LL TX-Green super computer
  - Selected top-5 candidates using mean Average Precision metric on Validation set
  - Selected two final candidates based on different architectures
    - LADI-v2-classifier-small (based on google/bit-50)
    - LADI-v2-classifier-large (based on microsoft/swinv2-large-patch4-window12to16-192to256-22kto1k-ft)
  - Will train final versions with validation and test set included for deployment
- apple/mobilevitv2-1.0-imagenet1k-256
  - facebook/convnextv2-huge-22k-384
  - facebook/convnextv2-large-22k-224
  - facebook/deit-base-distilled-patch16-224
  - facebook/deit-base-patch16-224
  - **google/bit-50**
  - google/efficientnet-b0
  - google/efficientnet-b7
  - google/mobilenet\_v1\_1.0\_224
  - google/mobilenet\_v2\_1.0\_224
  - google/vit-base-patch16-224-in21k
  - google/vit-base-patch32-224-in21k
  - google/vit-base-patch32-384
  - google/vit-huge-patch14-224-in21k
  - google/vit-large-patch16-224-in21k
  - google/vit-large-patch16-384
  - microsoft/beit-base-patch16-224-pt22k
  - microsoft/focalnet-base
  - microsoft/resnet-152
  - microsoft/resnet-50
  - microsoft/swin-large-patch4-window7-224-in22k
  - microsoft/swin-tiny-patch4-window7-224
  - microsoft/swinv2-large-patch4-window12-192-22k
  - **microsoft/swinv2-large-patch4-window12to16-192to256-22kto1k-ft**



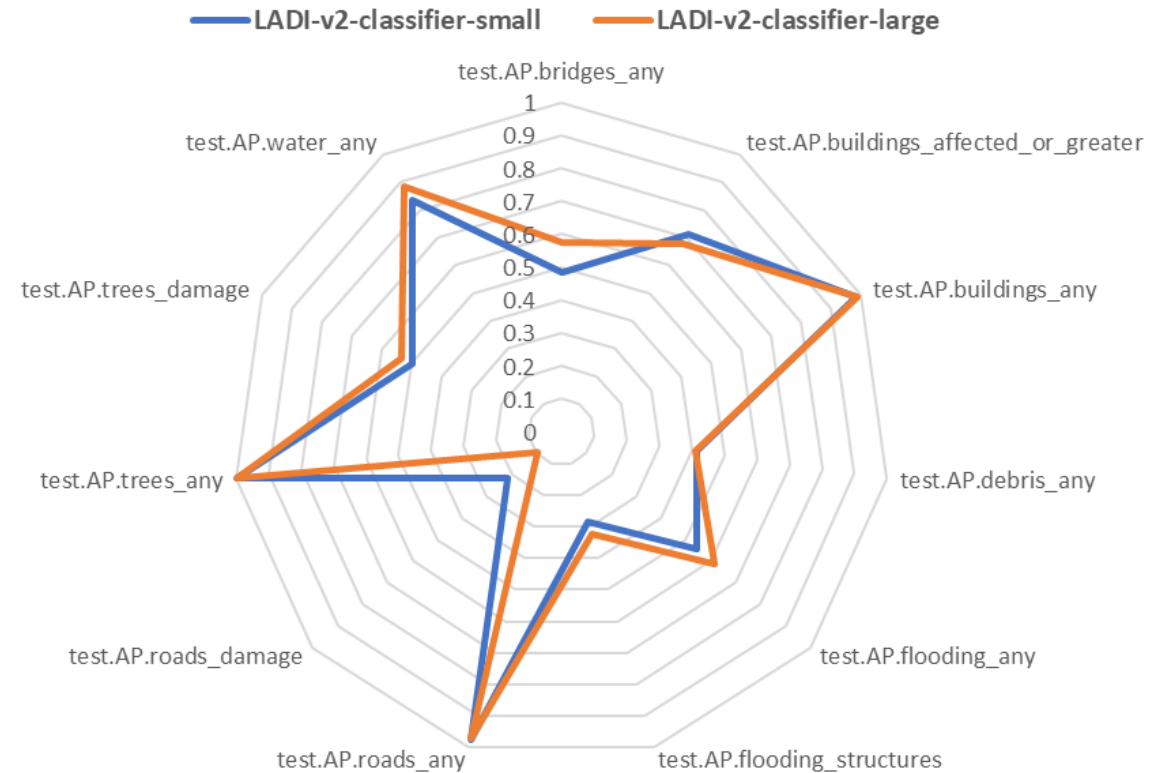
# Best candidate performance

Average precision (AP) of LADI-v2-classifiers (small, large) on validation and test sets. Higher is better.

### Release Candidate Performance (Validation Set)



### Release Candidate Performance (Test Set)

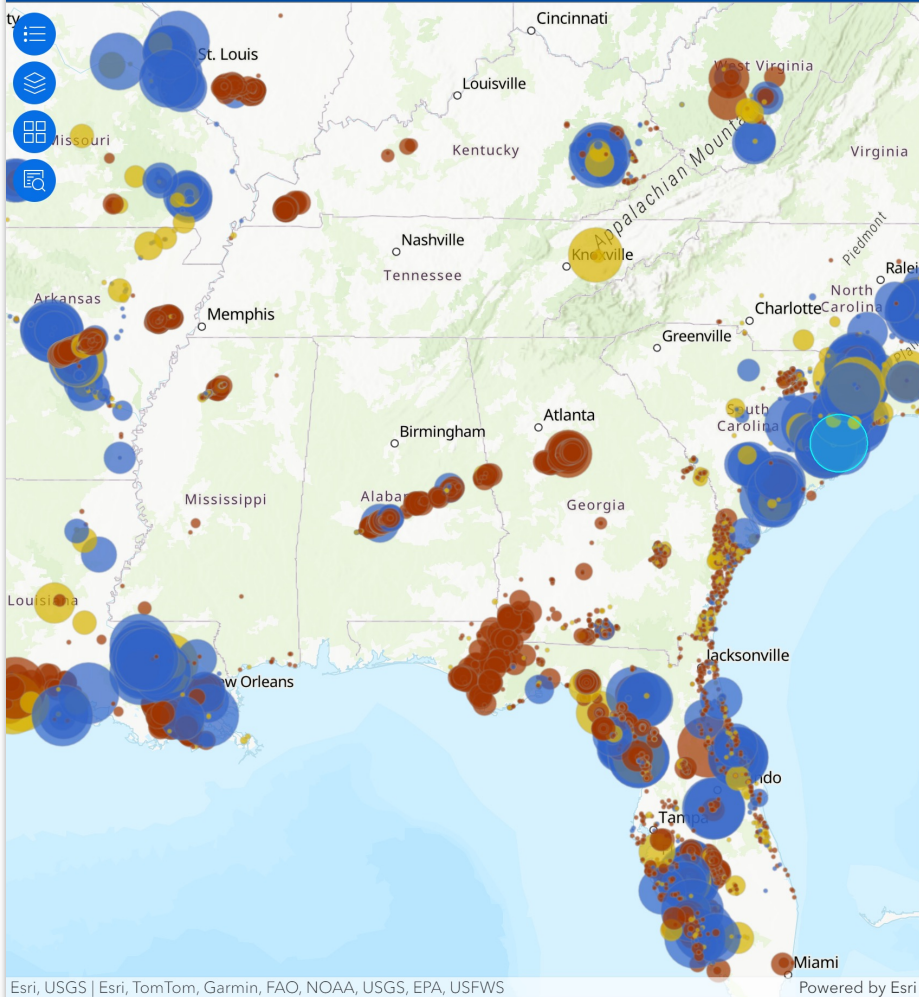


**Performance degradation observed in out-of-sample prediction (2023 test set vs. 2015-2022 validation set), especially in damage-related categories. Performance of presence of water, trees, roads, and buildings is robust.**



# Demo

## Demo CAP Image Filter App



### Filters

Toggle filters "on" to view images likely to contain each feature. Click the "down arrow" next to each feature to change the certainty threshold.

- Bridges
- Buildings
- Building damage
- Debris
- Flooding
- Flooded structures
- Road damage
- Tree damage
- Trees
- 2023 Incident(s)

0 Selected

< 6236 of 16219 >



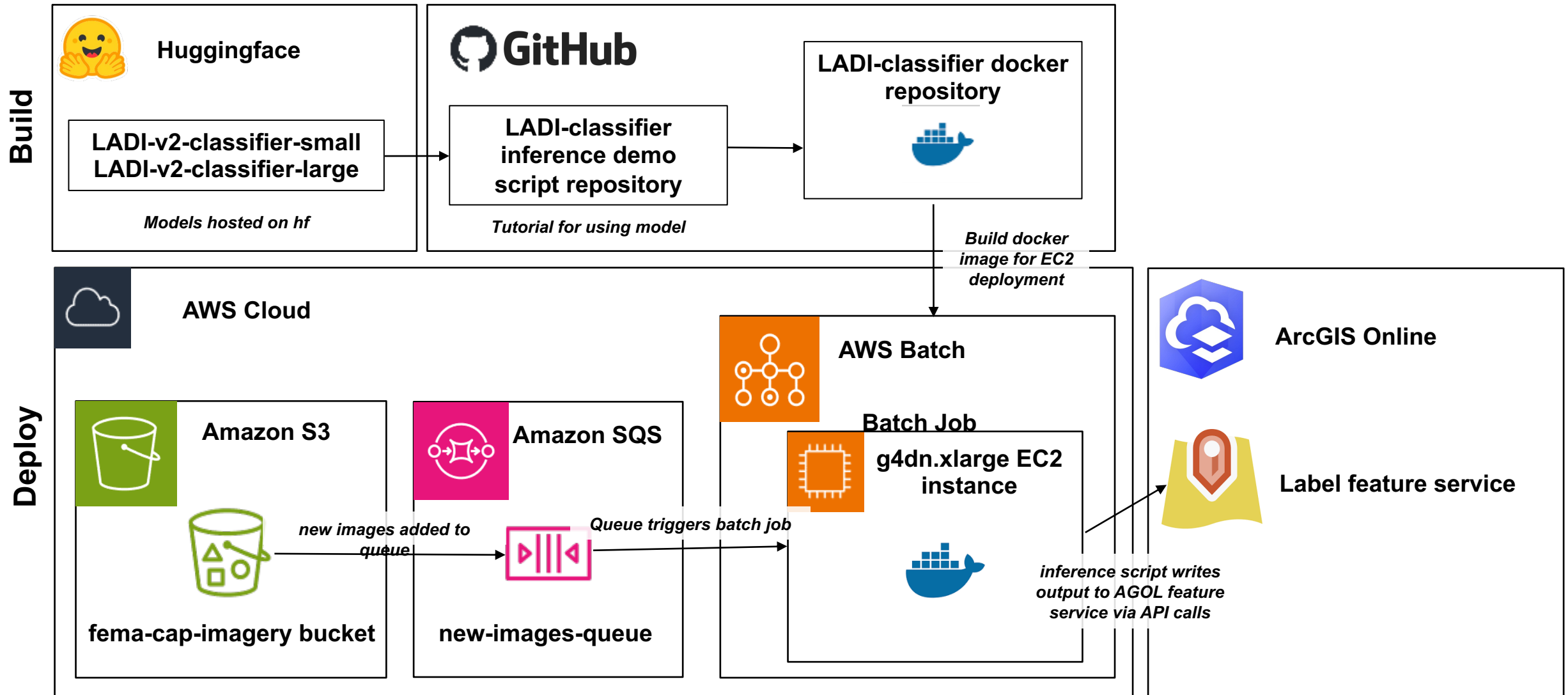
Link to source image	View
Buildings	1.00
Building damage	0.74
Water	1.00
Flooding	1.00
Flooded structures	0.94
Trees	1.00

<https://experience.arcgis.com/experience/136c316e81db4a0e9af924c7d20ec8cb>





# Proposed AWS implementation architecture

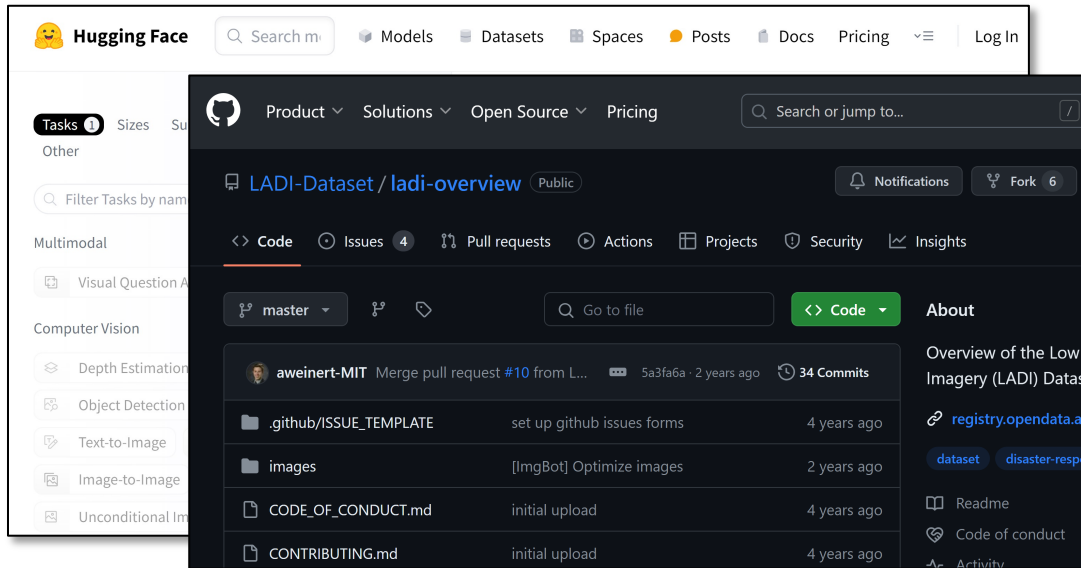




# Next steps

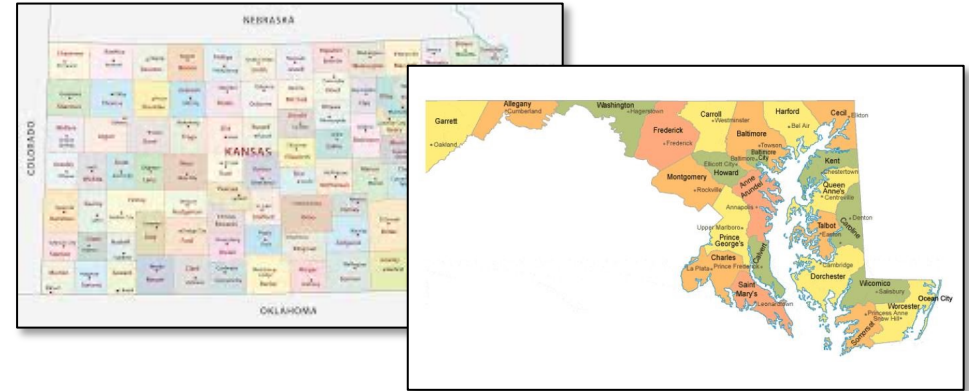
## Project Completion

1. Publish LADI v2 data and model open source on HuggingFace and GitHub by end of April, 2024
2. Support partners to deploy on AWS



## Future Development

1. Fine tune and deploy for state partners



2. Implementations with Vision-Language Models

### Examples of “notes” input by volunteers

Agricultural rows
railroad bridges, tree debris against bridge pylon.
Power line/power station infrastructure; area around this infrastructure is flooded.
The house without roof shingles in the upper right corner has a dumpster present - possibly regular work or remodel. Not clear enough to declare it major damage.
A couple foundations with debris and a bulldozer on site. Other imagery layers show empty clean lots. Not obvious that this is a storm-related destroyed building.



# Contact

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**For updates on the LADI v2 dataset and models, follow us on GitHub:**

**<https://github.com/ladi-dataset>**

