



Radiant Earth
Foundation

EARTH IMAGERY FOR IMPACT

Radiant MLHub:

Advancing Utilization of AI Applications on Earth Observations with Benchmark Training Datasets

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2nd NOAA Workshop on “Leveraging AI in Environmental Sciences”

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Mission

Empowering organizations and individuals globally with open Earth observation training data, standards and tools to cultivate a global community focused on machine learning and Earth observations to meet the world's most critical challenges.



Vision

Leveraging machine learning and Earth observation for positive global impact

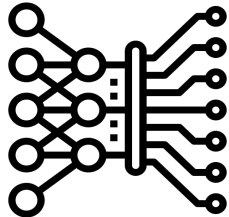


Motivation



- **Increasing volume of data**

- New satellites, in situ sensors, and models
- Governments and commercial entities



- **Advancements in Machine Learning techniques**

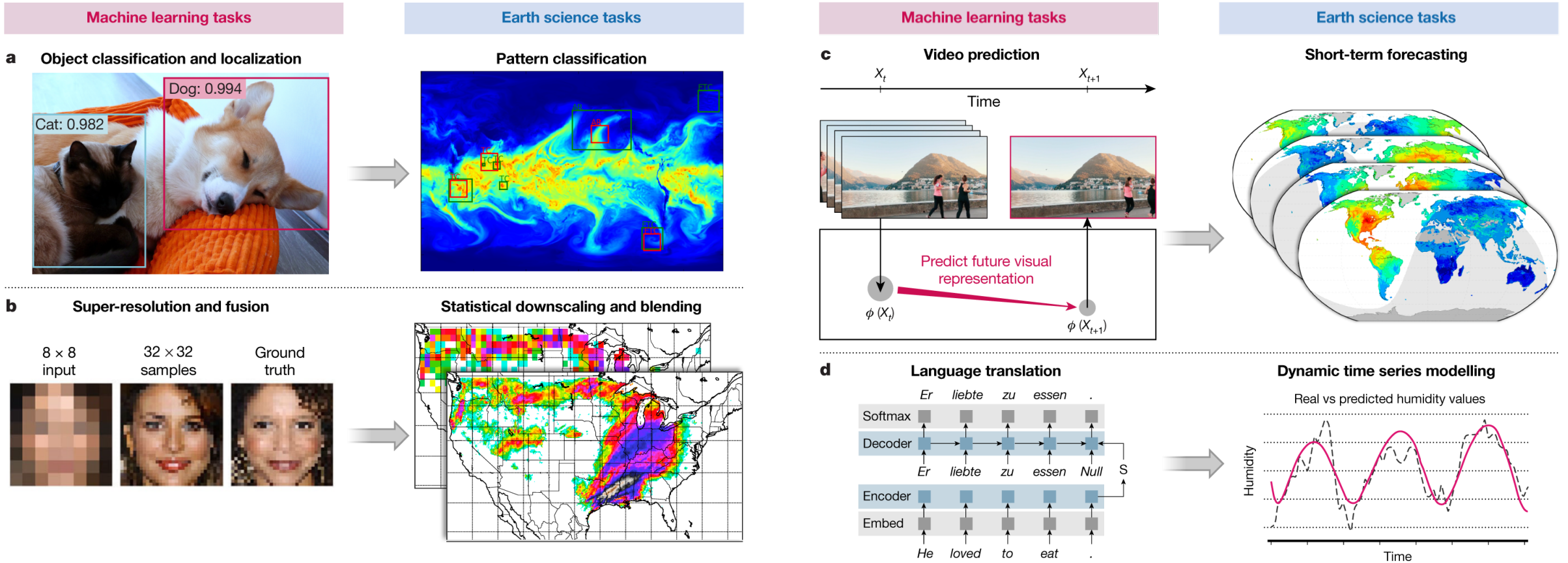
- Data-driven and fast iterations
- Capable of detecting complex and non-linear relationships

- **Availability of cloud services**



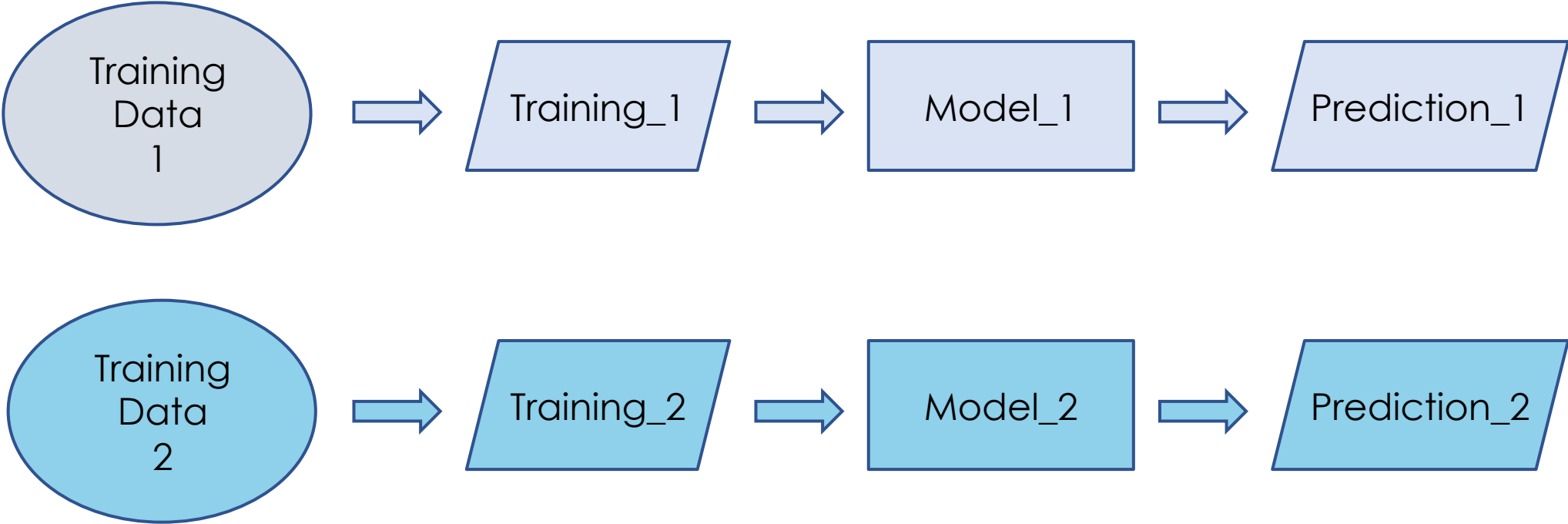
- Brining computation to data
- Scaling resources on-demand
- Serverless designs

ML Tasks in Earth Science

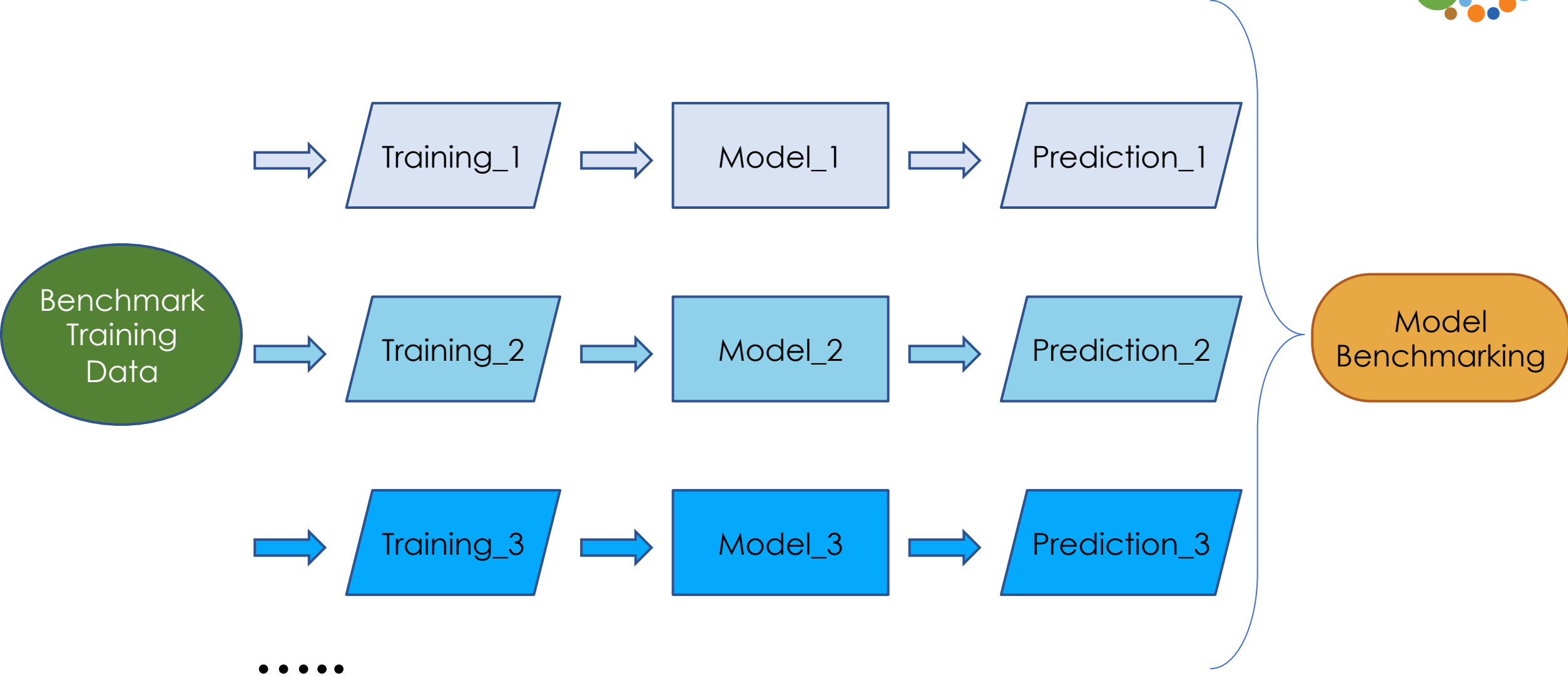


Source: Reichstein. et al., 2019

Existing Workflows

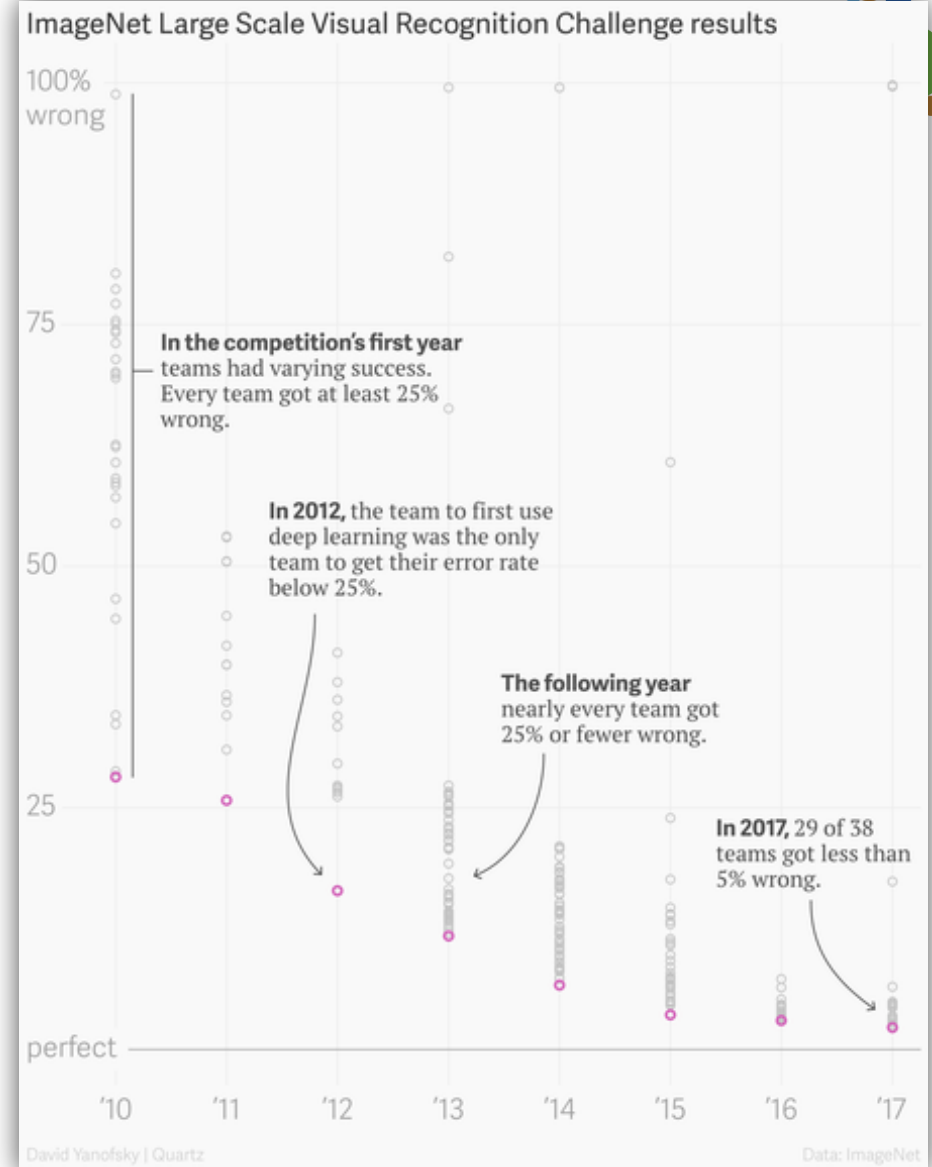


Ideal Workflow



ImageNet

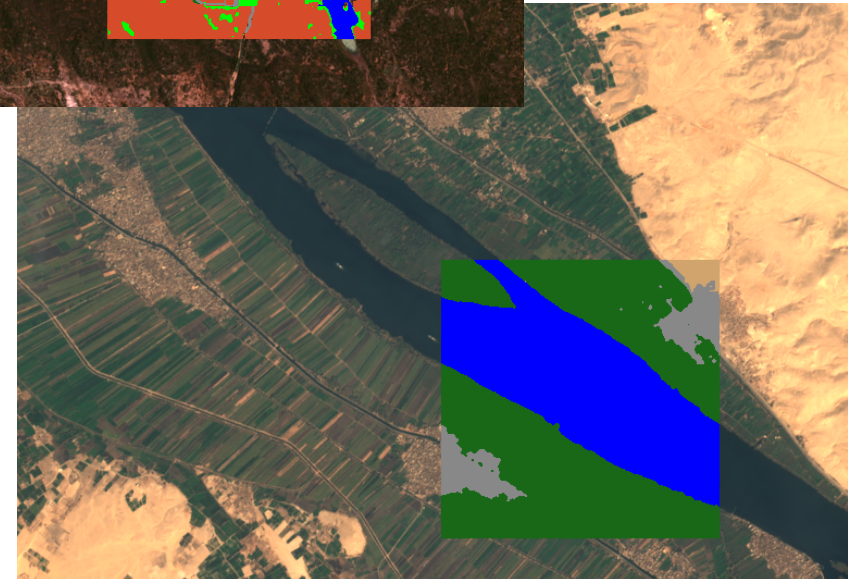
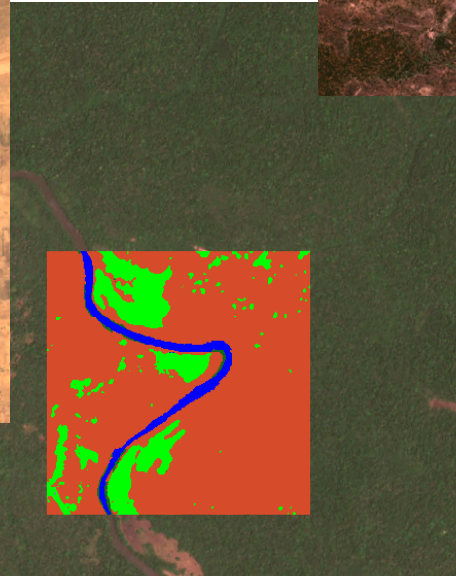
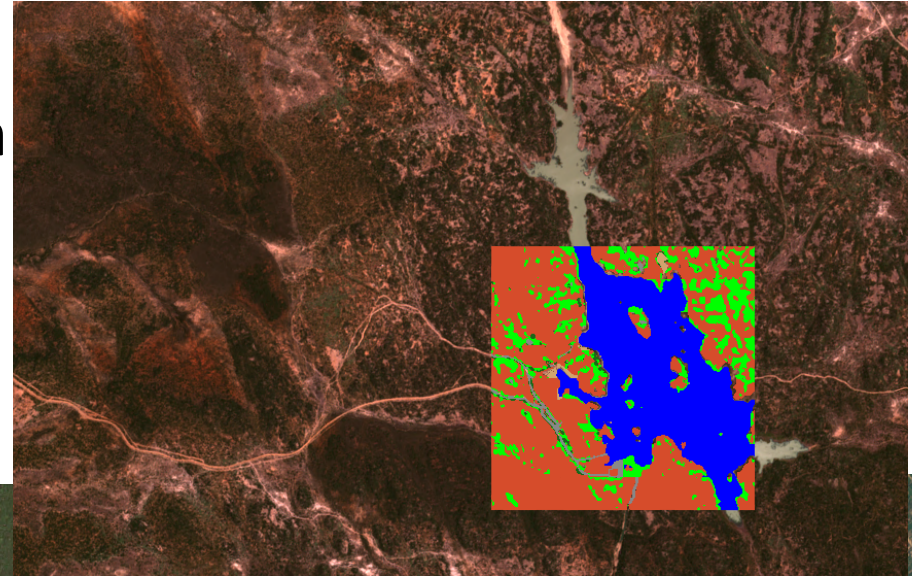
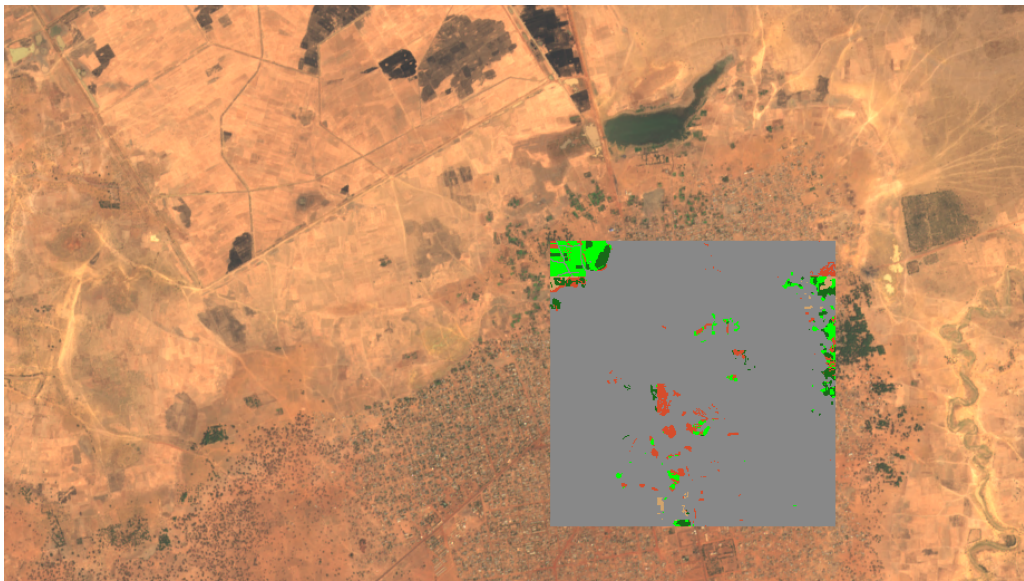
- 14 M annotated images including 1 M with object bounding boxes.
- 20 K categories of objects
- Open access
- Annual competition 2010-2017



LandCoverNet



- First geodiverse LC training dataset
- 130M labeled pixels at 10 m resolution
- 7 classes based on annual time series

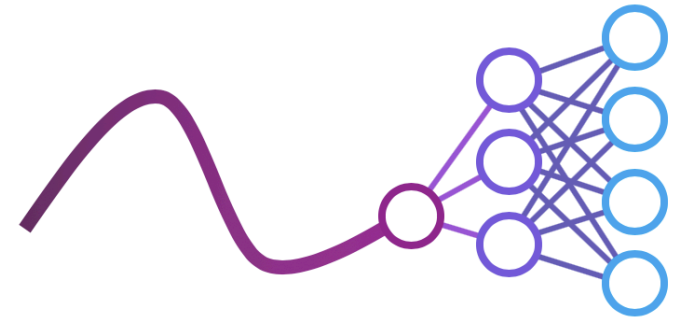


WeatherBench



- A benchmark dataset for data-driven weather forecasting (based on ERA-5)
- 14 variables including
 - Temperature
 - Relative Humidity
 - Vorticity
 - Precipitation
 - Cloud Cover
 -
- Evaluation metrics and baseline models

WeatherBench



[GitHub repo](#)

Training Data Challenges in Earth Science



Geospatial Training Data Catalogs:

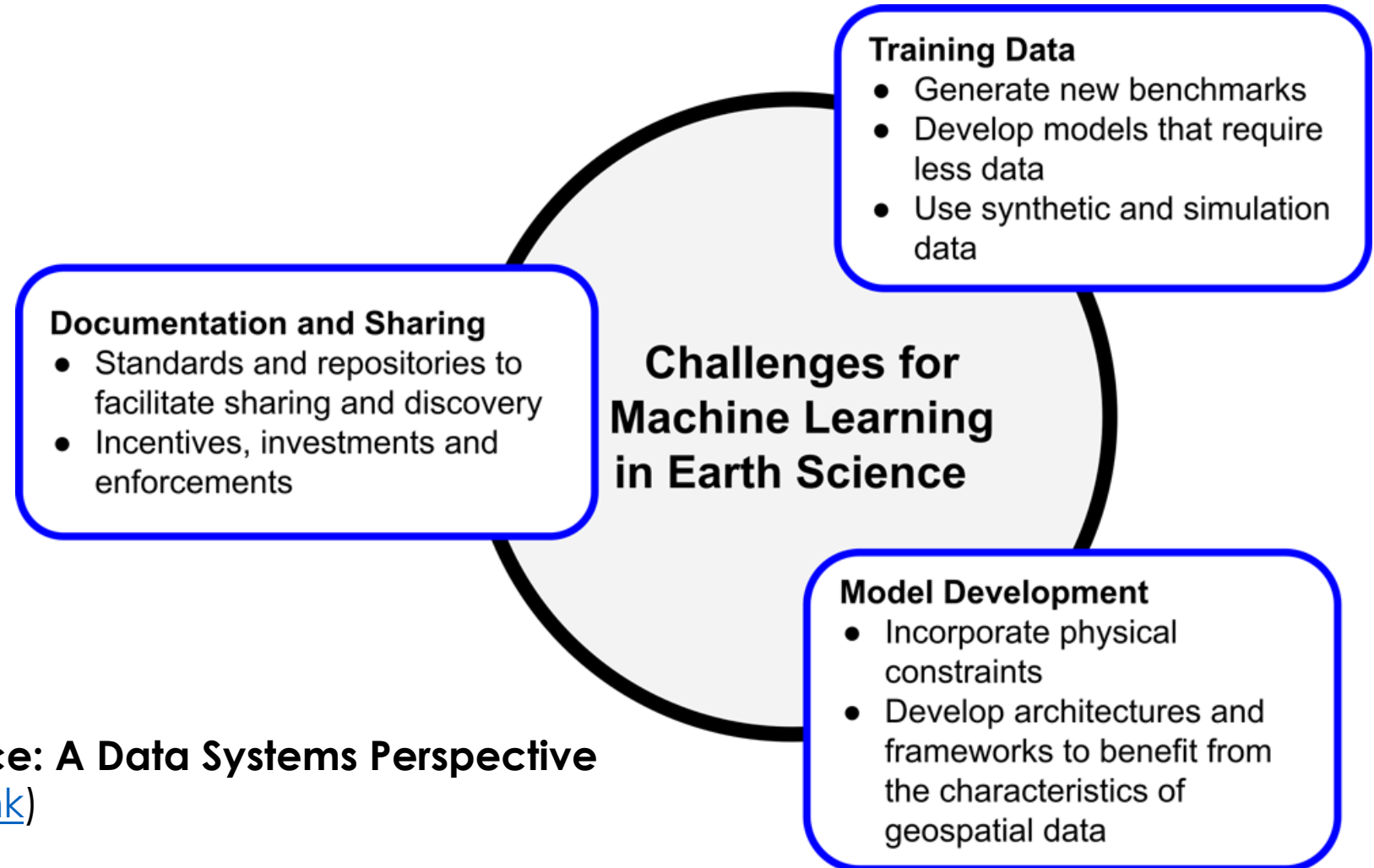
- Lack of Geo-Diversity
- Scarce data sources
- Data Accessibility
- Inter-Operability
- Machine learning-readiness



Result of Gaps in Training Data Catalogs:

- Biased or incorrect results
- Inability to capture wide range of possible outcomes in space and time

Challenges for ML in Earth Science



ML Commons for Earth Observation



Hub

- EO Training Datasets
- ML Models
- Competitions
- Image annotation + ground-referencing

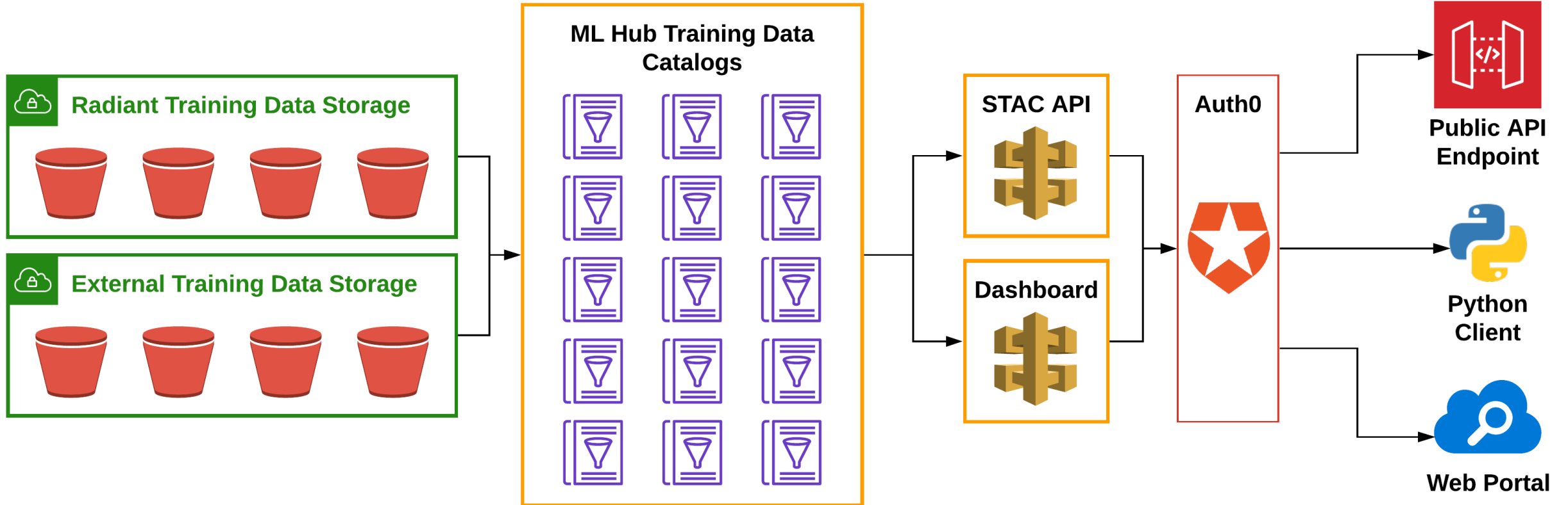
Community

- Convenings to develop standards for ML on EO
- Interoperability of datasets
- Technical Working Groups
- White Papers

Education

- EO market information
- Best practices on use of ML and EO
- Speaking engagements
- Media outreach

Radiant ML Hub



Radiant MLHub Repository



- Each dataset has a DOI with version and citation

- FAIR data principles

- Findable
- Accessible
- Interoperable
- Reusable

Radiant MLHub Training Data Registry

 Radiant MLHub
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Tropical Cyclone Wind Estimation Competition

<https://doi.org/10.34911/rdnt.xs53up>

goes regression tropical storm

Description

A collection of tropical storms in the Atlantic and East Pacific Oceans from 2000 to 2019 with corresponding maximum sustained surface wind speed. This dataset is split into training and test categories for the purpose of a competition.

Documentation

<http://doi.org/10.1109/JSTARS.2020.3011907>

Citation

M. Maskey, R. Ramachandran, I. Gurung, B. Freitag, M. Ramasubramanian, J. Miller "Tropical Cyclone Wind Estimation Competition Dataset", Version 1.0, Radiant MLHub. [Date Accessed]
<https://doi.org/10.34911/rdnt.xs53up>

STAC Collections

Description
GOES Training Source Imagery

Resource type
Source Imagery

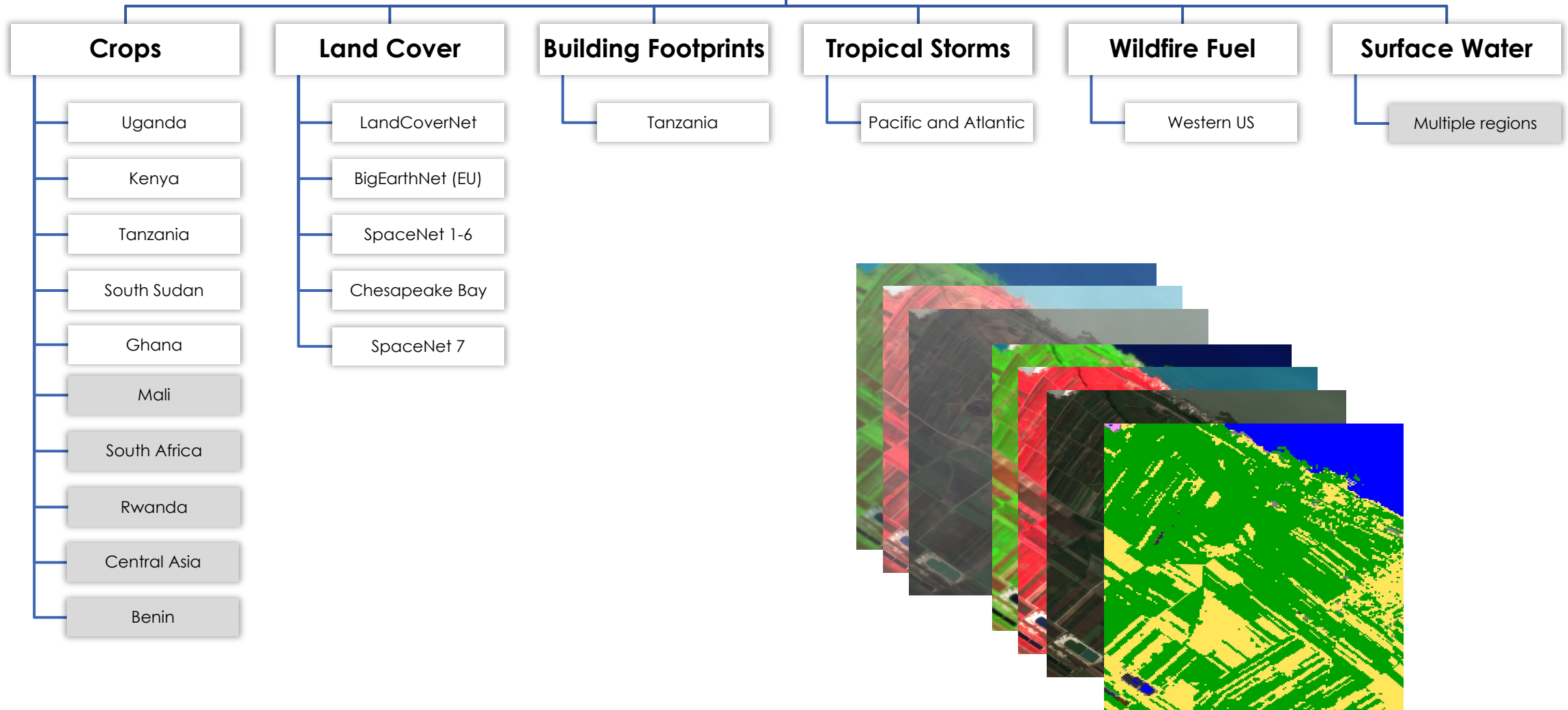
Collection ID
[nasa_tropical_storm_competition_train_source](#)

License
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Description



Radiant MLHub Data Catalog



Competitions



Past:

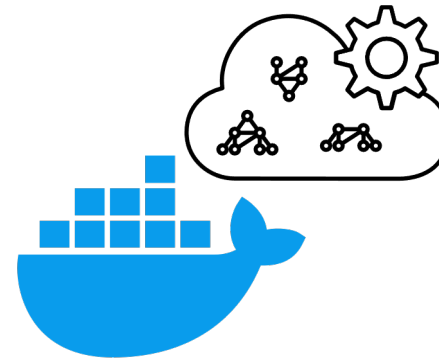
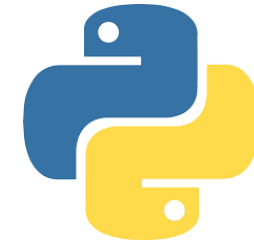
- Crop type classification in Kenya
Multiband and temporal Sentinel-2
- Cloud labeling
Optical data from Sentinel-2
- Tropical Cyclone Wind Speed Estimation
Temporal observations from GOES

Future:

- Crop Monitoring in South Africa (Summer)
Multiband and temporal Sentinel-2 and Sentinel-1

What's coming up?

- Enhance training data repository
 - Develop a Python Client (alpha version released)
 - Build integration with NASA EOSDIS CMR catalog
- Develop an ML model repository
 - Commit your model in GitHub
 - Register on Radiant MLHub
 - Share with the world!
- Generate new training datasets
 - Fusion of SAR and multi-spectral data



Thanks!

www.radiant.earth



www.mlhub.earth



MLHub Slack Channel:
bit.ly/MLHubSlack

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