



Center for Artificial Intelligence

Using AI/ML to Advance NOAA Missions *AI in Government*

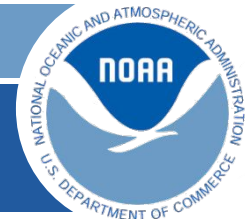
Eric A. Kihn, Ph.D.

NESDIS AI Lead, Chief of NCEI Oceans Geophysics Science and Services

Contributions from: NAIEC, NCAI Team

1/20/2022





NOAA People Supporting AI

AI Executive Committee (NAIEC) (as of January 2022*)

NMFS

Ben Richards (co-chair)
Brett Alger

OAR

Jebb Stewart (co-chair)
Venkatachala 'Ram' Ramaswamy

NESDIS

Eric Kihn
Vacant

NWS

Jamese Sims
Hendrik Tolman

NOS

Neil Weston
Tyler Christensen

OMAO

John Katebini
Vacant

OCIO

Frank Indiviglio

NAIEC Executive Secretary
Vacant

NCAI

Rob Redmon (Lead)
Jordan Watson (Deputy - LANTERN) until Feb 2022
Heather McCullough (Deputy - LANTERN) as of Feb 2022

AI Working Group (NAIWG)

Christin Khan (NMFS)
Erin Moreland (NMFS)
Christian Jones (NMFS)
Bryan Costa (NOS)
Mamoudou "Ama" Ba (NWS)
Vacant (NESDIS)
Vacant (OAR)
Vacant (OMAO)

Comms Team

Jennifer Fulford
Heather McCullough
Brian Meyer
Douglas Rao

Training Team

Douglas Rao
Chris Slocum

AI-Ready Data Team

Douglas Rao
Tyler Christensen through Jan 2022

NOAA AI WG (NAIWG)

Technical
Guidance

Oversight &
Guidance

NOAA AI Executive
Committee (NAIEC)

Reporting &
Annual Report

NOAA Center of AI (NCAI)

*Changes to NAIEC January 2022

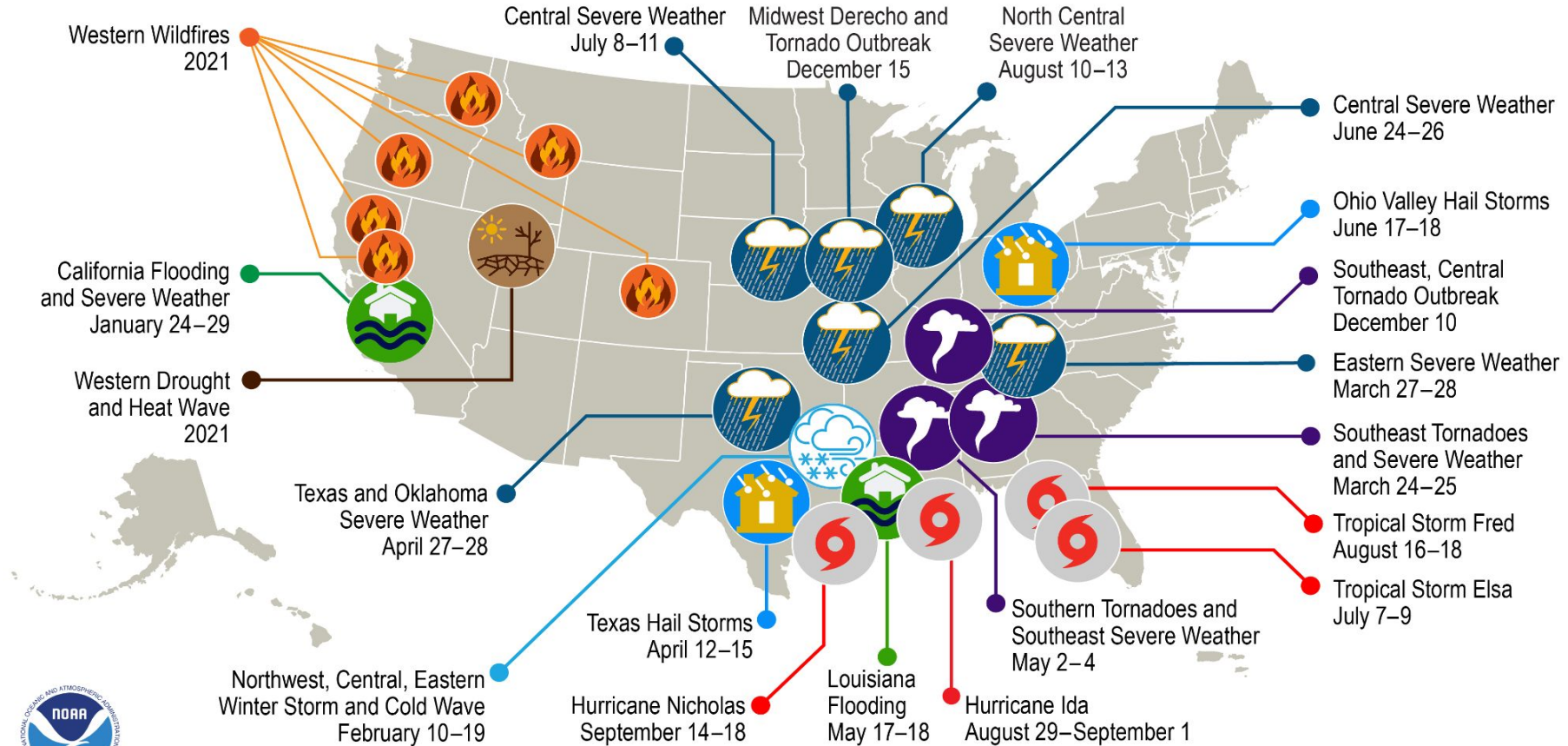
Tyler Christensen replaces Greg Dusek as NOS rep
Ben Richards replaces Greg Dusek as co-Chair



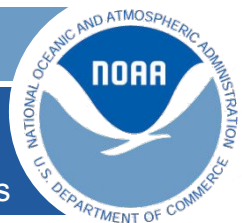
Outline

- Background
 - NOAA's Mission Space
 - Drivers for AI in NOAA
- NOAA's Center for Artificial Intelligence
 - Developing and Applying an "AI-ready" standard
 - Training the Workforce
 - Engagement and Partnerships - Leveraging cross sector investments
 - NOAA Community of Practice
- Summary and Next Steps

U.S. 2021 Billion-Dollar Weather and Climate Disasters

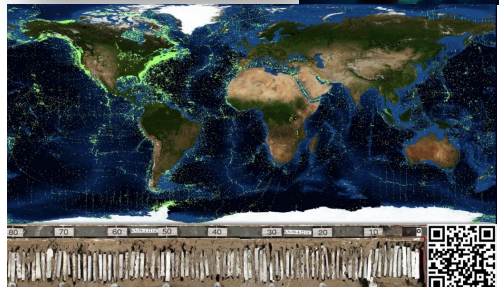
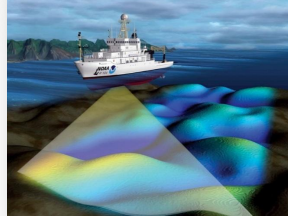
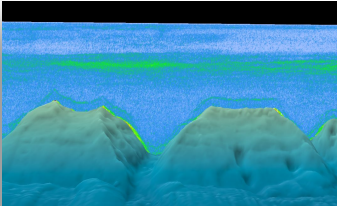
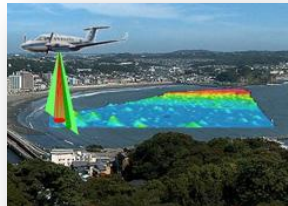
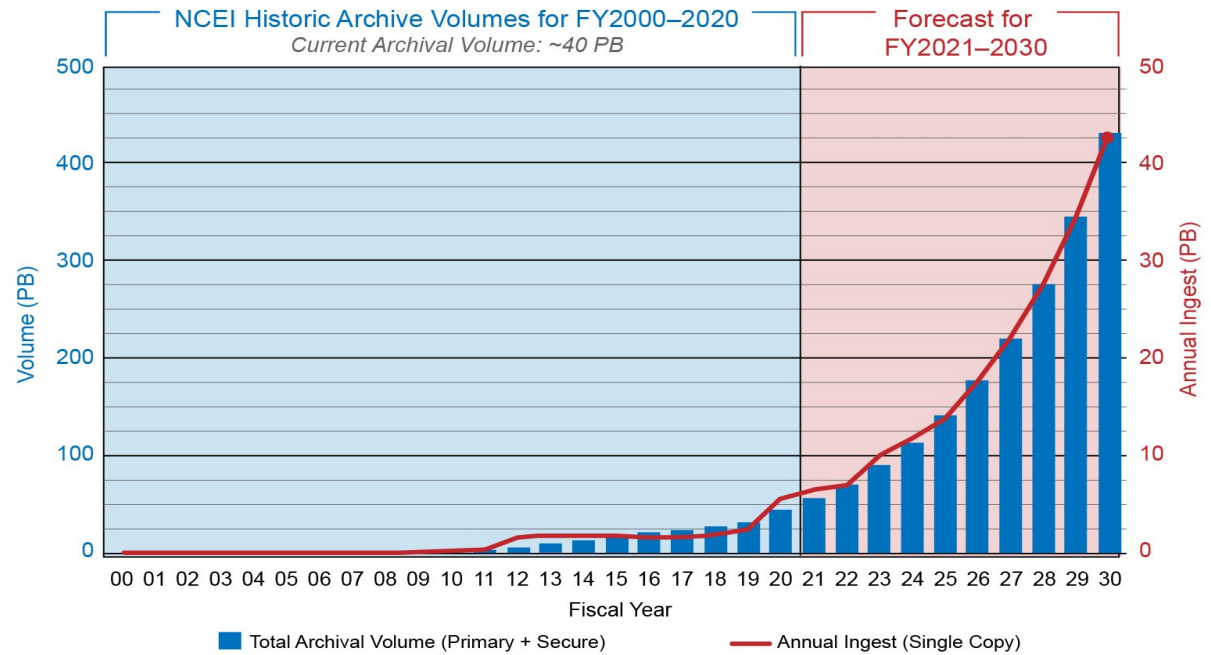


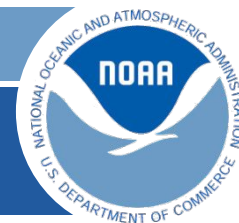
This map denotes the approximate location for each of the 20 separate billion-dollar weather and climate disasters that impacted the United States in 2021



NCEI Archival Volume History and Forecast

Increasing Data Volumes from Station, Model, Radar, UxS, Acoustics, 'Omics, and Satellite Sources

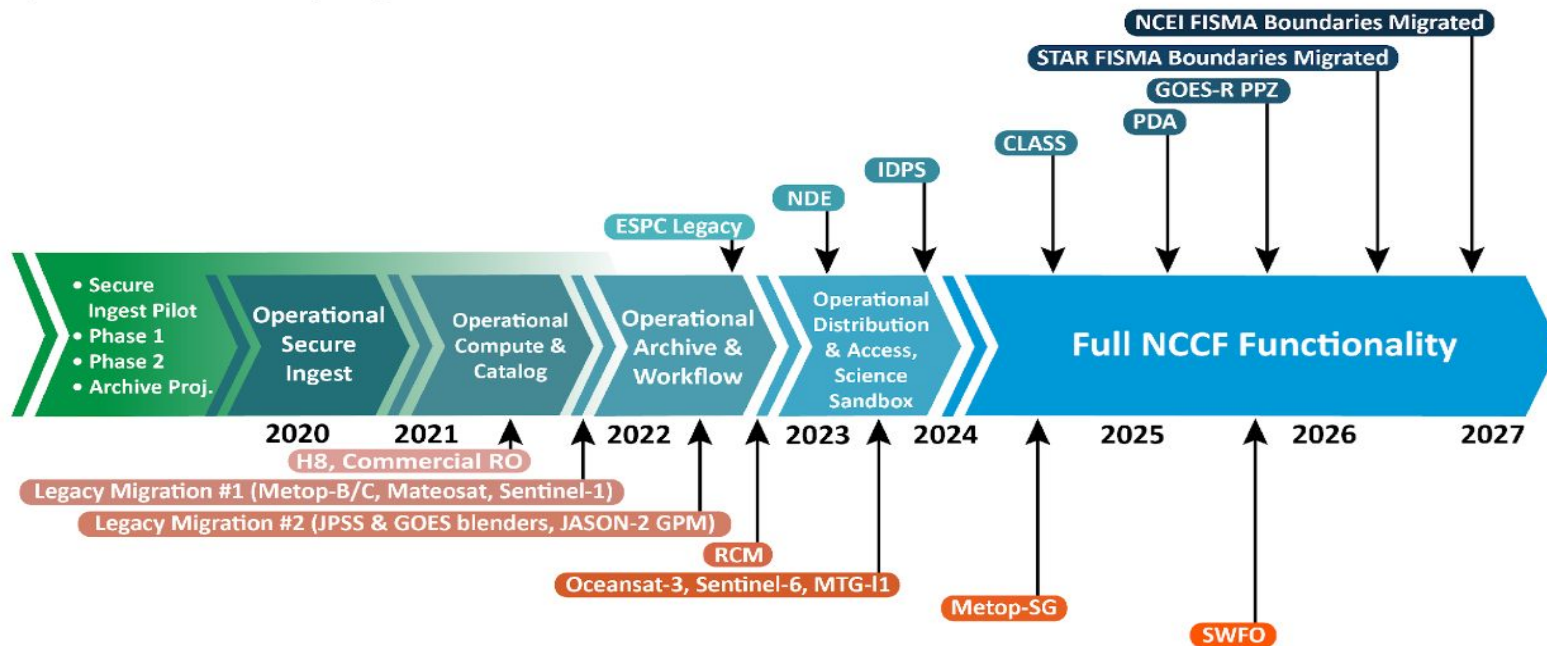




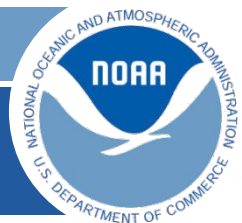
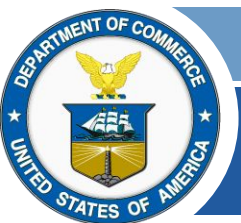
NOAA Cloud Migration

NESDIS Common Cloud Framework Roadmap

System Functionality Migration to NCCF



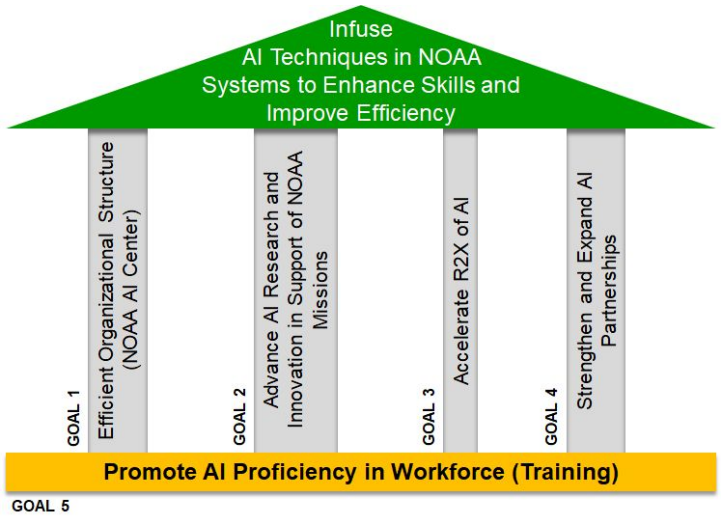
Data Stream into NCCF Operational Functions

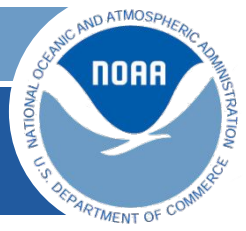


Background

National AI Initiative Act of 2020:

“The Administrator of the National Oceanic and Atmospheric Administration (hereafter referred to as “the Administrator”) shall establish, a Center for Artificial Intelligence”

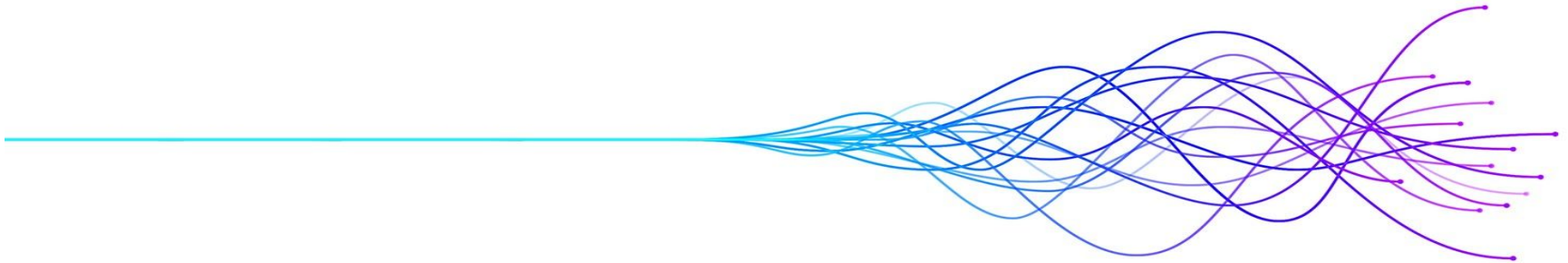


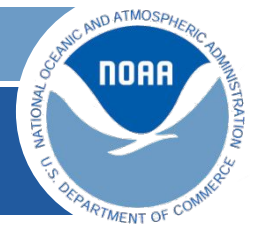


NOAA Center for AI

noaa.gov/ai

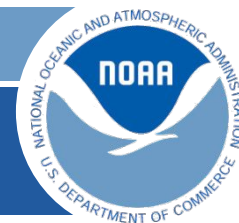
NCAI's Vision - "To empower the proliferation of AI at NOAA by lowering the cost of engaging curiosity through AI applications to capture mission value through its Community of Practice."





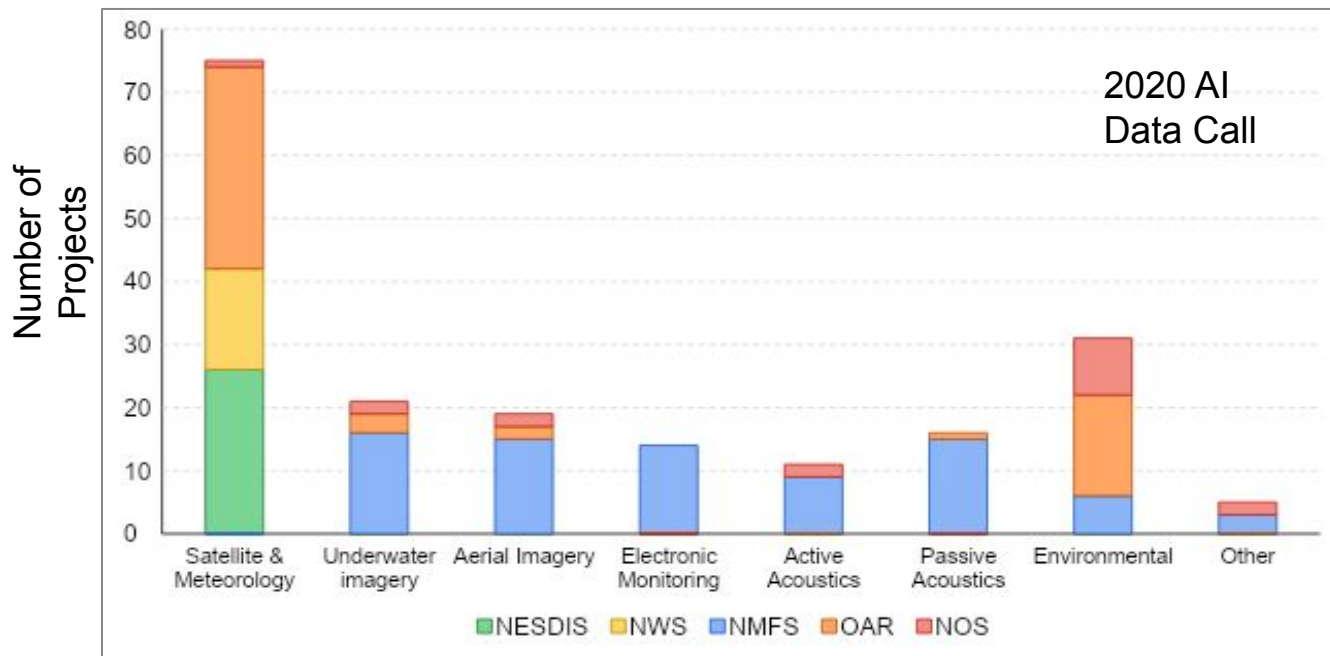
NCAI Initial Efforts

- Understanding the NOAA AI Portfolio
 - Pilots->Operations-> Next Generation
- AI-Ready Data
 - Data deluge → Automating data wrangling → Actionable information
- Training our AI-ready Agency & Workforce
 - Differing Perspectives → Basic Fluency → Efficient R2X
- Partnership Development
 - Cross-sector Investments → Develop Partnerships → Harvesting & Cost Avoidance

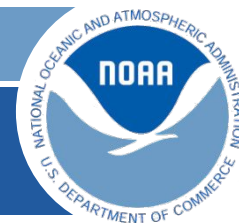


NOAA 2020 AI Data Call – Data Types

NOAA has applied AI-ML to a variety of environmental data demonstrating its interdisciplinary research and operational capabilities in support of its cross-functional mission requirements.

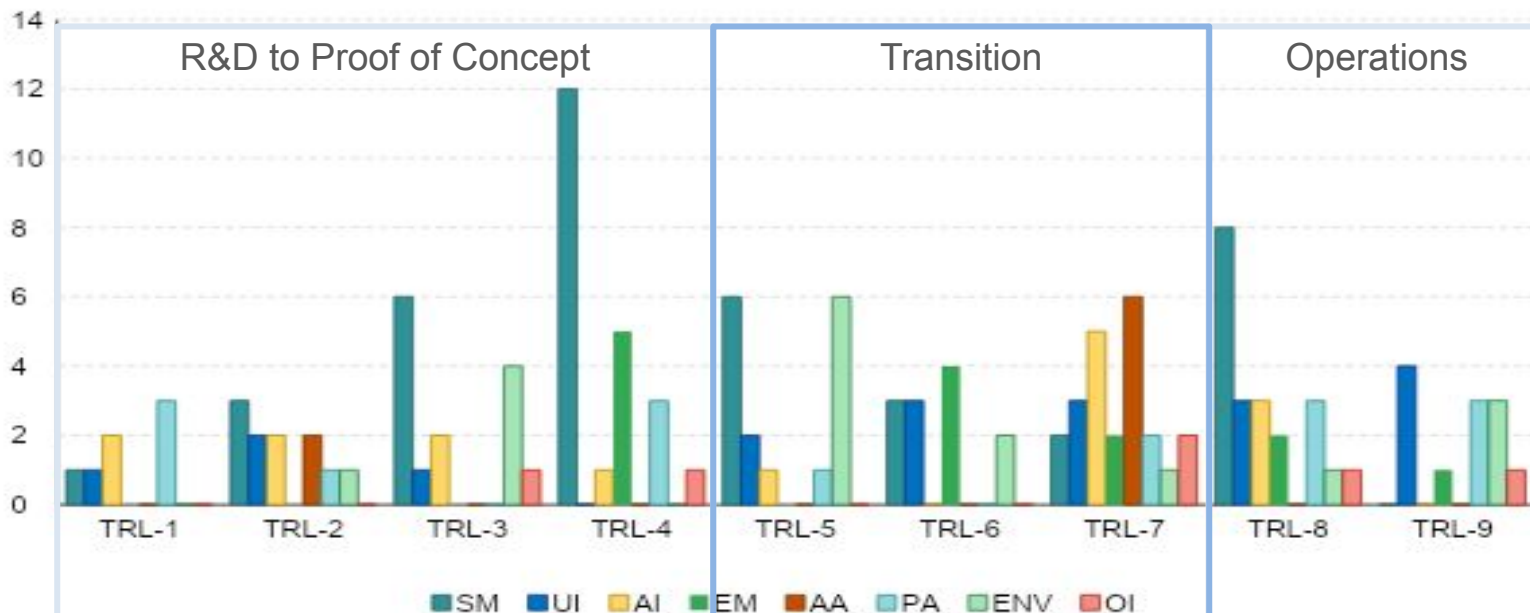


Analyzing the results of the 2022 Call is In Progress



NOAA AI Data Call – Data types & TRLs

NOAA has made progress in transitioning the applications of AI-ML into operational efficiencies, as shown by the Technology Readiness Levels (TRLs). About 37% of the NOAA AI projects are in transition, while 25% reached the operational phase.



SM (Satellite/Weather), UI (Underwater Imagery), AI (Aerial Imagery), EM (Electronic Monitoring), (AA) Active Acoustics, (PA) Passive Acoustics, ENV (Environmental Observations), OI (Other imagery)



Advancing AI Research through Life Saving R2O

Detecting Rip Currents in Coastal Webcam Imagery

National Ocean Service

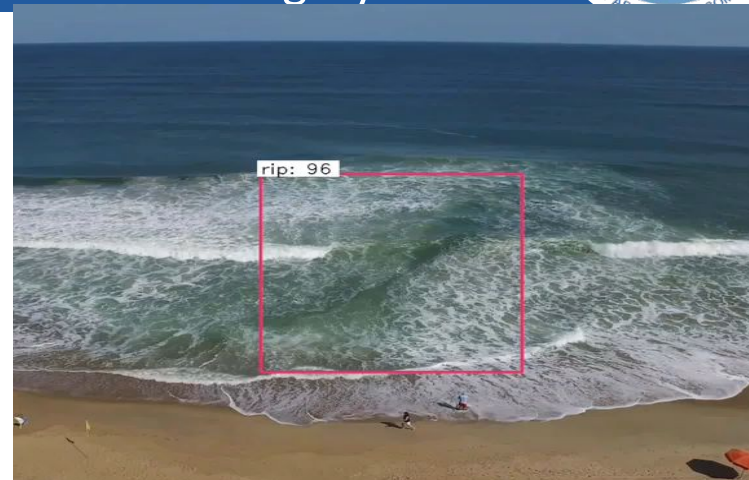
Objective: Detect rip currents in coastal webcam imagery using regional convolutional neural networks

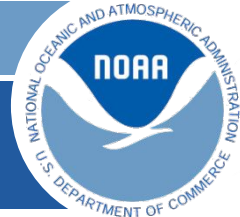
Benefits: The algorithm will be applied to webcam imagery collected by the NOAA-funded WebCOOS camera network in the Southeastern US. The results will support implementation and improvement of the NOAA rip current forecast model

Partnerships: UC Santa Cruz; UNC Wilmington; U of South Carolina; SECOORA; USGS

De Silva, A., I. Mori, G. Dusek, J. Davis and A. Pang (2021), Automated rip current detection with region based convolutional neural networks, Coastal Engineering, <https://doi.org/10.1016/j.coastaleng.2021.103859>

Principal investigators: Gregory Dusek (gregory.dusek@noaa.gov), Alex Pang (pang@soe.ucsc.edu)





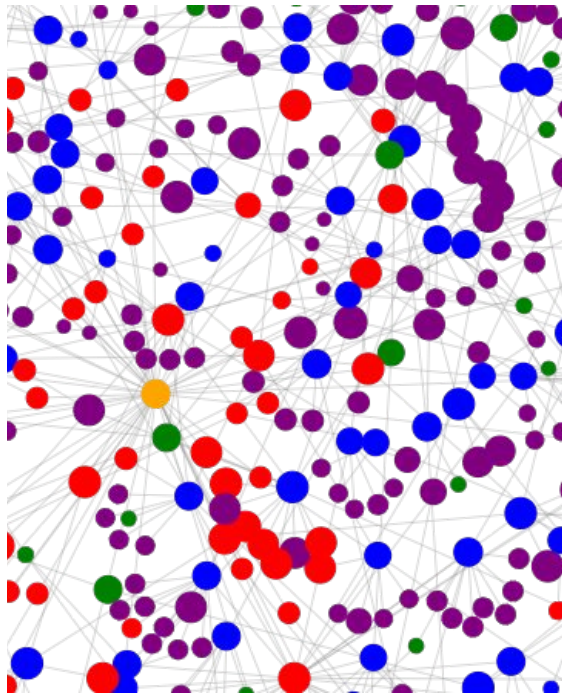
Example: Natural Language Processing

Machine Learning technique where algorithms identify patterns and context of words to find meaning in unstructured text documents

Example pilot application: A summer [Hollings Scholar project](#) used NLP to improve data discovery by harvesting structured metadata from unstructured text documents, demonstrating improved search precision

Example pilot application: Can machine learning find the link between the words used in research articles and the data referenced in the article?

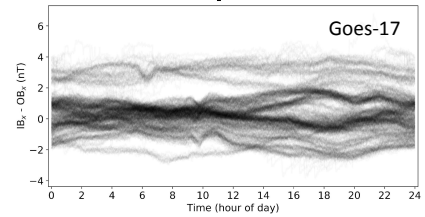
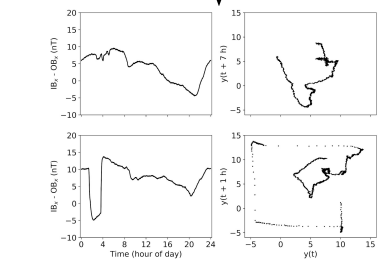
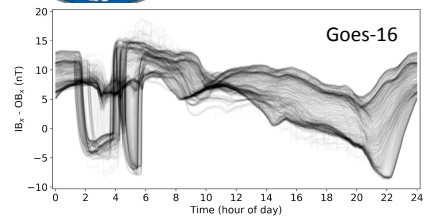
- Collaboration led by the Coleridge Initiative
- [Recent Kaggle competition](#) using NOAA, USDA, and NSF datasets
- Early effort shows promise; [planning phase 2](#) of "Show Us the Data".
- Potential benefits
 - Help researchers find data used in their research topic
 - NOAA understands our data users and shows the value of investing in our data



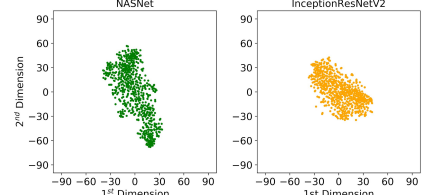
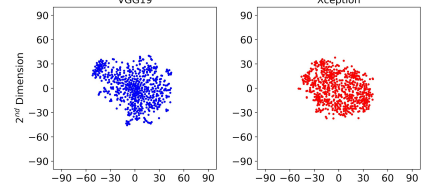
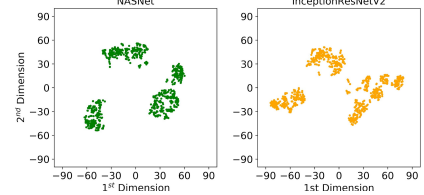
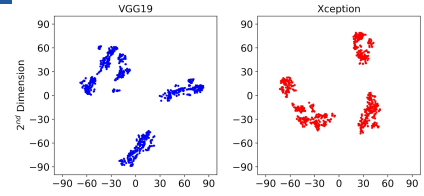
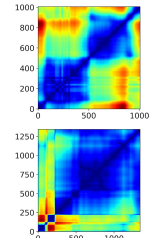


Advancing AI Research through Exploring Critical Problems

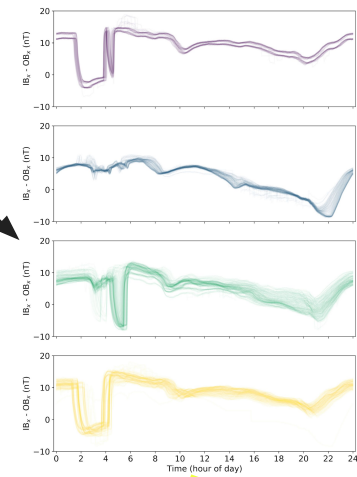
GOES-R Magnetometer Challenges - Classification of Differences



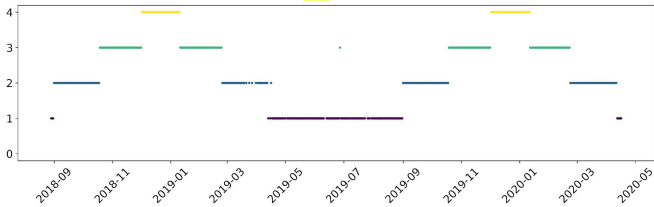
Time series → Phase space → Recurrence Plot



Transfer Learning for feature extraction & Dimension Reduction using TSNE



- Observed differences can be clustered into four main classes,
- Temporal distribution of classes points to major orbital effects on the measured magnetic field strengths by Inboard and Outboard Magnetometers,
- This result could potentially be used to build a step function to correct GOES-R MAG data that is used to provide real-time warnings of geomagnetic storms.



→ K-Means Clustering → How clusters are distributed in time



AI-Ready Data

“The most common reason I can't use a dataset is just that it doesn't have the right content, either in terms of what variables are in it, or its coverage, or its resolution, or the purpose it was created for, so having all of that information readily available speeds up discovery a lot.”

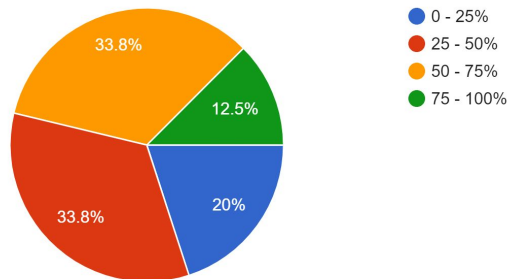
“Complete and consistent data sets are even more important in AI/ML projects”



AI-Ready Data: Why?, What? and How?

In your typical AI/ML application development, roughly what percentage of your time do you spend on finding, accessing, and preprocessing data?

80 responses



Almost half of respondents spend **at least half of their time on data wrangling**, before they can get to work on the science questions they are trying to answer.

Goal: users spend less time data wrangling, more time on AI / ML

How can data users find data that is easy to use in AI/ML?

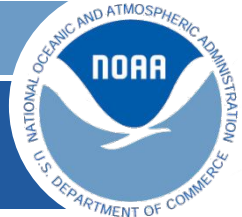
How can data providers assess and improve usability?

What's needed:

- Specific community driven definition of AI-readiness requirements
- Assessment tools for data providers
- Way to represent readiness level so providers can report data readiness and users can compare
- Feedback and iteration to improve the standard
- Ideally, a formally published standard (or set of standards)

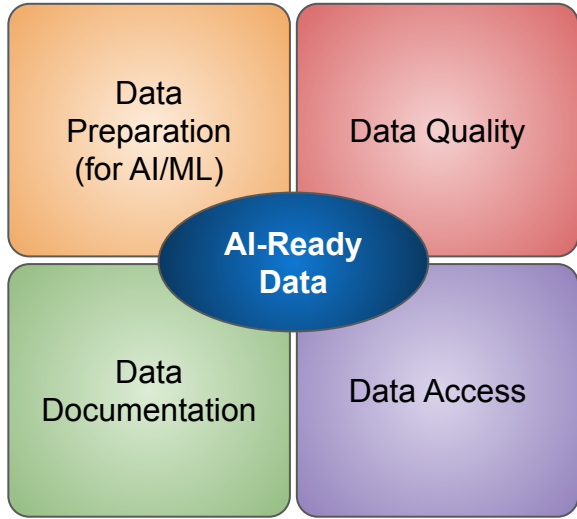
NOAA is leading a collaboration under the Earth Science Information Partners (ESIP) working to develop the standard. Membership includes US Federal agencies, universities, NGOs, private sector, and international

- https://wiki.esipfed.org/Data_Readiness



AI Ready Data Survey

Req's for Open Environmental Data → Enable AI Applications



FY-22 Delivery: Develop a preliminary AI-ready data standard by engaging across NOAA and external stakeholders via ESIP, and workshops. Present the preliminary standard at AMS, AGU or ESIP Winter Mtg (FY22 Q2). Test the standard against a pilot set of data sets (FY22 Q3). Include the standard in at least 1 call for proposals or funded opportunity (FY22 Q4).

What makes a data set "AI-Ready"?

What usability improvements should providers prioritize?

Survey Categories and Sample Questions:

Demographic / Background - sector & research domain

Data Preparation - Gap filling, gridding, outliers, labels

- e.g. Which of these data preparation factors is most important for your most common application needs?

Data Quality - Completeness, consistency, bias, provenance

Documentation - Metadata, DOIs, example code

- e.g. Which of these data documentation factors is most important for your most common application needs?

Data Access

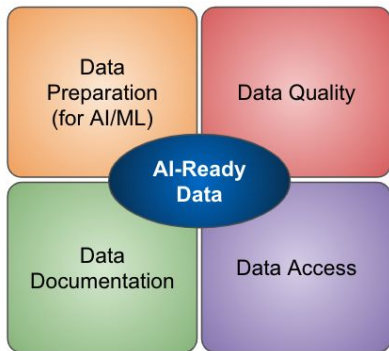
- e.g. Which file formats can you work with in your AI/ML applications? Which do you prefer?

Training Data Reuse - Sharing labeled datasets



AI Ready Data Survey

Req's for Open Environmental Data → Enable AI Applications



Initial findings January 2022:

(93 responses to date with most from USG, 9% NGO, 12% Private)

Data Preparation: Outliers included & tagged, gridded in space & time, labeled targets

Documentation: Metadata w/details about all parameters, example code/Notebooks, and information about space/time extent,

Data Quality: Consistency, Completeness, Resolution, Lack of Bias

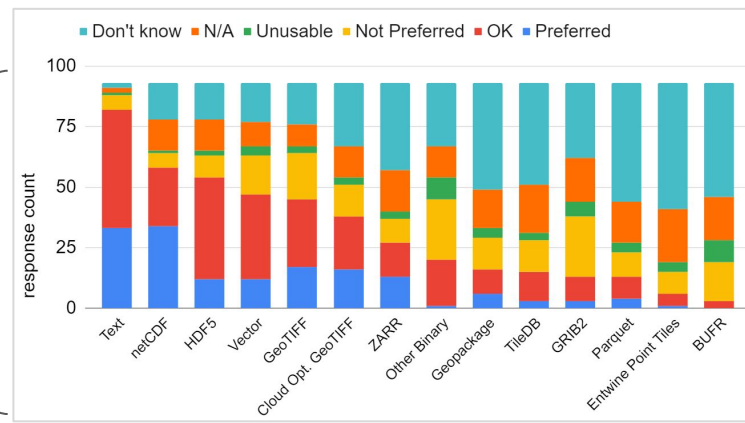
Data Access: Cloud, File download, API are fairly evenly split

Training Data Re-Use: 58% published their training data, and 50% used training data from another group

What formats can you work with for AI/ML?

- Self-describing formats preferred
- Text formats (e.g. csv) also good
- Some prefer cloud-optimized

Flexible: 67% can handle 4 or more formats



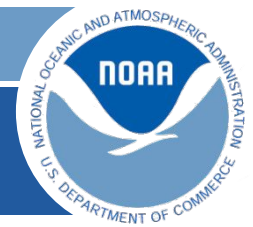


Training our AI Ready Agency & Workforce



“Tell me and I forget, teach me and I may remember,
involve me and I learn.”

– Benjamin Franklin



Assessing NOAA's training resources & requirements

NOAA AI Strategic Plan:


“complete annual assessment of NOAA AI training resources and requirements”

(Strategic Action 5.1.1; starting FY2022)

NCAI Goals for Survey:

- Prioritize training needs
- Identify gaps & develop training plan
- Engage Community of Communities of Practice in development of training plan

(Draft NCAI Actions 5.1.1. A and B)

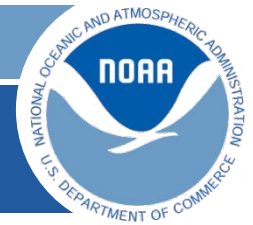


NCAI Resource Assessment

This is a request for information on NOAA artificial intelligence (AI) specific to workforce development. The information collected here will assist in ranking training/resource priorities, focusing development of NOAA-specific training needs and activities, and highlighting existing AI training material and toolkits across the agency.

This assessment has three components, and you may select any of the three to contribute. The broad topics are:

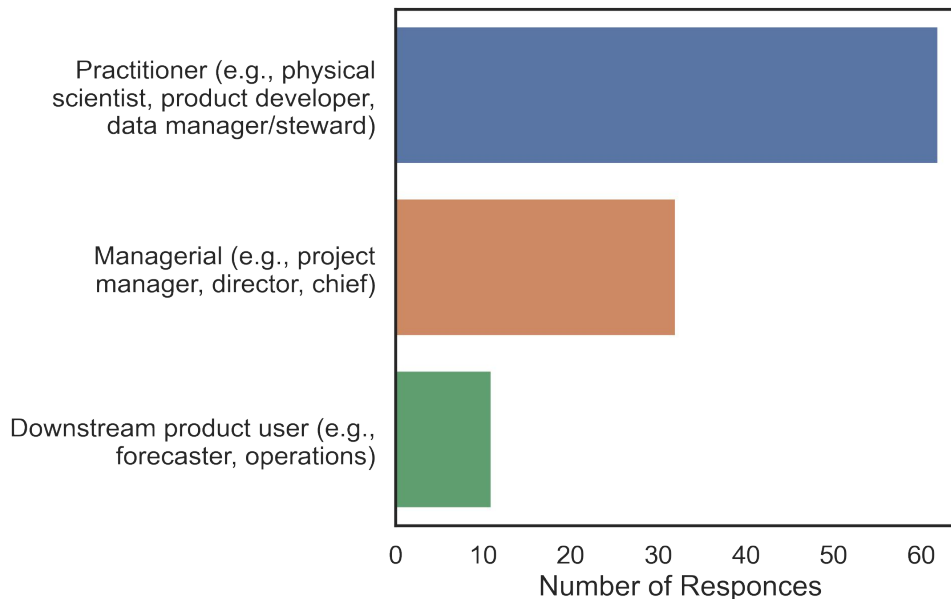
- Training priorities input: help NCAI rank priorities for identified training needs based on community of practice input.
- Specific training needs/resources: if NCAI missed a critical training priority, submit a new need.
- Existing resource awareness/contribution: if your office has a resource, please alert NCAI to its existence.

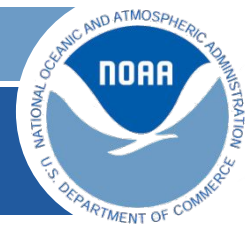


Assessing NOAA's training resources & requirements

High Level Summary

- 81 respondents
 - 75% – federal employees
 - 25% – affiliates
- 75% of the respondents are AI/ML practitioners
- ~40% respondents with managerial roles





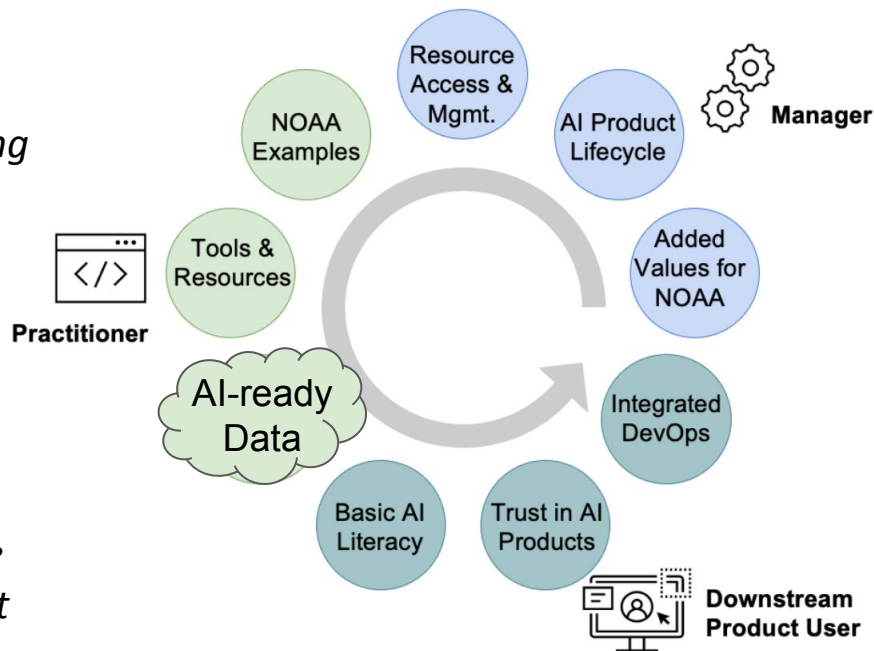
Powering Discovery and Innovation with Training, Trustworthy, and Equitable AI-ready Services

Training + AI-ready data for Trustworthy + Equitable AI-ready Services

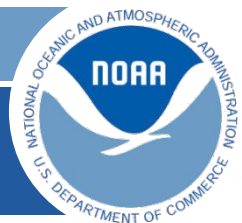
Factsheet: <https://www.noaa.gov/ai/training>

Urgent: need NOAA-specific training material using NOAA data and computing resources to **remove common barriers** to the “Research to Operations, Applications, and Services” pipeline.

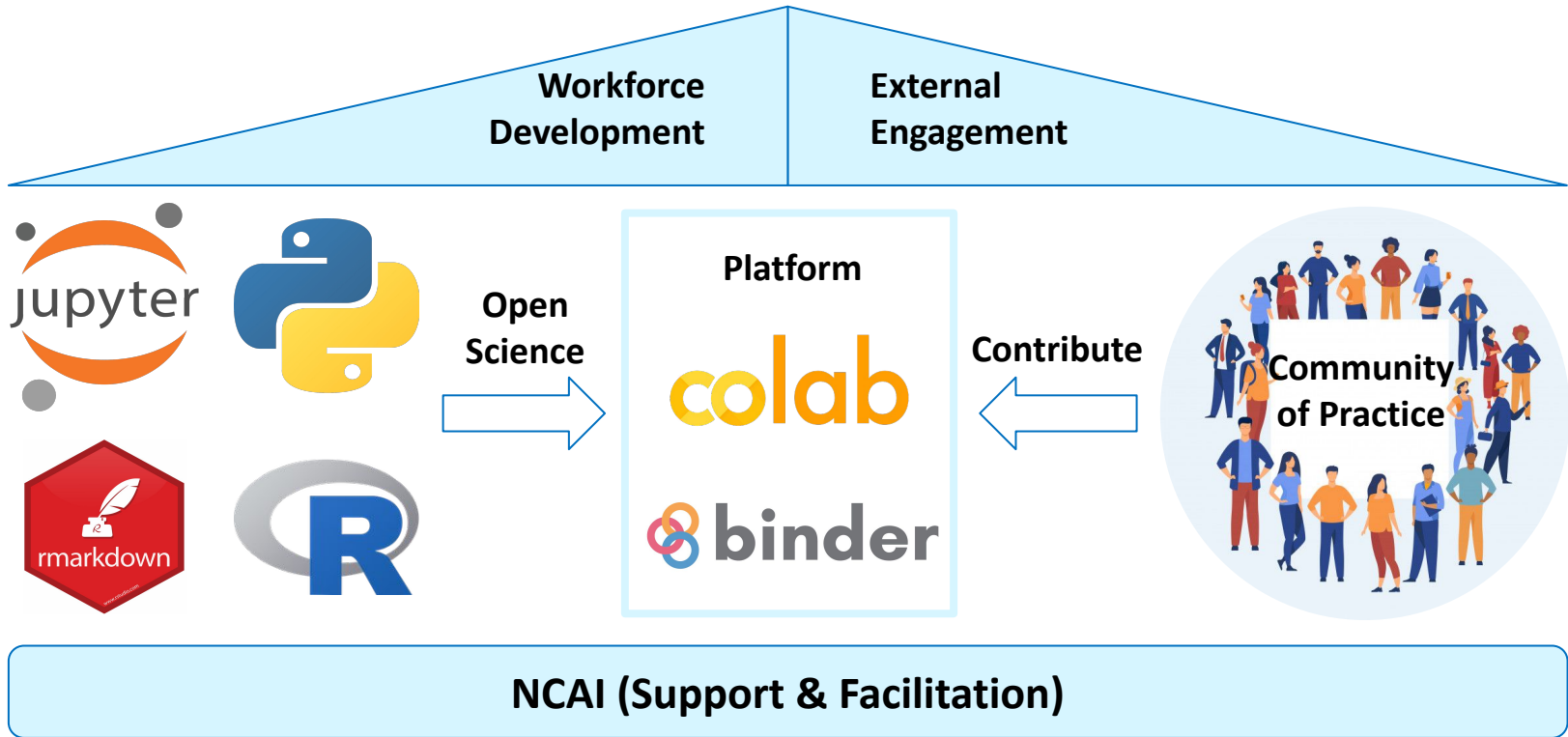
*To address needs, resource creation should be prioritized to **convert NOAA AI success stories** into interactive training material in a **sandbox computing environment** that allows the workforce to immediately apply learning outcomes to support NOAA’s mission via the AI strategic plan.*

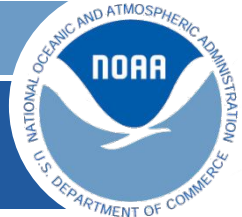


NOAA training action priority lifecycle highlighted by ²² workforce role and relationship to AI. ([noaa.gov/ai/training](https://www.noaa.gov/ai/training))



A Flexible Training Framework Driven by Open Science





Learning Journeys to Empower Diverse Learners

Beginner Users

- No previous background
- Need comprehensive info about problems & overall workflow

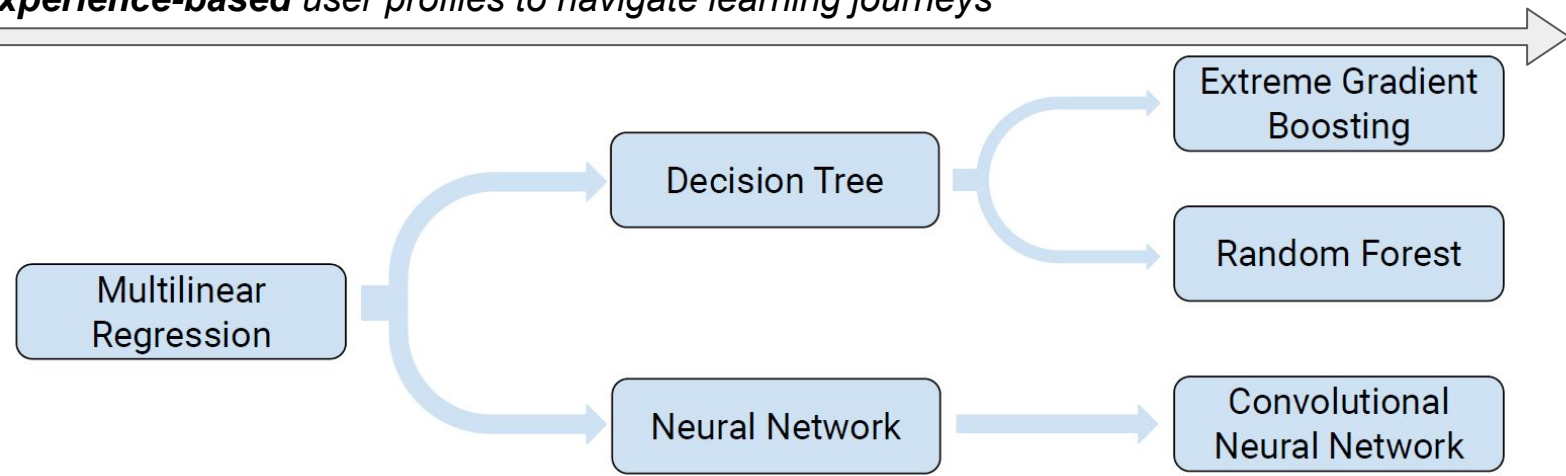
Intermediate Users

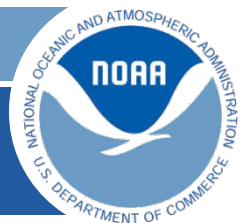
- Have basic knowledge & experiences
- Want to learn advanced AI/ML tools for applications

Advanced Users

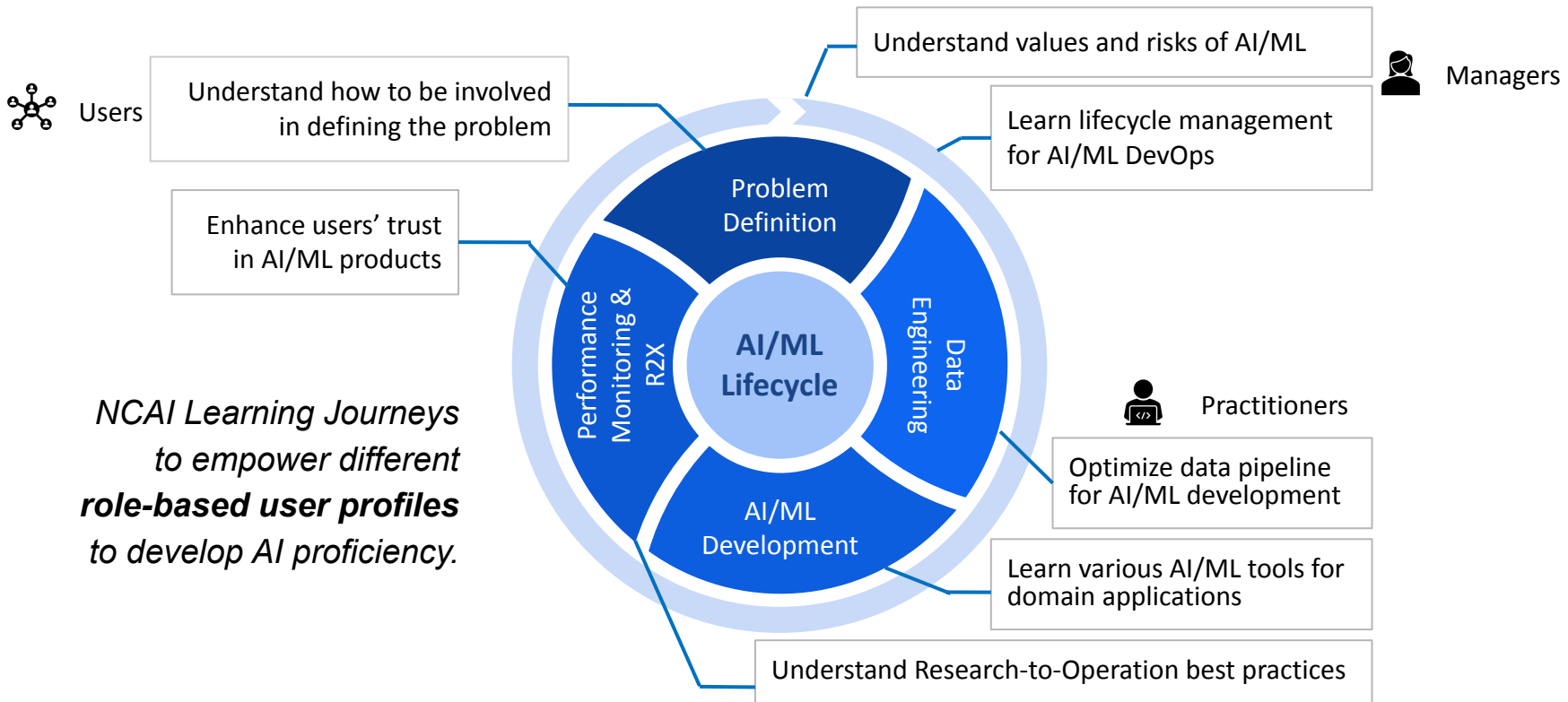
- Experienced in AI/ML applications
- Want to keep up with tools & best practices

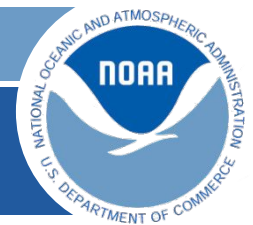
Experience-based user profiles to navigate learning journeys





Learning Journeys to Empower Diverse Learners





Tools Development to Empower the Community

Jupyter notebook template with guidebook

External_Collaborative_Training_Notebook_template.ipynb

File Edit View Insert Runtime Tools Help

Table of contents

- Collaborative Training Notebook Template (note to content creator)
- <> Notebook Title
- Tutorial Material
- Exercises
- Next steps
- Examples in the community
- Data statement
- References
- Metadata
- License
- Disclaimer (Optional)
- Section

Collaborative Training Notebook Template (note to content creator)

The purpose of the Training Notebook Template is to make sure that material has a uniform appearance and that contain similarly structured material across different communities of practices. The main sections are:

- **Overview** of the broad topic covered
 - **Prerequisites** for what background information is needed to go through the notebook
 - **Targeted level** of this notebook
 - **Learning outcomes** from completing the notebook
- The **Tutorial** with a balance of explanation and activity
- **Exercises** for students to try that do not have solutions but maybe have an answer or benchmark to facilitate understanding
- **Next steps**
 - Potential follow on material
 - Other relevant notebooks
- **Examples in the community**
- **Data statement**
- **References**

As you fill out your notebook, make sure to delete the suggestion text.

The notebook template is maintained by Chris Slocum (christopher.slocum@noaa.gov) and Douglas Rao (douglas.rao@noaa.gov).

Notebook readability assessment

Processing: ENS0_Seasonal_Forecasting.ipynb

Notebook metadata

- Language: Python 3
- Number Markdown Cells: 35
- Number of URLs: 7
- Number of valid URLs: 7
- Number Code Cells: 22
- Percent Text Output: 59%
- Percent Display Output: 50%

Readability metrics:

- Flesch reading ease:
 - Score: 46.7
 - Text interpretation: Difficult
 - Grade level: College
- Flesch-Kincaid Grade Level: 12.8

Reading time estimates:

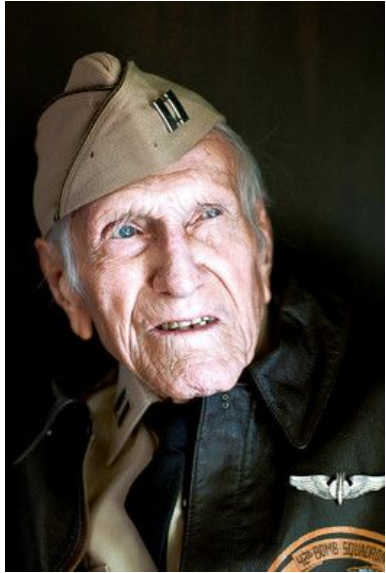
Estimated text reading time:	13 to 25 min
Estimated code reading time:	10 to 20 min
Estimated code annotation reading time:	4 to 7 min
Estimated total reading time:	23 to 45 min

https://github.com/ESIPFed/earth-science-community-ML-tutorials/tree/main/tutorial_template

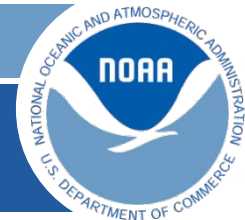
RATCHET - Readability Assessment
Tool for Code that Helps with Effective
Training



Partnership Development



Everyone needs that support-even if at first you don't think you do. Look around. See who's on your side and in your corner. You don't have to go it alone. - *Louis Zamperini*

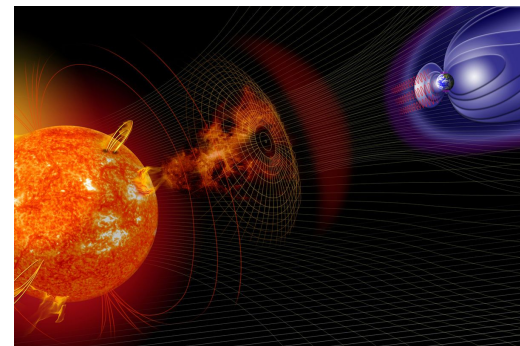
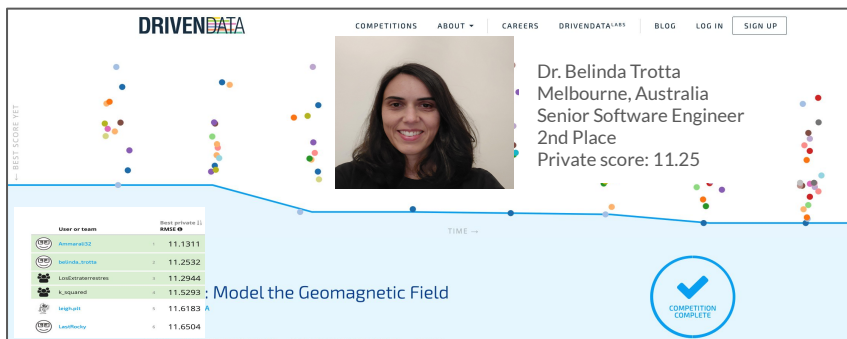


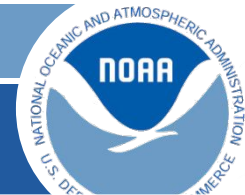
Partnership Development - through Competition

MagNet: Model the Geomagnetic Field - Internationally Open Machine Learning Competition

- *NCEI innovates* 2020 funded [an open data-science competition to improve our ML model](#) to predict *Dst* from solar-wind data, mitigating the impacts of space weather geomagnetic storm events on magnetic navigation systems
 - 600 participants and 1200 model submissions, with prize finalists achieving ~11.1 – 11.5 nT RSME on the private test set, very close to the theoretical limit.
 - NCEI internally validated all four of the winning models
- R2O: Incorporate the winning model in the HDGM-RT (ongoing)
- Training: Documentation - Develop an NCAI Learning Journey interactive notebook (ongoing)
- Partnerships: Sponsored by NOAA, with support from NASA's Center of Excellence for Collab. Innovation (CoECI). Challenge was conducted by DrivenData and [HeroX](#).
 - Long term collaboration with the 2nd place winner, Dr. Belinda Trotta
- Contacts: Manoj Nair (PI), Rob Redmon (Gov't POC)

Article in Room Space Magazine

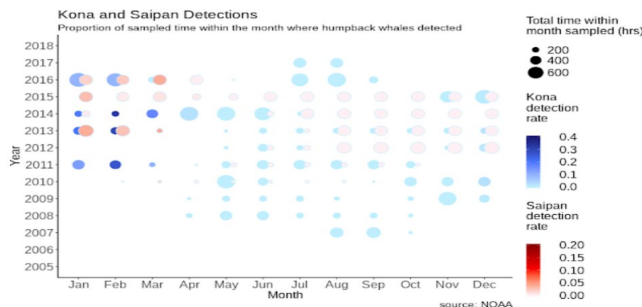




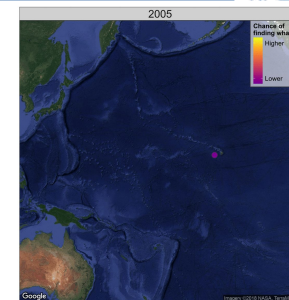
Partnership Development - through Common Goals

OK Google, Find the humpback whales

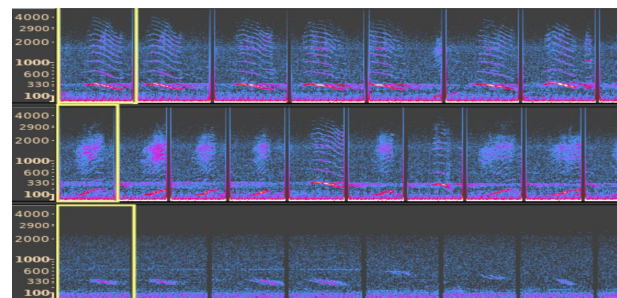
- PIFSC deployed bottom mounted HARPs since 2005
 - 13 Pacific Island sites
 - 200+ TB
- **AI for Social Good**
 - Identify humpbacks
 - Decimated to 9.2 TB
 - 200 kHz -> 10 kHz
 - Classifier for 75 sec segments attains 90% precision at 90% recall for supervised learning
 - Unsupervised learning retrieves similar song units



Time density of presence for Kona and Saipan



Relative call abundance identified by the classifier

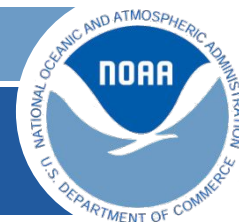


Manually chosen calls (boxed) and nearest neighbors

PIFSC: Ann Allen & Karlina Merkens

Google: Matt Harvey, Jiayang Liu, Julie Cattiau, Aren Jansen, Rif Saurous, & Lauren Harrell





Partnerships Representative Callout

(Examples here - not an exhaustive list)

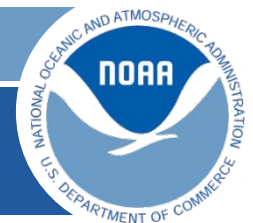
- National Science Foundation
 - [AI2ES](#) - Research on Trustworthy AI in Wx, Climate, & Coastal Oceanography, AI4ES Summer
 - AI-Ready data (via [ESIP Data Readiness](#)); Explainable AI; Trustworthy and Responsible AI
- Industry
 - Exploring OTAs, CRADAs and other opportunities to build on public private investments and partnerships; and develop our workforce.
- UKMet Joint Centre for Excellence in Env. Intelligence
 - Exploring co-developed Learning Journeys, Digital Twins

Partnerships are built around the relative strengths of the partners

Pilot Projects	WWCB Societal*	AI-Ready	Learning Journey	Training Data
<p>Combining the power of citizen science and artificial intelligence to make new discoveries in our ocean soundscapes</p> <p>Crowdsourced annotation of acoustic Big Data and streamlined application to AI models will decrease latency in developing data products critical to managing the Blue Economy empowering citizen scientist exploration, leading to AI models to address NOAA science and management questions.</p>	1			Y



NATIONAL OCEANIC AND ATMOSPHERIC ADMINISTRATION



Transforming Weather, Climate Services, and Blue Economy with Artificial Intelligence

203

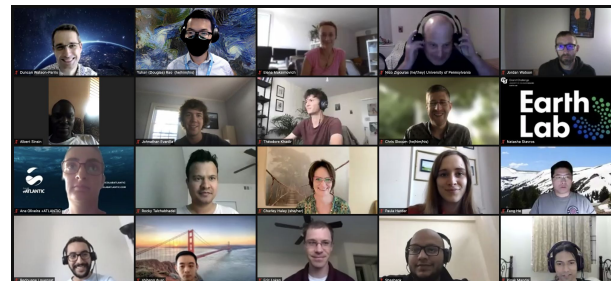
Submitted Abstracts

from government, academia, private industries, and international partners

1325

Registered Participants

from 34 countries



~ 50%

are students and early- career professionals

~ 60%

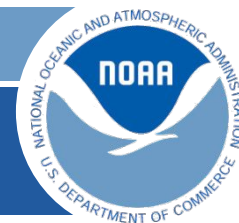
are non-NOAA participants

~175

international participants from outside of U.S.

~150

registered hackathon participants



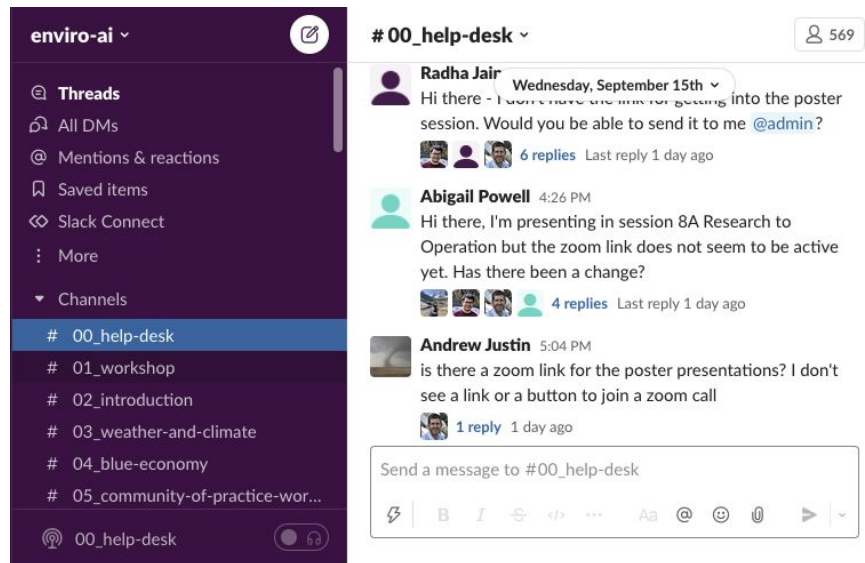
Partnership Development Through Workshop Engagement

580

***Slack Workspace
Members***

3381

***Messages in Slack
during the
workshop***



193 messages

#00_help-desk
Asking for help for workshop

118 messages

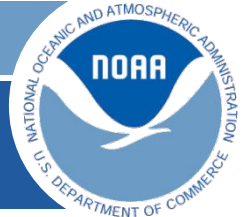
#02_introduction
Creating networking
opportunities

435 discussions

#03_weather-and-climate
Active discussion for AI research
in weather and climate

156 discussions

#10_hackathon
Active engagement during
hackathon events



NOAA's 4th AI Workshop Ideation

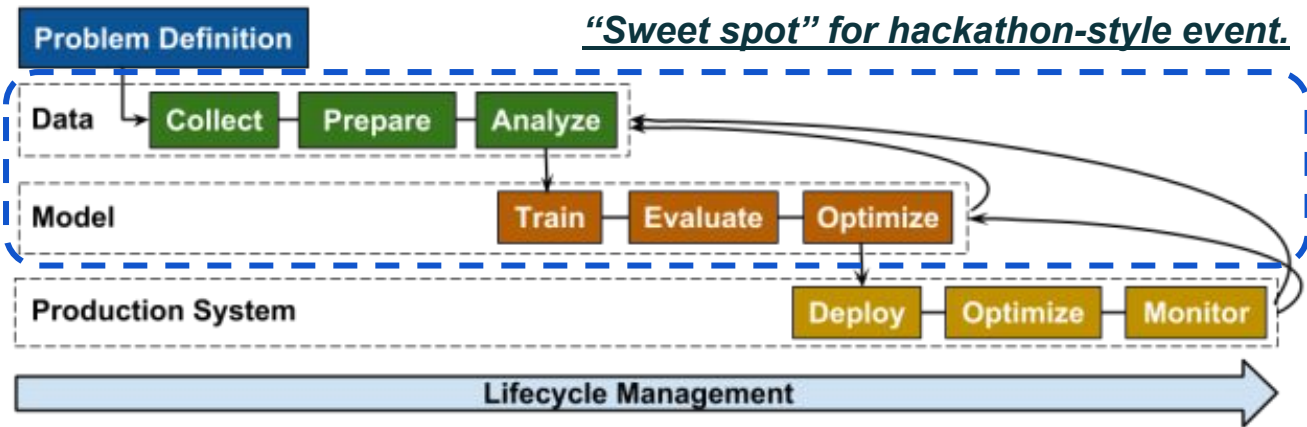
Bridging the Gap Between Awareness and Actions

In 2022, the 4th NOAA AI Workshop will be a Hackathon approach.

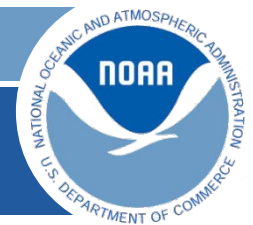
Improving **AI-readiness** for NOAA data

Addressing NOAA priorities through **co-development**

Infusing **NOAA priority areas** in problem definition process by **engaging with diverse stakeholders**.



Facilitate **workforce training** & **partnership** to accelerate AI adoption.



Connect with Us!

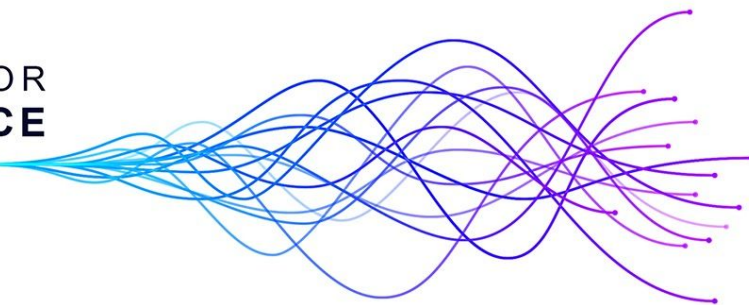
Join our *Public Community*

- **NEW!** noaa.gov/ai
- *Connect with NOAA's Community of Practice around AI for Earth system science*
- Mailing List: <https://tinyurl.com/y2ehvhfg>
- Contact the NCAI Team: ncai-team@noaa.gov

Inside NOAA's Ecosystem?

- <https://sites.google.com/a/noaa.gov/ncai/>

NOAA CENTER FOR
ARTIFICIAL INTELLIGENCE



Help add to our
AI Collaboration
User Stories
on noaa.gov/ai

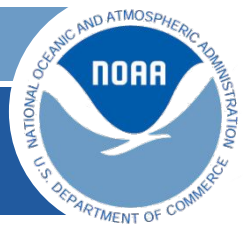
News and events



3rd NOAA Workshop on
Leveraging AI in Environmental
Sciences



NATIONAL OCEANIC AND ATMOSPHERIC ADMINISTRATION

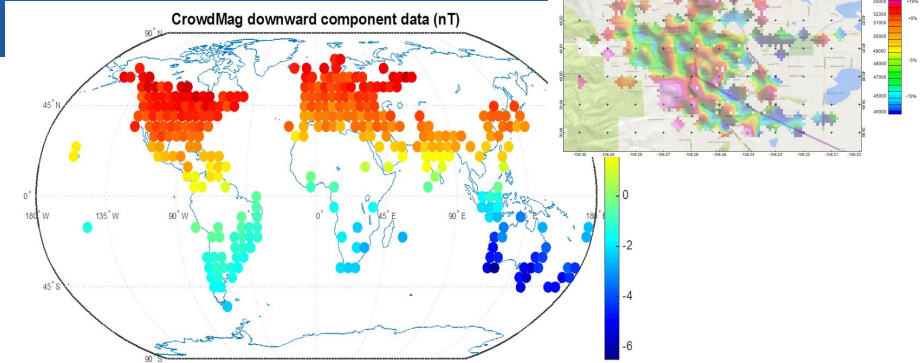


Backup Slides



CrowdMag

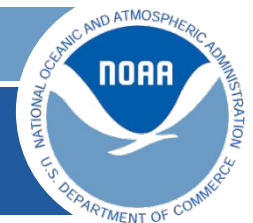
- NCEI develops and hosts the CrowdMag crowdsourced magnetic field smartphone app
 - From passive observing and background data collection to “Magtivities” to work within OS restrictions
- CrowdMag has over 12k participants on 6 continents
- Create multiple scales of magnetic field models
 - City/Regional and Global
- Flight Mode in development to turn commercial flights into aeromag data collection surveys
- ***Developing AI/ML algorithms to help clean data and identify biases based on phone make/model***



Help Improve Navigation for The World

CrowdMag
The First Crowd-Sourced Geomagnetic Mapping App

Download on the App Store
GET IT ON Google Play



NOAA's Center for Artificial Intelligence

Team

Rob Redmon (NCAI Lead)

Jordan Watson (Deputy, LANTERN)

Eric Kihn (NESDIS AI Representative)

Douglas Rao

Chris Slocum

Jennifer Fulford

Heather McCullough

Brian Meyer

Jebb Stewart

Ben Richards

Paul DiGiacomo

Ken Casey

Dave Fischman

Huai-min Zhang

Stephanie Herring

Tyler Christensen (AI-ready, LANTERN)

*Focusing on: Workshop and Engagement,
Training, AI-Ready Data, Strategy*

Coordination across NOAA Line Offices

NOAA AI Executive Committee

S&T Synergy Committee

NOAA AI Working Group (onboarding)

Foundations

[National AI Initiative Act of 2020
\(DIVISION E, SEC. 5001\)](#)

"The Administrator of the National Oceanic and Atmospheric Administration (hereafter referred to as "the Administrator") shall establish, a Center for Artificial Intelligence"

Goals

Goal 1: Efficient Governance

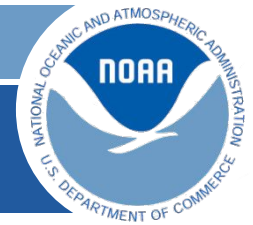
Goal 2: Advance AI Research

Goal 3: Accelerate R2X

Goal 4: Partnerships

Goal 5: AI Workforce Training





Background - AI Initiative Act of 2020

The [National AI Initiative Act of 2020 \(DIVISION E, SEC. 5001\)](#) became law on January 1, 2021, providing for a coordinated program across the entire Federal government to accelerate AI research and application for the Nation's economic prosperity and national security.

- *"IN GENERAL.—The Administrator of the National Oceanic and Atmospheric Administration (hereafter referred to as "the Administrator") shall establish, a Center for Artificial Intelligence"*
- *"AUTHORIZATION OF APPROPRIATIONS.—There are authorized to be appropriated to the Administrator to carry out this section \$10,000,000 for fiscal year 2021."*

This legislative achievement had significant support from elected officials, e.g. Rep. Mike Garcia (R-CA) on the bill *"the bill invests in America's military families, readiness and modernization. I'm particularly pleased SST priorities are represented in the bill including my amendment to create a dedicated center for Artificial Intelligence at NOAA, helping detect and prevent wildfires."*

The White House's Office of Science and Technology Policy (OSTP) launched the [National Artificial Intelligence Research Resource Task Force \(NAIRRTF\)](#) on June 10th, and [RFI on the Implementation Plan \(due Oct 1\)](#).



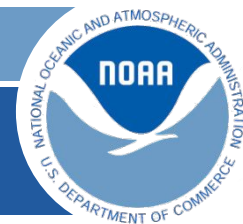
NCAI Fully Resourced Capabilities

Status: Developed a proposal framework for FY22+, pending FY24 PCS; exploring IJA and WWCB connections.

Once fully resourced, NCAI will achieve the following Strategic Goals:

- **Goal 1: Organization & Process - *Stand up Congressionally Authorized NCAI***
 - Program Office coordinated with Public and Private Partners (modeled on RESTORE);
 - AI expertise embedded in each LO supporting Mission Scientists;
- **Goal 2: Advance AI Research and Innovation in Support of NOAA's Mission**
 - Stimulation of AI outcomes across all mission areas with long-term impacts via Grants and Partnerships
- **Goal 3: Accelerate the Transition of AI Research to Applications (R2X)**
 - Bridging the R2X “valley of death” with a fully curated repository of AI software, apps, and policies on ethics, mission validation metrics, ops reqs and an AI App Handbook;
- **Goal 4: Strengthen & Expand Partnerships**
 - A robust and fully realized AI partnership program to leverage capabilities from commercial, academic and government partners.
- **Goal 5: AI Proficiency**
 - AI fully capable workforce established through widespread benchmark AI-ready data, Learning Journeys library, multiple developmental sandboxes, and professional training





Advancing AI Research - Representative Pilots

Pilot Projects	WWCB Societal*	AI-Ready	Learning Journey	Training Data
Valuing NOAA's Data with User Publications Use Natural Language Processing (NLP) to value NOAA's data and products, and automate stewardship efforts, by finding uncited or inconsistently cited data in journal articles.	1,2,3,4,5,6	Y		
Downscaling of Global Climate Projections to Regional Scales Advance Trustworthy and Equitable AI for Climate data services for fisheries management by exploring AI methods to relate global forcing to the regional response, trained on our existing, dynamically downscaled results.	5(?)		Y	
Sea ice detection in Sairdrone images Reduce Sairdrone collisions with sea ice floes by assessing the reproducibility and integrity of current convolutional neural net methodologies for detecting ice in images captured by the drones.	1	Y	Y	Y
Improvements to Numerical Weather and Climate Prediction Systems Use AI to speed up and improve model physics within the Global and Regional Forecast System by neural network emulation of components of Global Forecast System (GFS) physics suite.	5		Y	

*WWCB Societal Challenges:

1. Changing Oceans

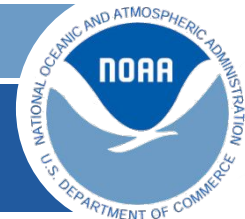
3. Effects of Space Weather

5. Monitoring and Modeling for Climate Change Mitigation

2. Coastal Resilience

4. Extreme Events & Cascading Hazards

6. Water Availability, Quality, and Risk



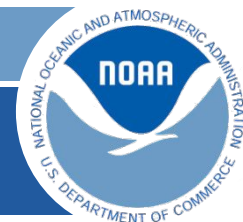
AI Ready Data - Representative Pilots

Pilot Projects	WWCB Societal*	AI-Ready	Learning Journey	Training Data
New Training Dataset for Tropical Cyclones This project would provide a case study for how NOAA datasets can be evaluated and updated for AI readiness before being made publicly available by NCEI.	2, 4, 5(?)	Y	Y	Y
Cloud optimized Water Column Sonar Develop a data lake of archived water column sonar data translated into cloud-friendly formats to improve interoperability and scaled processing. This effort will lower the learning curve to process the data more easily, especially in a cloud environment.	1	Y	Y	
Towards Ops - Conventional Observation Reanalysis (COrE) Distributing COrE, via the cloud, in AI-Ready formats will hasten the adoption process and speed access for transition of ongoing operations and developmental work.	5	Y	Y	
Improving Seal Detection Accuracy in Aerial Imagery Refining and testing algorithms so that we can produce critical high-quality datasets for marine mammal assessments.	1	Y	Y	

*WWCB Societal Challenges:

1. Changing Oceans
3. Effects of Space Weather
5. Monitoring and Modeling for Climate Change Mitigation

2. Coastal Resilience
4. Extreme Events & Cascading Hazards
6. Water Availability, Quality, and Risk



Research to Apps (R2X) - Representative Pilots

Pilot Project	WWCB Societal*	AI-Ready	Learning Materials	Training Data
AI Quality Control of NOAA Water Level Observations Water level processing system to provide near real-time QC'd tide gauge data, accelerating use for downstream products (e.g. Sea Level & High Tide Flooding monitoring).	2, 5		Y	
Near real-time ageing of fishing using AI Operationalize the integration of automated image analysis on fishing vessels for real or near real-time analysis, dramatically shortening ingest into Fisheries management timeline.	1	Y	Y	Y
AI + 'Omics: forecasting reef coral susceptibility to climate change Online "coral health map" to inform stakeholders, including FKNMS managers, which reefs within their jurisdiction are under greatest threat of climate change and which are more resilient to future stress events	1, 2, 5		Y	
Testing AI applications for weather forecasting in NOAA testbeds Leverage NOAA testbed infrastructure to put AI tools to the test with operational forecasters in a pseudo-operational environment. Aid training, and unique challenges of testing AI applications.	4?	Y	Y	Y

*WWCB Societal Challenges:

1. Changing Oceans

3. Effects of Space Weather

5. Monitoring and Modeling for Climate Change Mitigation

2. Coastal Resilience

4. Extreme Events & Cascading Hazards

6. Water Availability, Quality, and Risk