#### 3<sup>rd</sup> Workshop on Machine Learning for Earth Observation and Prediction

## **ECMWF** esa

- CHAIRS: Rochelle Schneider, Alan Geer, Alessandro Sebastianelli
- 16 in person, 74 virtual
- Organisations: Academia, Operational NWP/Forecasting, Industry, Space agencies



Working Group 2:

#### 3<sup>rd</sup> Workshop on Machine Learning for Earth Observation and Prediction

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- 1. CURRENT ML APPLICATIONS IN THE THEMATIC AREA:
  - a) Solar Energy/Radiation Forecasting
  - b) Wildfires Detection
  - c) Snow/Flooding Detection, Depth and Extent Estimation
  - d) Precipitation Retrievals
  - e) Environmental Health
  - f) Pollution
  - g) Oil spills
  - h) Oceanography/Chlorophyll detection
  - i) Crops classification
  - j) Urbanization
  - k) Geodesy
  - I) Clouds detection/motion modelling

- 1. CURRENT ML TOOLS AND APPROACHES IN THE THEMATIC AREA:
  - a) Segmentation
  - b) Classification/Detection
  - c) Forecasting
  - d) Regression
  - e) Domain Translation
  - f) Super Resolution/Downscaling

- 2. LIMITATIONS, CHALLENGES AND OPPORTUNITIES:
  - a) Reproducibility  $\rightarrow$  Scalable  $\rightarrow$  Operational  $\rightarrow$  Maintainable  $\rightarrow$  Transferrable  $\rightarrow$  Explainable
  - b) Continuous learning / retraining
  - c) Usability for society
  - d) Physics aware ML  $\rightarrow$  prior physical knowledge + constraints
  - e) Interoperability and integration with existing non ML tools
  - f) Lack of training data
    - a) citizen science + gamification (non-expert can label, e.g. zooniverse.org)
    - b) Domain expert required for labelling
    - c) Ground truth observation and generalization (e.g. from Europe to New Zealand)
  - g) supervised and semi-supervised
    - a) Few shot learning (zebra)
    - b) Meta-learning (learning how to learn)
    - c) Non domain foundation models (e.g. cats and dogs VS crop types)

Working Group 1: Machine Learning for Earth Observations

- 3. ADVANTAGES (DISADVANTAGES?) OF ML TECHNIQUES FROM TRADITIONAL STATISTICAL METHODS:
  - a) Fill the gaps around existing classical models
  - b) ML for diagnostics / evaluation / understanding outside of an existing framework
  - c) Using ML to do things that existing systems can't do
    - a) Replacing part of the process done by humans
    - b) Complex non linear correlations
    - c) Finding complex patterns in large datasets
  - d) NLP for mining/indexing unstructured data
  - e) Sector-specific transformative applications: forecasting future losses and gains in solar energy production  $\rightarrow$  reducing CO<sub>2</sub> emissions
  - f) Speed up in model performance and outcomes (training is slow, prediction is fast)

#### 4. FUTURE DIRECTIONS

- a) Federated learning
  - a) data protection issues (e.g. health data)
- b) AI on board satellites
  - a) Events detection (smart satellites)
  - b) Data compression / prioritization
  - c) Cooperating / connected satellites
  - d) Make sure we still keep the raw data for future learning
- c) Smart sensors  $\rightarrow$  connected environment  $\rightarrow$  smart observing systems / grids
- d) Emergent intelligence  $\rightarrow$  aggregation of more limited systems / agents
- e) Embodied intelligence  $\rightarrow$  proactive human-like learning
- f) Data fusion  $\rightarrow$  connecting diverse sources of data

- 5. EXTRA DISCUSSION CLOUD AI BASED PLATFORMS
  - a) More expensive than in house computing resources (re-train costs double)
  - b) On the other hand TPUs and GPUs speed a lot the processing
  - c) Multiple users interaction
  - d) Open online services (e.g. google earth engine) with private backends
  - e) Lock-in by the "open online services"
  - f) Rapid scale up
  - g) Lack of support
  - h) Data protection and security