#### **Project Introduction**

#### **Improving Precipitation Forecast Accuracy**

This internship centers on developing a robust weather prediction pipeline with an emphasis on precise rainfall forecasting. Over two weeks, the intern will design, implement, and evaluate a series of models and verification tools using historical weather data:

- Week 1: The focus is on data collection, preprocessing, baseline forecast development (logistic regression for rain/no-rain, ARIMA/linear models for accumulation), and the implementation of core verification metrics including Threat Score (TS), RMSE, MAE, Precipitation Accuracy Score (PAS), and Fractional Skill Score (FSS) <a href="vlab.noaa.gov">vlab.noaa.gov</a>.
- Week 2: Advancing into machine learning (Random Forest, XGBoost, SVR) and deep learning approaches (LSTM, CNN-LSTM), the intern will apply self-supervised bias correction, ensemble strategies for uncertainty quantification, and threshold-focused validation. The goal is to optimize both event detection and rainfall volume accuracy—especially for extreme events—assessed via metrics like Brier Score, ROC-AUC, reliability diagrams, and ensemble spread analysis.

By the end of the internship, the candidate will have delivered:

- A structured modeling pipeline (baseline  $\rightarrow$  ML  $\rightarrow$  DL  $\rightarrow$  ensemble)
- A comprehensive verification toolkit covering spatial, probabilistic, and volume-based metrics
- A final presentation showcasing performance comparisons and recommendations for enhancing operational forecast systems

## **Week 1 — Data & Modelling Foundations**

#### Day 1 – Data Acquisition & Exploration

- Objective: Collect clean historical weather data.
- Actions:
  - o Retrieve ~2 years of hourly data (rain amount, temperature, humidity, wind, pressure).
  - o Analyze data: missing values, distributions, seasonality, and quality.

#### Day 2 – Preprocessing & Feature Engineering

- Objective: Prepare meaningful model inputs.
- Actions:
  - o Impute missing data (linear interpolation, KNN).
  - o Engineer features: lagged rainfall (past 1h, 3h, 6h), cumulative rainfall, humidity trends, and calendar attributes.
  - o Normalize/standardize features for regression/ML models.

#### **Day 3 – Baseline Precipitation Forecasts**

- Objective: Set baseline for rain detection & amount.
- Actions:
  - Logistic regression for rain/no-rain and linear regression/ARIMA for rain amount.
  - Review evaluation metrics:
    - Threat Score (TS) for classification egusphere.copernicus.org+9wpc.ncep.noaa.gov+9wcrp-climate.org+9
    - RMSE/MAE for continuous forecasts

## **Day 4 – Precipitation Scoring Metrics**

• **Objective:** Expand metric suite for event and spatial accuracy.

#### • Actions:

- Compute Precipitation Accuracy Score (PAS) to capture both occurrence and volume <u>researchgate.net</u>
- o Implement **Fractional Skill Score (FSS)** for spatial consistency across neighborhoods wcrp-climate.org+15journals.ametsoc.org+15html.rhhz.net+15

## Day 5 - Advanced Verification Tools

• Objective: Build probabilistic and spatial diagnostic capacity.

#### Actions:

- o Build probabilistic classifiers or regression with thresholded events.
- o Score with Brier Score, Reliability Diagrams, ROC-AUC.
- Compute ESO metrics: CSI, FAR, Bias, and use FSS to handle neighborhood displacement issues <u>arxiv.orgscores.readthedocs.io+2html.rhhz.net+2researchgate.net+2wcrp-climate.org</u>

#### Day 6 - Machine Learning Modeling

• **Objective:** Improve predictions with data-driven methods.

#### • Actions:

- Train ML models (Random Forest, XGBoost, SVR) on engineered features to forecast amount and rain presence.
- o Evaluate using thresholds (e.g., ≥1 mm) with TS, PAS, FSS, MAE/RMSE.

## **Day 7 – Deep Learning Exploration**

• **Objective:** Leverage temporal/spatial patterns for better forecasts.

#### • Actions:

- o Prototype LSTM or 1D-CNN for sequential temporal features.
- o If spatial grids available, explore CNN-LSTM.
- Score models with PAS, FSS, continuous and probabilistic metrics.

#### Week 2 — Refinement, Ensembles & Reporting

## Day 8 – Self-Supervised Post-Processing (SSLPDL)

- **Objective:** Enhance predictive reliability via self-supervised learning.
- Actions:
  - o Fine-tune NWP-based forecasts using SSLPDL for rainfall probability estimation en.wikipedia.org+3learningweather.psu.edu+3arxiv.org+3wcrp-climate.org+1egusphere.copernicus.org+1arxiv.org+1journals.ametsoc.org+1
  - o Compare improvements in RMSE, PAS, Brier Score.

#### Day 9 – Ensemble Forecasting & Uncertainty

- Objective: Combine models for robustness.
- Actions:
  - o Generate ensembles (e.g., averaging across baseline, ML, DL models).
  - o Evaluate ensemble spread, calibration, and reliability diagrams.
  - o Use FSS to compare ensemble vs single-model spatial skill.

#### Day 10 – Spatial & Extreme Event Focus

- **Objective:** Fine-tune toward heavy rainfall accuracy.
- Actions:
  - Test models over thresholds (e.g.,  $\ge$ 10 mm).
  - Score with PAS across thresholds
    journals.ametsoc.org+4egusphere.copernicus.org+4wcrp climate.org+4wired.com+1sciencedirect.com+1researchgate.net+11researchgate.n
    et+11mausamjournal.imd.gov.in+11sciencedirect.com+5wcrp climate.org+5nature.com+5
  - Analyze FSS at varying neighborhood scales for performance localization arxiv.org+3html.rhhz.net+3arxiv.org+3

#### **Day 11 – Tooling & Efficiency**

• **Objective:** Ensure computational efficiency and proper use of metrics.

#### Actions:

- Optimize FSS with summed-area table methods researchgate.net+1learningweather.psu.edu+1
- o Ensure scalable pipelines and avoid "double-penalty" effects in evaluations.

#### **Day 12 – Final Model Selection & Validation**

- **Objective:** Consolidate insights and pick top-performing model/ensemble.
- Actions:
  - o Compare performance across models using RMSE, PAS, TS, FSS, Brier.
  - o Evaluate on unseen test data period (e.g., last 6 months).
  - o Identify best structured system for daily use.

## **Day 13 – Reporting & Presentation Prep**

- **Objective:** Summarize learnings and propose next steps.
- Actions:
  - o Create summary slides/report with:
    - Data sources and feature summaries
    - Evaluation metric definitions
    - Performance tables and charts
    - Reliability diagrams and FSS spatial maps
    - Precipitation threshold analysis
  - o Present roadmap:
    - Incorporate spatial grids, NWP outputs, transformer models
    - Apply SSLPDL for real-time traffic
    - Operational deployment suggestions

## Day 14 - Feedback, Documentation, & Handover

• **Objective:** Finalize deliverables and gather feedback.

## Actions:

- o Polish notebooks, codebase, and documentation (README, use guidelines).
- o Submit final code, presentations, and datasets.
- o Deliver final walk-through to mentor/team and seek improvement suggestions.

# **Metric Overview**

Metric	Туре	Purpose
Threat Score (TS)/CSI	Binary event	Hit rate for thresholded rainfall ( <a href="https://doi.org/html.rhhz.net">html.rhhz.net</a> , <a href="https://doi.org/ncep.noaa.gov">arxiv.org</a> , <a href="https://doi.org/ncep.noaa.gov">arxiv.org</a> , <a href="https://doi.org/ncep.noaa.gov">wired.com</a> , <a href="https://doi.org/ncep.noaa.gov">journals.ametsoc.org</a> )
RMSE / MAE	Continuous	Quantifies forecast error magnitude
Precipitation Accuracy Score (PAS)	Event + volume	Combines timing & volume accuracy
Fractional Skill Score (FSS)	Spatial/neighborhood	Mitigates location bias
Brier Score / ROC-AUC / Reliability	Probabilistic	Evaluates calibration of rain probability
Ensemble Stats	Spread & calibration	Quantifies uncertainty and forecast reach

# **Deliverables at Completion:**

- 1. Organized code and notebooks
- 2. Final model pipelines (baseline, ML, DL, ensemble)
- 3. Complete scoring scripts for all metrics
- 4. Results documentation (tables & graphics)
- 5. Final presentation and recommendation deck