

# CS3: Extreme Weather Image Classification - Student Rubric

**Goal:** Successfully construct and rigorously evaluate a Deep Learning model for binary classification of satellite imagery (Extreme vs. Normal Weather).

## Task Summary

Your work begins with the original 9,081 image dataset (Harvard Dataverse) and must proceed through the complete pipeline: data normalization, class consolidation, model development (baseline and advanced), full testing, and interpretable analysis.

## Assessment Criteria (What Constitutes Successful Completion)

Category	Criterion	Exceeds Expectations (100%)	Meets Expectations (80%)	Needs Work (50%)
<b>A. Data Preparation</b>	<b>Replication of Pipeline</b>	Correctly implements stratified 80/10/10 split AND applies all necessary pre-processing (224x224 resizing, deduplication).	Successfully executes the stratified 80/10/10 split on the consolidated 2-class dataset.	Class consolidation is incorrect or the splitting process shows significant error.
<b>B. Model Selection</b>	<b>Advanced Architecture Use</b>	Implements and trains <b>two</b> distinct models: a foundational CNN (baseline) AND a sophisticated Transfer Learning architecture.	Implements and trains <b>one</b> appropriate Transfer Learning model.	Only a basic, non-transfer learning model is attempted.

<b>C. Performance Metrics</b>	<b>Target F1-Score Attainment</b>	Achieves a Macro-Averaged F1-Score of <b>0.90 or greater</b> on the Test Set.	Achieves a Macro-Averaged F1-Score between <b>0.85 and 0.89</b> on the Test Set.	The final F1-Score is below the 0.85 threshold.
<b>D. Interpretation</b>	<b>Grad-CAM Visualization</b>	Generates clear Grad-CAM heatmaps AND includes thoughtful analysis of which meteorological features the model attends to.	Generates the required Grad-CAM visualizations and provides a sufficient description of the model's focus.	Model interpretability visualization is omitted entirely.
<b>E. Documentation</b>	<b>Quality &amp; Reproducibility</b>	All scripts are thoroughly commented, easily reproducible, and the final report offers a substantive discussion of potential bias.	Code executes reliably, and the report adequately documents the key steps and final metrics.	The submitted code is non-functional, difficult to follow, or lacks adequate documentation.

## Required Deliverables (Submitted via GitHub Repository)

- Code ([Scripts/](#) folder):** One Python script/Jupyter Notebook ([classification\\_pipeline.py](#)/[ipynb](#)) documenting the entire workflow.
- Results ([Reports/](#) folder):** Final Confusion Matrices and Classification Reports.
- Visualizations ([Visuals/](#) folder):** Key Grad-CAM heatmaps.
- Final Reflection (README, 300-500 words):** A summary addressing:
  - Justification for the chosen transfer learning model and its effectiveness.
  - Assessment of the most difficult analytical choice (Class Collapsing, Resizing, or Deduplication).
  - Comparison of final metrics (Accuracy, F1-Score) against original hypothesis targets.