

# DATA 606 CAPSTONE IN DATA SCIENCE

## Project Title: BlizzardX: Cracking the Code of Snow & Temperature Swings

### Team H:

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### Research Question:

How does snowfall and snow depth influence temperature fluctuations? Can machine learning models predict temperature changes and extreme cold events with high accuracy?

### Project Overview:

This project explores the relationship between **snowfall, snow depth, and temperature fluctuations** in a given area. The key objectives are:

- Understanding how snowfall influences maximum and minimum temperatures.
- Developing **predictive models for extreme cold events** using snowfall data.
- Enhancing insights into microclimate variations and providing useful forecasting tools.

### Dataset Information:

#### Dataset Source:

- **Data:** Directory with “.dly” files for all of GHCN-Daily  
[<https://www.ncei.noaa.gov/pub/data/ghcn/daily/all/>]
- **ghcnd-inventory.txt:** File listing the periods of record for each station and element  
[<https://www.ncei.noaa.gov/pub/data/ghcn/daily/ghcnd-inventory.txt>]
- **ghcnd-stations.txt:** List of stations and their metadata (e.g., coordinates)  
[<https://www.ncei.noaa.gov/pub/data/ghcn/daily/ghcnd-stations.txt>]

#### Dataset Description:

This dataset contains information about **snowfall, temperature fluctuations, and other relevant factors**.

- **Dataset Size:** Approximately **1,29,648 station records** across world.
  - **Attributes:**
    - **Snowfall (SNOW):** Daily snowfall amount.
    - **Snow Depth (SNWD):** Depth of snow on the ground.
    - **Maximum Temperature (TMAX):** Maximum daily temperature.
    - **Minimum Temperature (TMIN):** Minimum daily temperature.
    - **Observed Temperature (TOBS):** Actual temperature observed at the station.
    - **Station Identifier:** Unique identifier for each station.
    - **Year:** Year of the observation.
    - **Month:** Month of the observation.
- (Additional station metadata available in a separate CSV file):*
- **Station Name:** The name of the station.
  - **Coordinates (Latitude, Longitude):** Geographical location of the station.
  - **Elevation:** Elevation of the station above sea level.

## Planned Analysis & Methodology:

### Exploratory Data Analysis (EDA):

- **Correlation Analysis:** Examining the relationship between snowfall and temperature fluctuations (TMAX, TMIN).
- **Seasonal Trends:** Investigating how snowfall impacts temperature across different months.
- **Geospatial Analysis:** Analyzing the impact of station coordinates (latitude, longitude) and elevation on temperature and snowfall patterns.
- **Visualizations:** Creating scatter plots and line graphs to compare temperature fluctuations on snowy vs. non-snowy days.

### Machine Learning Models:

#### Regression Models (for Predicting Temperature Changes):

- **Gradient Boosting Regressor**
- **Long Short-Term Memory (LSTM)**

*(Other regression models may be explored as the project progresses.)*

#### Classification Models (for Predicting Extreme Cold Events, where temperature drops below a specific threshold):

- **Logistic Regression**
- **XGBoost**

*(Additional classification models may be considered in later phases.)*

These models will predict temperature variations based on snowfall and snow depth and classify days that are likely to experience extreme cold temperatures.

## Performance Evaluation:

### Regression Models ( Evaluation of Temperature prediction Models )

- **R<sup>2</sup> Score**
- **Root Mean Squared Error (RMSE)**

### Classification Models ( Predicting Cold Events )

- **Accuracy**
- **F1-Score**
- **Confusion Matrix**

## User Interface:

We are planning to develop an interactive web application to visualize the predicted temperature variations and extreme cold event forecasts, with real-time data fetching from NOAA.

## Conclusion:

This project aims to improve our understanding of how **snowfall influences temperature fluctuations** and develop **machine learning models to predict extreme cold events**.

The predictive insights gained from this study could be valuable for:

- **Weather forecasting systems.**
- **Emergency preparedness for severe winter conditions.**
- **Urban planning and infrastructure adaptation.**
- **Agricultural planning to mitigate frost damage.**

By leveraging **historical weather data, machine learning techniques, and geospatial analysis**, this project goes beyond traditional weather apps, providing **data-driven insights and predictive tools** that help forecast extreme cold events more accurately.