# **Main Question**

How is mountain snow affected by air temperature?

# **Data Source**

### Air Temperature Data

- · Data Source: Western juniper-dominated Experimental Catchments in south western Idaho, USA
- Metadata URL: https://data.gov/
- Data URL: https://ndownloader.figshare.com/files/44334659
- Content: Air temperature data of six places at south wester Idaho, USA in celsius, water\_year, month, day calendar\_year from 2007-2013.

|   | water_year | month | day | calendar_year | hour | sme2_ta_C | smf1_ta_C | smg1_ta_C | smg2_ta_C | smm1_ta_C | smm2_ta_C |
|---|------------|-------|-----|---------------|------|-----------|-----------|-----------|-----------|-----------|-----------|
| 1 | 2008       | 10    | 01  | 2007          | 00   | 9.0       | 9.4       | 9.7       | 9.2       | 8.8       | 8.7       |
| 2 | 2008       | 10    | 01  | 2007          | 01   | 8.1       | 8.7       | 9.0       | 8.4       | 8.6       | 8.1       |
| 3 | 2008       | 10    | 01  | 2007          | 02   | 7.2       | 8.4       | 8.8       | 8.0       | 8.4       | 7.6       |
| 4 | 2008       | 10    | 01  | 2007          | 03   | 6.6       | 7.8       | 8.3       | 7.5       | 7.8       | 7.1       |
| 5 | 2008       | 10    | 01  | 2007          | 04   | 6.2       | 7.7       | 7.9       | 7.1       | 7.5       | 6.7       |

### **Snow Depth Data**

- Data Source: Western juniper-dominated Experimental Catchments in south western Idaho, USA
- Metadata URL: https://data.gov/
- Data URL: https://ndownloader.figshare.com/files/44334722
- · Content: Snow depth data of six places at south wester Idaho, USA in millimeter, water\_year, month, day calendar\_year from 2007-2013.

|   | water_year | month | day | calendar_year | hour | sme2_sd_mm | smf1_sd_mm | smg1_sd_mm | smg2_sd_mm | smm1_sd_mm | smm2_sd_mm |
|---|------------|-------|-----|---------------|------|------------|------------|------------|------------|------------|------------|
| 1 | 2008       | 10    | 01  | 2007          | 00   | 0          | 0          | 0          | 0          | 0          | 0          |
| 2 | 2008       | 10    | 01  | 2007          | 01   | 0          | 0          | 0          | 0          | 0          | 0          |
| 3 | 2008       | 10    | 01  | 2007          | 02   | 0          | 0          | 0          | 0          | 0          | 0          |
| 4 | 2008       | 10    | 01  | 2007          | 03   | 0          | 0          | 0          | 0          | 0          | 0          |
| 5 | 2008       | 10    | 01  | 2007          | 04   | 0          | 0          | 0          | 0          | 0          | 0          |

### **Data Structure and Quality**

These datasets are collected from authentic sources. Which is reflect the real world data and accurate. All dataset information is complete and very few missing values.

### **Data Source Licenses**

All data are received from US government open sources and available for public use. The licence is us-pd and has no copyright restrictions.

# **Data Pipeline**

The data pipeline, includes the following stages:

### **Technology Used**

The data pipeline is implemented by Python using libraries like SQLite for data storage, Pandas for data manipulation andm os for interacting with the operating system.

#### **Transformation**

The transformation stage is critical in preparing the data for analysis. Key steps in this process include:

- 1. Date Table Creation: Two comprehensive date tables are generated to ensure consistency in date-related data across both datasets. This function, create\_date\_table, creates a DataFrame with a range of dates, including various components like year, month, day, and quarter. This standardization aids in accurate temporal analysis and comparison. This table is joined with the covid and crime datasets to perform analysis.
- 2. Reanaming and deleting columns: Removes the unnecessary columns from the datasets, like water\_year. and marge month, day, calendar\_year column into a single column name date. it also reanames some columns for efficient use during analysis.
- 3. Data Type Conversions: The script includes conversions of data types, ensuring that each column in the datasets is of the appropriate data type for analysis. For instance, converting strings to dates.

The above steps ensure that the data is clean, consistent, and ready for analysis. This detailed process is crucial for obtaining reliable and accurate insights from the data.

```
In [ ]: # URLs of the datasets
        air_temperature_url = 'https://ndownloader.figshare.com/files/44334659'
        snow_depth_url = 'https://ndownloader.figshare.com/files/44334722'
In [ ]: # SQLite database file path
        data_dir = './data'
        db_path = os.path.join(data_dir, 'climate_data.sqlite')
In [ ]: def run(self):
            # Ensure the data directory exists
            if not os.path.exists(os.path.dirname(self.db_path)):
            os.makedirs(os.path.dirname(self.db_path))
            # Connect to the SQLite database (it will be created if it doesn't exist)
            conn = sqlite3.connect(self.db_path)
            # Transform and save the air temperature dataset to SQLite
            self.save_to_sqlite(self.air_temperature_url, 'air_temperature', conn)
            # Transform and save the snow depth dataset to SQLite
            self.save_to_sqlite(self.snow_depth_url, 'snow_depth', conn)
            # Close the database connection
            conn.close()
In [ ]: # Load the dataset
        df = pd.read_csv(url)
In [ ]: # Remove the "water_year" column if it exists
        if 'water_year' in df.columns:
           df.drop(columns=['water_year'], inplace=True)
In [ ]: # Rename air temperature column name
        if 'sme2_ta_C' in df.columns:
            df.rename(columns={'sme2_ta_C': 'sme2 temperature in celsius'}, inplace=True)
        if 'smf1_ta_C' in df.columns:
            df.rename(columns={'smf1_ta_C': 'smf1 temperature in celsius'}, inplace=True)
        if 'smg1_ta_C' in df.columns:
            df.rename(columns={'smg1_ta_C': 'smg1 temperature in celsius'}, inplace=True)
        if 'smg2_ta_C' in df.columns:
            df.rename(columns={'smg2_ta_C': 'smg2 temperature in celsius'}, inplace=True)
        if 'smm1_ta_C' in df.columns:
            df.rename(columns={'smm1_ta_C': 'smm1 temperature in celsius'}, inplace=True)
        if 'smm2_ta_C' in df.columns:
            df.rename(columns={'smm2_ta_C': 'smm2 temperature in celsius'}, inplace=True)
```

### **Challenges Encountered**

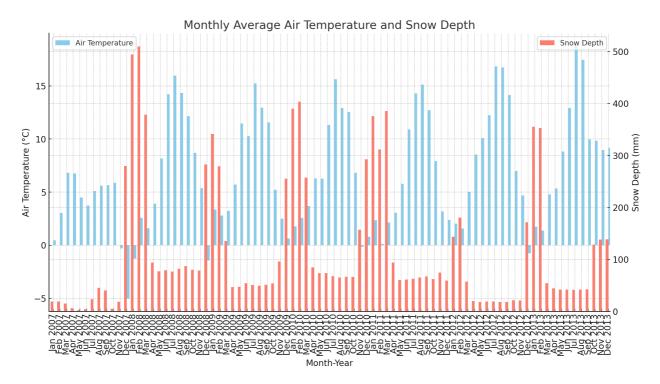
- 1. Incorrect data in the dataset, needed careful analysis to resolve the issue.
- 2. Handling missing data presented a significant challenge, requiring careful consideration to avoid wrong results.

3. Inconsistent data formats across sources, required standardization.

## Results

The data is presented in the form of plots and bargraphs. These representations aim to depict trends and patterns without imparting any interpretation. Key visualizations include:

- 1. Two tables in the SQLite database represent the output of the data pipeline during the time of November 2007 to September 2013 air temperature increases, snow depth tends to decrease.
- 2. This inverse relationship is consistent with climate science, where warmer temperatures lead to snowmelt and reduced snow accumulation.
- 3. Warmer temperatures cause more rapid snowmelt, exposing bare ground.
- 4. Rising air temperatures lead to shifts in ecosystems and landscapes. Glaciers recede, permafrost thaws, and plant zones migrate, altering the face of our planet.



# Limitations

- 1. Long tearm data required to get accurate result.
- 2. The study period may not be sufficient to establish long-term trends.