

Task 1.2: Ethics & Direction of Machine Learning Programs.

Part 1

- **Protecting Personal information:** When using machine learning to analyze climate change data, ClimateWins must be aware of potential biases and ethical concerns. While climate data does not typically include personal information, integrating geo-specific details could unintentionally expose information about certain locations or communities. ClimateWins need to ensure that data anonymization and limiting unnecessary geographic details can help mitigate this risk.
- **Regional and cultural biases:** The historical weather data comes from different stations across Europe, but it may not equally represent all regions. In the 1800s, data collection was often more advanced in areas with better technology, which could result in skewed predictions. To correct this, statistical adjustments or weighting techniques should be applied to make the dataset more balanced.
- **Human bias:** The way researchers collect, process, and interpret data can shape predictions, sometimes reinforcing existing assumptions rather than objective trends. Additionally, older measurement instruments were less precise, making historical data less reliable. Using Calibration techniques and careful data selection can help address these inconsistencies.
- **Preventing Misleading or Harmful Decisions:** If historical data is incomplete or inaccurate, ML models may make faulty predictions, potentially leading to harmful decisions. For example, areas prone to frequent natural disasters or extreme weather events might not be properly represented, resulting in a lack of preparedness for these risks. This could have serious consequences, including loss of life, if the models fail to adequately predict dangerous conditions.