
ClimAlytics AT

David Kalteis, Dominik Forsthuber, Michael Kerscher
BIG Data - Project

HAGENBERG | LINZ | STEYR | WELS



UNIVERSITY
OF APPLIED SCIENCES
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Dataset

- Provider: GeoSphere Austria
- Agency: Bundesanstalt für Geologie, Geophysik, Klimatologie und Meteorologie

GeoSphere Austria - Dataset API Frontend

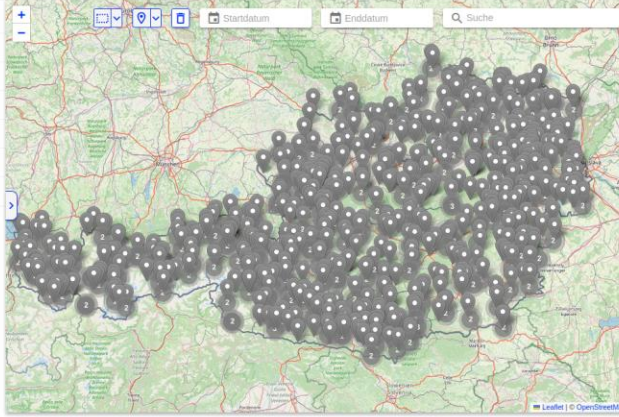
Messstationen Monatsdaten v2 1 Stationen wählen (0) 2 Parameter wählen (0) 3 Downloaden

Bitte wählen Sie hier Ihre Stationen mittels Tabelle oder Karte aus!

FILTER 1134/1134 KARTE

	<input type="checkbox"/>	Anzahl	ID ↑	Name	Startdatum
▼	<input type="checkbox"/>	1	1	Aflenz	Mai 1983
▼	<input type="checkbox"/>	2?	2	Aigen im Ennstal	März 1999
▼	<input type="checkbox"/>	2?	3	Allentsteig	Okt. 1983
▼	<input type="checkbox"/>	6?	4	Amstetten	Jan. 1936
▼	<input type="checkbox"/>	2?	5	Bad Aussee	Sept. 1983
▼	<input type="checkbox"/>	3?	6	Bad Bleiberg	Jan. 1996

Zeilen pro Seite: 6 ▼ 1 - 6 von 1134 | < > > |

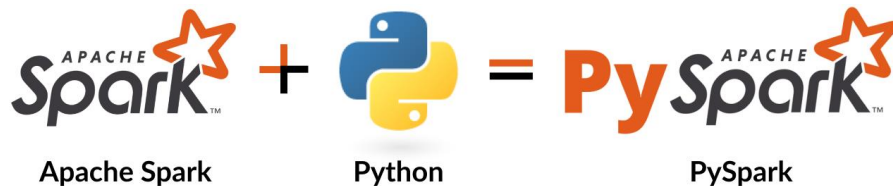


ABBRECHEN ZURÜCK WEITER

Technology Stack



Parquet



Spark SQL

Research Questions

- *How does the long-term trend in mean monthly temperature vary with elevation?*
- *Which geographic zones (valleys, plateaus, alpine corridors) show the largest shifts in “hot days” ($\geq 30\text{ }^{\circ}\text{C}$) and “frost days” ($\leq 0\text{ }^{\circ}\text{C}$) since 1970?*
- *How do station installation dates and validity periods create spatio-temporal gaps, and where are the largest “data deserts”?*
- *Which seasonal windows and locations optimize safety—combining sunshine hours, wind-gust flags, and frost/heat indicators?*

RQ1: Long-Term Temperature Trends

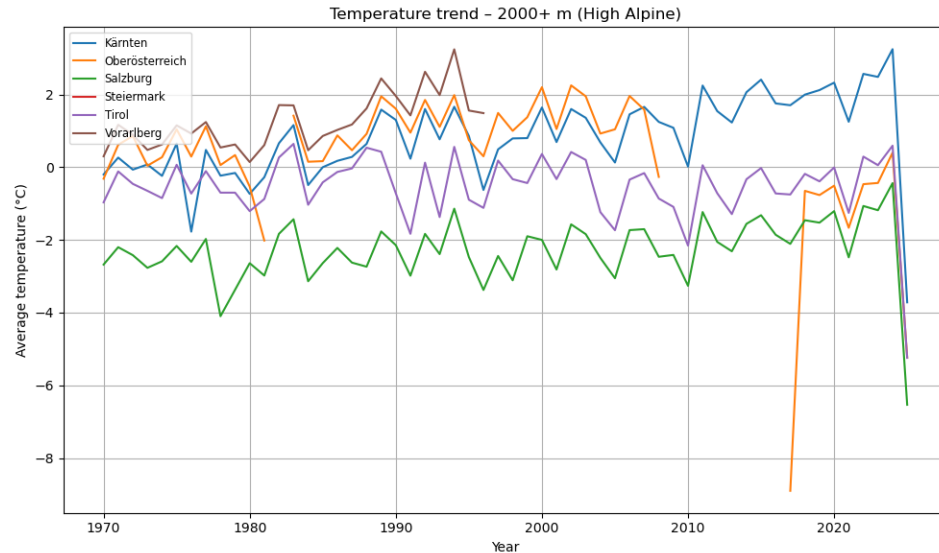
Objective:

- Long-term temperature trends differ by elevation

Methodology:

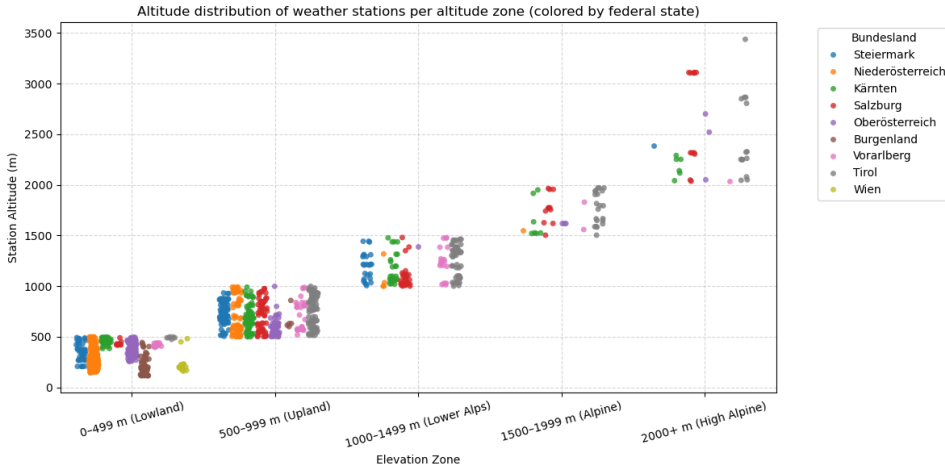
- Dataset of measurements with relevant columns: stations_id, date, tl_mittel
- Join stations metadata to classify into one of five elevation bands:
 - 0 – 499 m -> Lowland
 - 500 – 999 m -> Upland
 - 1000 – 1499 m -> Lower Alps
 - 1500 – 1999 m -> Alpine
 - 2000+ m -> High Alpine
- Aggregate by year, elevation zone and region

RQ1: Long-Term Temperature Trends



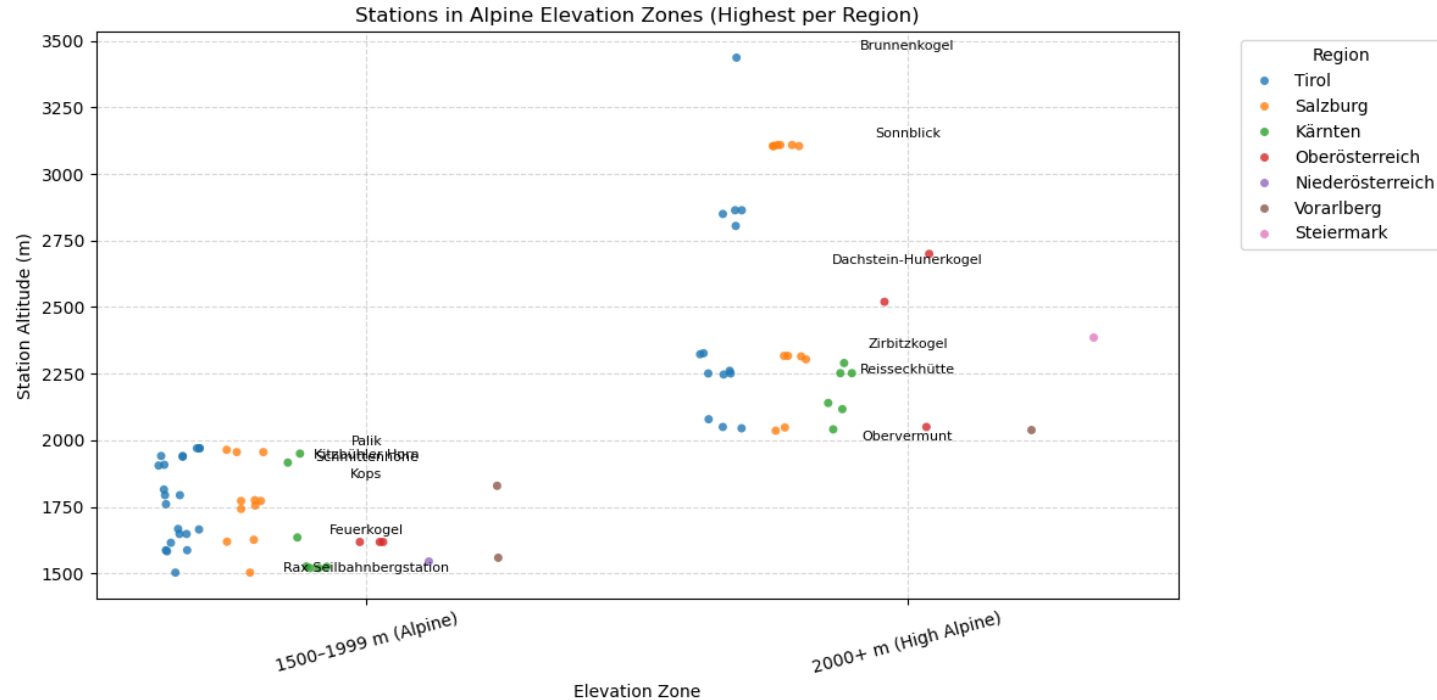
- Average temperature from 1970 – 2025
- All regions with High Alpine zone
- General increase
- Changes and missing values are visible
 - > Sensor malfunctions
 - > Missing or incomplete historical data
 - > Gaps due to station shutdowns

RQ1: Long-Term Temperature Trends



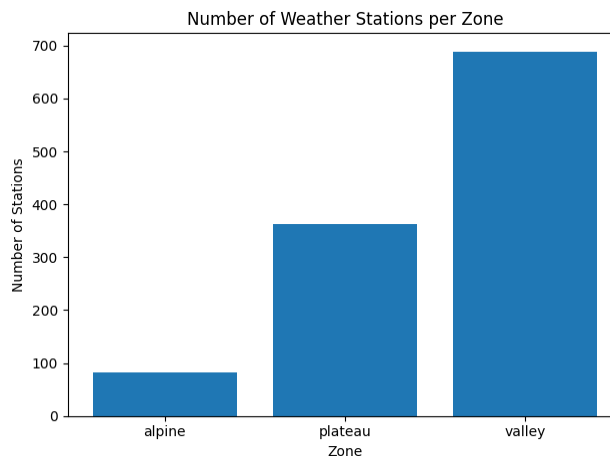
- Distribution of weather stations per elevation zone
- To validate
 - > the results of temperature trends
 - > Stations in High Alpine are located well above 2000+
- Regional diversity is given

RQ1: Long-Term Temperature Trends



RQ2: Spatial Patterns of Extremes

- Trends in “hot days” ($\geq 30\text{ }^{\circ}\text{C}$) and “frost days” ($\leq 0\text{ }^{\circ}\text{C}$) since 1970
 - > **tage_tropen**, *Tropentage*, "Zahl der Tropentage, Tagesmaximum der Lufttemperatur in 2 m Höhe $\geq 30.0\text{ }^{\circ}\text{C}$ ", d
 - > **tage_frost**, *Frosttage*, "Zahl der Frosttage, 24-Stunden-Minimalwert der Lufttemperatur in 2 m Höhe $< 0.0\text{ }^{\circ}\text{C}$ ", d
- Zones defined by elevation:
 - > **Valley**: $\leq 700\text{ m}$
 - > **Plateau**: 701–1500 m
 - > **Alpine**: $> 1500\text{ m}$



RQ2: Full Time Series Analysis

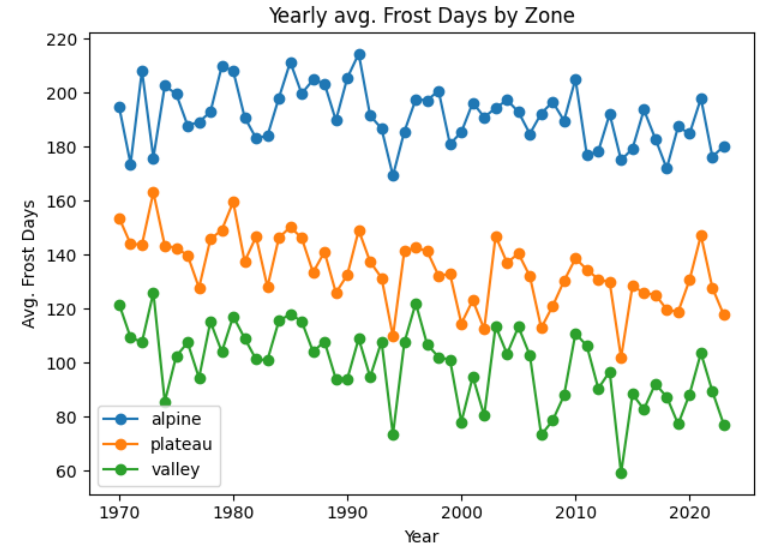
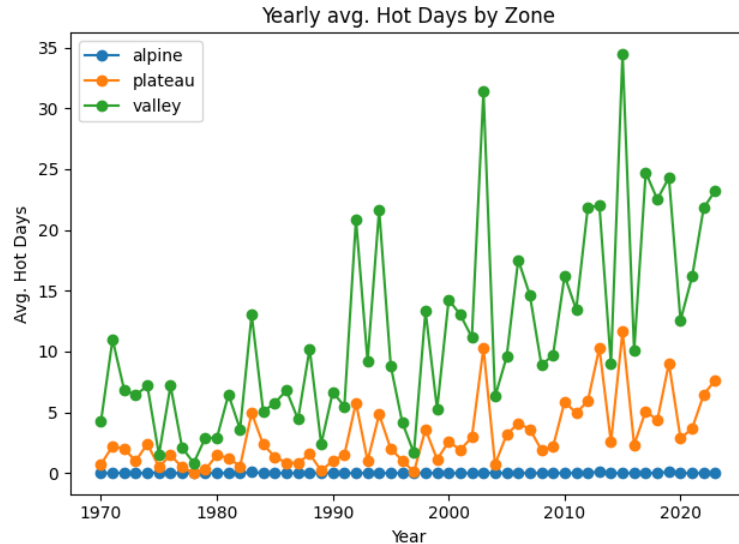
Data Processing

- Ingest raw CSV into Spark; parse date & extract year
- Join station metadata; attach elevation → assign zone
- Repartition by year & zone, write to Parquet for fast queries

Visualization

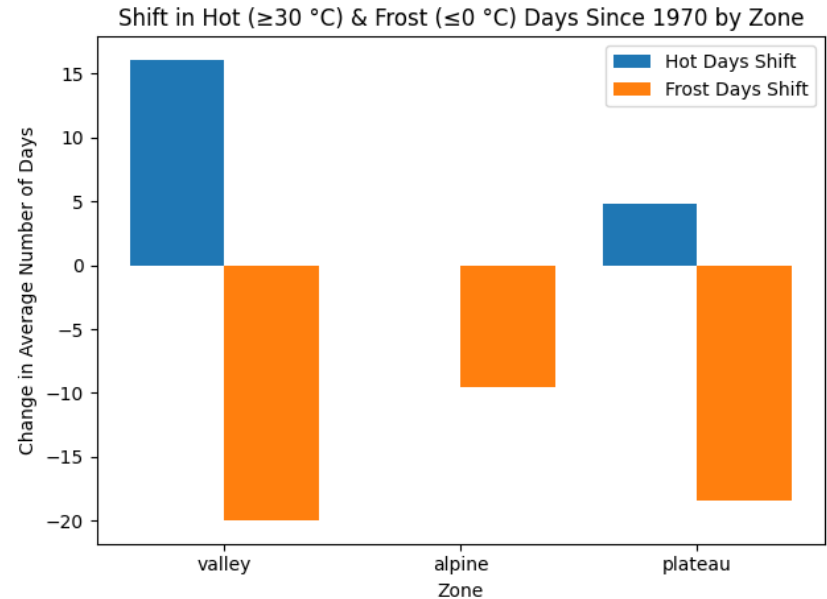
- Compute annual, per-station averages of hot & frost days
- Plot time series (1970–2023) for each zone

RQ2: Full Time Series Analysis



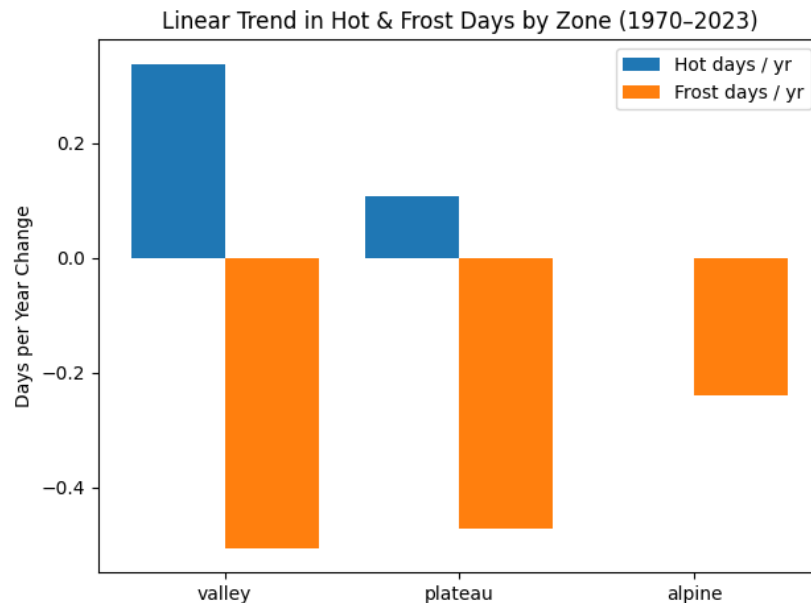
RQ2: End–Minus–Start Difference

- Calculate $\text{mean}(2014\text{--}23) - \text{mean}(1970\text{--}79)$ by zone
- Highlights net shifts over 50 years
- Valleys: largest \uparrow hot days & \downarrow frost days; plateaus moderate; alpine minimal



RQ2: Linear Regression Trends

- Fit a straight trend line to each zone's annual hot-day and frost-day averages
- Slope = average change in days per year (\uparrow or \downarrow)
- Incorporates every year's data, making the trend resistant to outlier years



RQ3: Spatio-Temporal Data Coverage Pro.

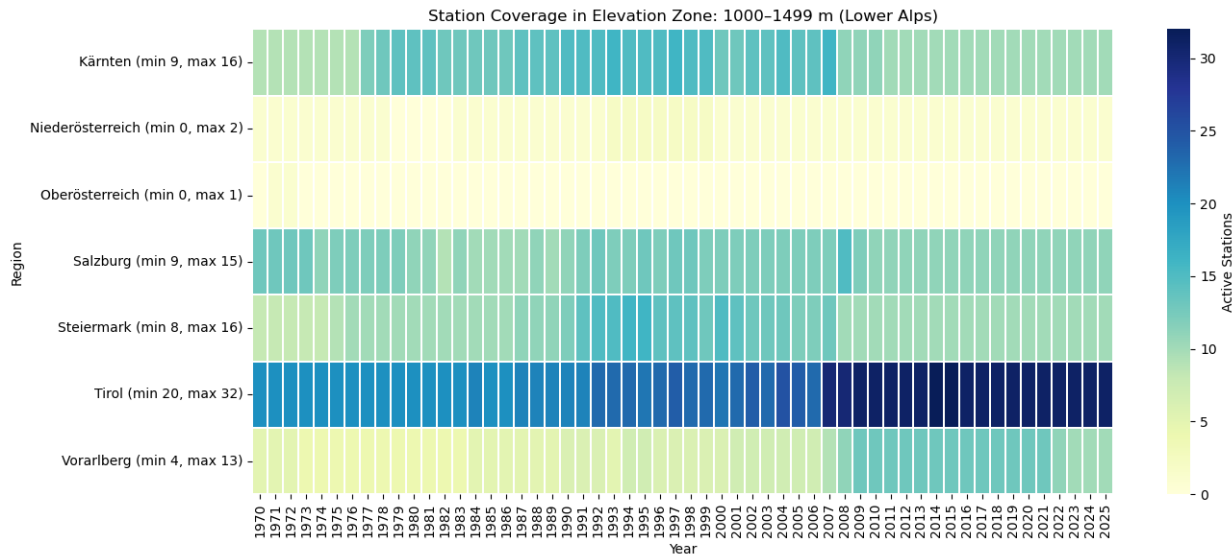
Objective:

- Identify spatial and temporal coverage gaps

Methodolgy:

- Two main steps:
 1. Metadata-Based Coverage
 2. Real Measurement-Based Coverage
 - To verify acutal data presence

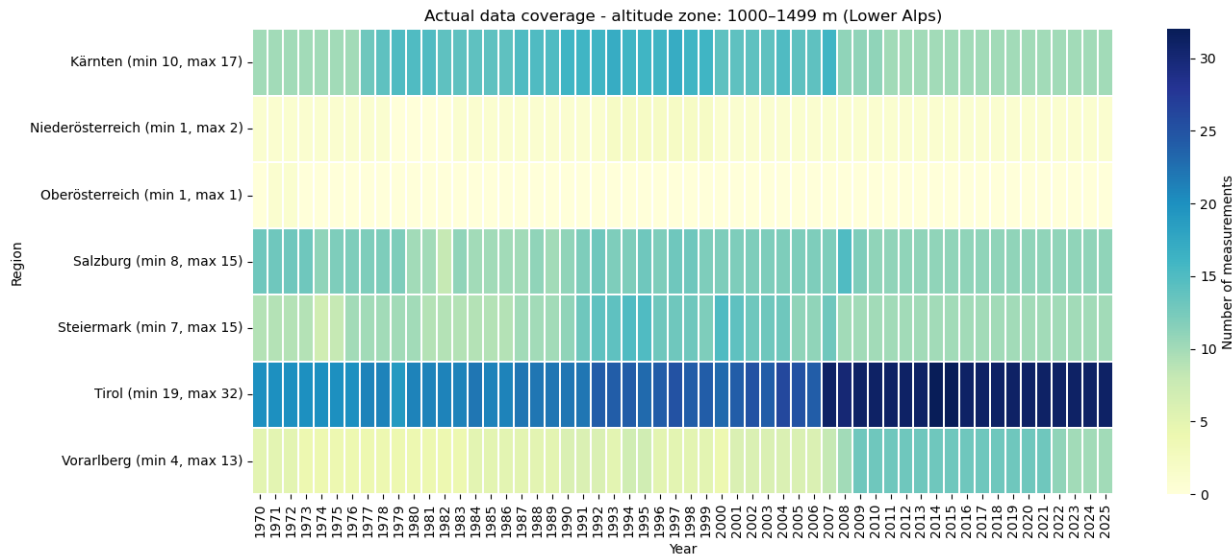
RQ3: Spatio-Temporal Data Coverage Pro.



Metadata-Based Coverage

- Using station metadata
- Filter by operational dates -> start / end
- Classify into elevation zones by altitude
- Count active stations per year x elevation zone x Bundesland

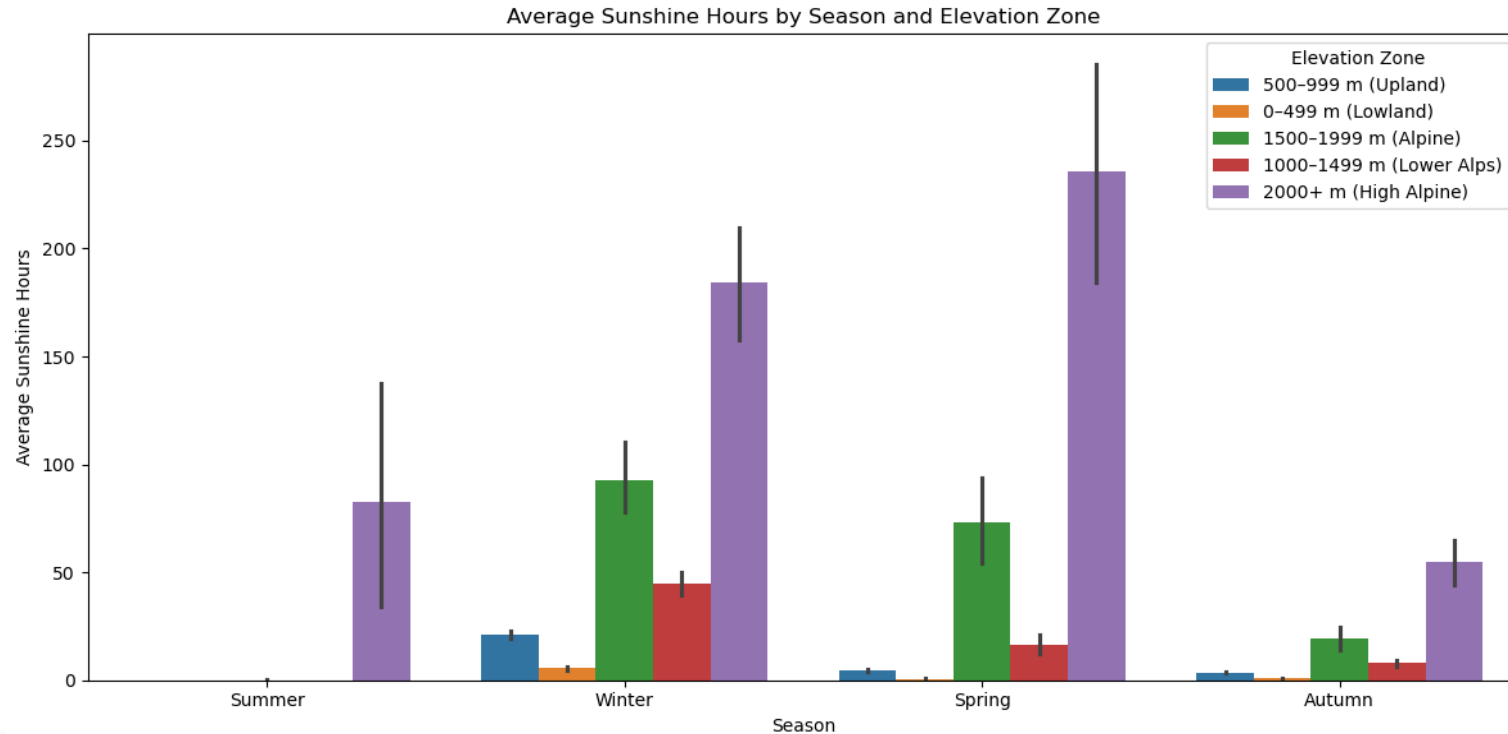
RQ3: Spatio-Temporal Data Coverage Pro.



Real Measurement-Based Coverage

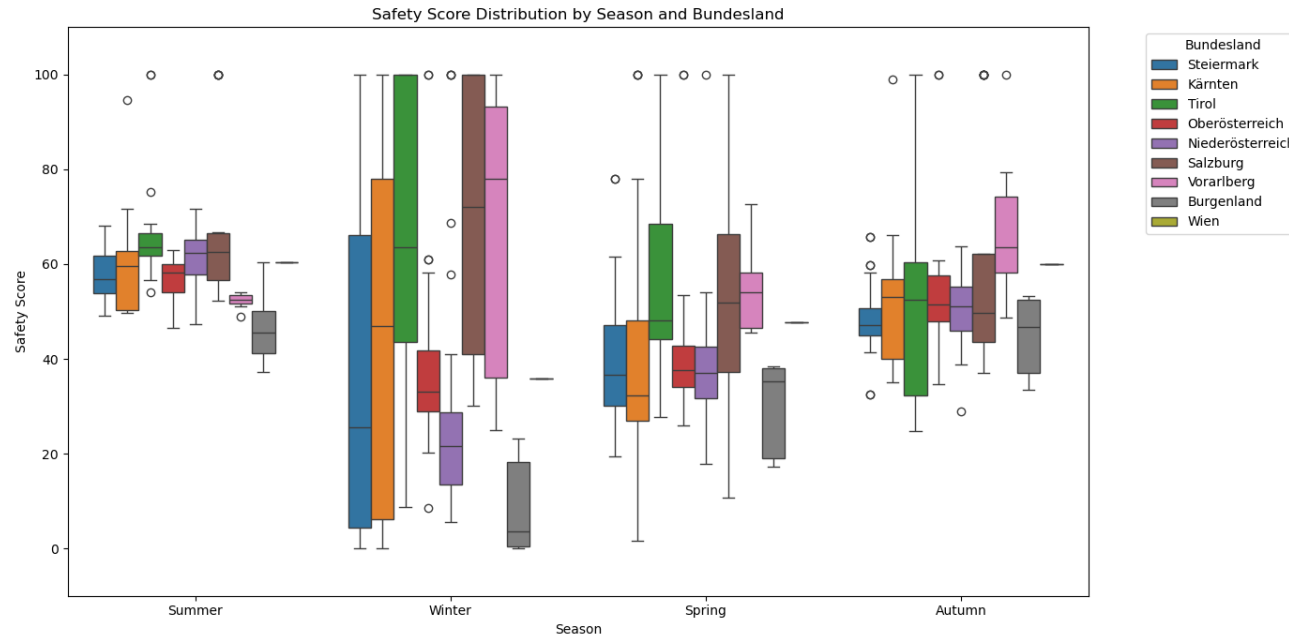
- Filter climate dataset for rows with non-null measurements
- Join with metadata and classify by elevation zone and Bundesland
- Count distinct active stations per year x zone x Bundesland

RQ4: Operational Window Optimization.

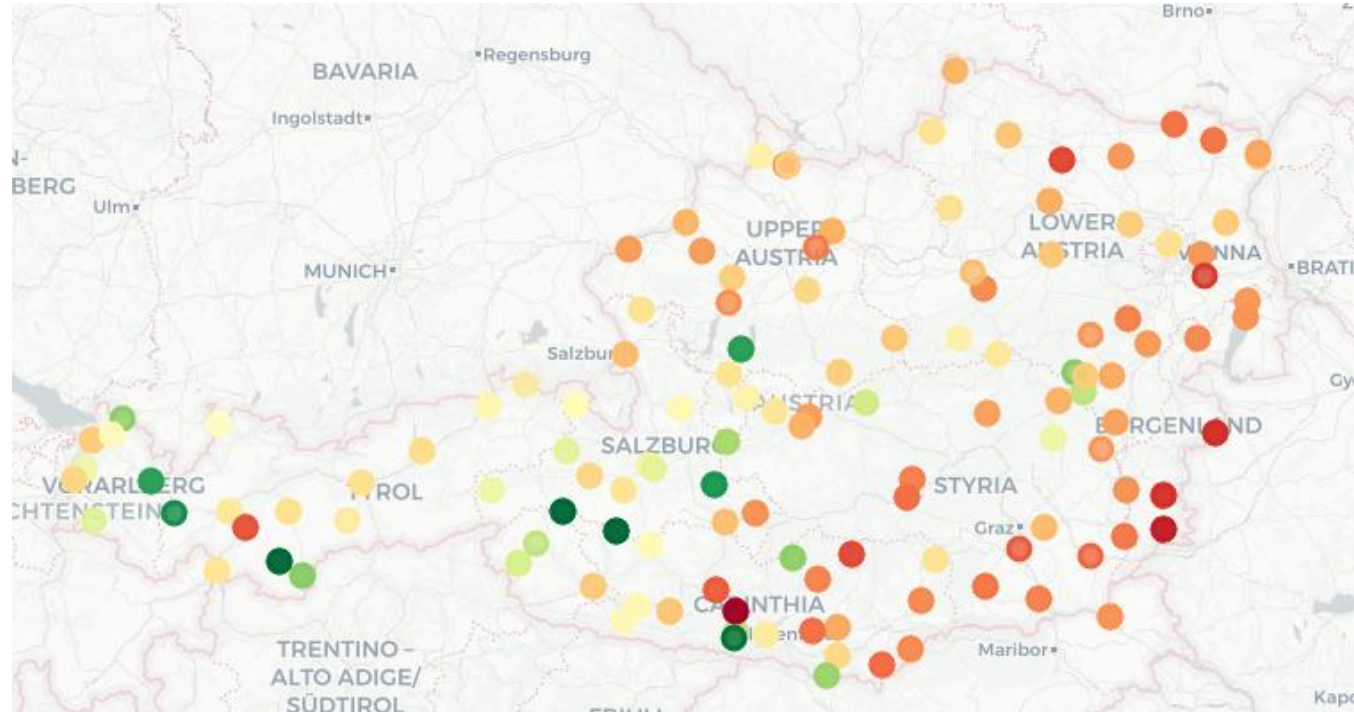


RQ4: Operational Window Optimization.

— Score = Sunshine - Wind Gust Freq - Frost Freq - Heat Freq



RQ4: Operational Window Optimization.



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