

Shows a quick summary of the cleaned dataset, helping to check column types, missing values, and overall structure before analysis. # ------ 2) EDA ----print("\n=== Info ===") display(weather_df.info()) O === Info === cclass 'pandas.core.frame.DataFrame'>
RangeIndex: 370 entries, 0 to 369
Data columns (total 13 columns): # Column Non-Null Count Dtype 370 non-null int64 weather id 370 non-null 370 non-null station_id int64 date tmax_f datetime64[ns] 370 non-null float64 370 non-null 370 non-null float64 float64 tmin_f wspd_mph precipitation_in 370 non-null name 370 non-null latitude 370 non-null float64 object float64 longitude elevation 370 non-null 370 non-null 370 non-null float64 11 year 370 non-null int32
12 month 370 non-null int32
dtypes: datetime64[ns](1), float64(7), int32(2), int64(2), object(1) memory usage: 34.8+ KB Summarizes the main weather variables numerically, giving a quick snapshot of central tendency, spread, and ranges. print("\n=== Summary statistics ===") display(weather_df[["tmax_f","tmin_f","wspd_mph","precipitation_in"]].describe()) === Summary statistics === **∓*** tmax_f tmin_f wspd_mph precipitation_in ## 370.000000 count 370.000000 370.000000 370.000000 mean 19.963243 18.263784 5.148108 0.034459 51.269059 52.445734 21.233156 0.560701 -0.950000 min -69.000000 -88.000000 -30.000000 -0.450000 **25%** -25.000000 -25.000000 -13.000000 50% 22.000000 18.500000 5.000000 0.000000 **75%** 65.000000 62.000000 23.000000 0.550000 0.950000 max 109.000000 120.000000 40.000000 Checks data completeness by showing the number of missing values per column. print("\n=== Missing values ===") display(weather_df.isna().sum()) === Missing values === **∓*** weather_id 0 station_id 0 date 0 tmax_f tmin_f 0 0 wspd_mph precipitation_in 0 0 latitude 0 Ionaitude 0

0

0

elevation year

```
year 0
month 0
dtype: int64
```

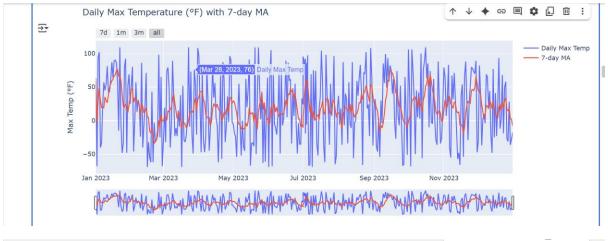
Shows how many daily records exist for each year, helping to spot gaps or uneven coverage in the dataset.

Turns threshold rules into flags, adds them to the dataset, and then shows how often each extreme condition occurred.

```
# -------3) APPLY THRESHOLDS -------
op_map = {">":np.greater,"<":np.less,">=":np.greater_equal,"<=":np.less_equal,"==":np.equal}
extreme_cols = []
for _, row in threshold_df.iterrows():
    var, op, val, name = row["variable"], row["operator"], row["value"], row["name"]
    if var in weather_df.columns and op in op_map:
        col = f"is_{name.lower().replace(' ','_)}"
        weather_df[col] = op_map[op](weather_df[var], val)
        extreme_cols.append(col)
    weather_df["is_extreme"] = weather_df[extreme_cols].any(axis=1)
    print("\n== Threshold flags summary ==")
    display(weather_df[extreme_cols + ["is_extreme"]].sum())</pre>
```

```
=== Threshold flags summary ===
θ
is_hot_day 28
is_cold_day 211
is_windy_day 105
is_extreme 265
dtype: int64
```

Creates an interactive line chart of daily maximum temperature with a 7-day moving average, including zoom/range controls and a date slider for easier exploration.



Creates an interactive bar chart where the user can toggle between seeing the number of extreme weather days per year or per month, using on-chart buttons.





Computes and plots the number of extreme days in each 7-day window, showing short-term clusters of extreme weather with an interactive time slider.



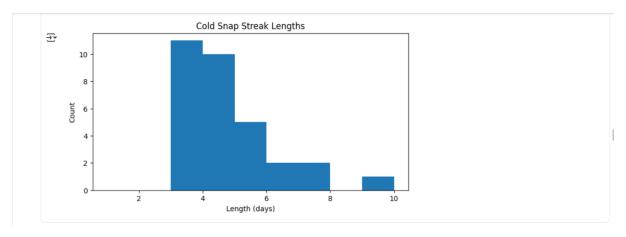
Groups consecutive flagged days into streaks by splitting whenever dates aren't exactly 1 day apart, then keeps only streaks at least STREAK_MIN_LENGTH days long (default 3).

Previews the detected cold snap streaks (minimum 3 days long), just like the earlier preview for heatwave streaks.

```
print("\n=== Cold Snap Streaks (23 days) ===")
display(coldsnap_streaks.head(10))
```

```
=== Cold Snap Streaks (≥3 days) ===
∓*
       start_date end_date length_days 🔠
     4 2023-01-18 2023-01-21 4
    5 2023-01-24 2023-01-26
    8 2023-02-04 2023-02-06
                                    3
    11 2023-02-19 2023-02-23
                                     5
    13 2023-03-01 2023-03-04
    19 2023-03-22 2023-03-25
    23 2023-04-06 2023-04-08
                                    3
    27 2023-04-20 2023-04-24
    28 2023-04-26 2023-04-29
    30 2023-05-05 2023-05-11
```

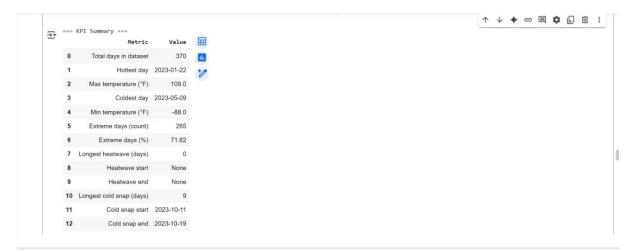
Visualizes the distribution of streak lengths for heatwaves and cold snaps, so you can see whether most streaks are short or if long streaks are common.

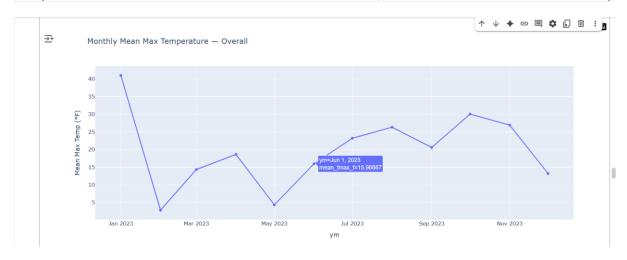


Computes headline metrics (coverage, hottest/coldest days, extreme-day stats) and extracts the longest heatwave/cold-snap streak, then displays everything in a tidy table.

```
"Min temperature (°F)": float(cold_row["tmin_f"]) if cold_row is not None else None,
    "Extreme days (count)": int(weather_df["is_extreme"].sum()),
    "Extreme days (%)": round(100*weather_df["is_extreme"].sum()),
    "Longest heatwave (days)": longest_streak_info(heatwave_streaks,"heatwave")["longest_heatwave_length"],
    "Heatwave start": longest_streak_info(heatwave_streaks,"heatwave")["longest_heatwave_stret"],
    "Heatwave end": longest_streak_info(heatwave_streaks,"heatwave")["longest_heatwave_end"],
    "Longest cold snap (days)": longest_streak_info(coldsnap_streaks,"coldsnap)["longest_coldsnap_length"],
    "Cold snap start": longest_streak_info(coldsnap_streaks,"coldsnap")["longest_coldsnap_start"],
    "Cold snap end": longest_streak_info(coldsnap_streaks,"coldsnap")["longest_coldsnap_end"],
}

print("\n=== KPI Summary ===")
    kpi_table = pd.DataFrame(list(kpi.items()), columns=["Metric", "Value"])
    display(kpi_table)
```







```
[17]

✓ Os
      # ----- MONTHLY DISTRIBUTION & SHARES-----
            import calendar
             # Prep: month label (Jan-Dec) in correct order
             monthly_day = weather_df.dropna(subset=["date","tmax_f"]).copy()
            monthly_day("month_num") = monthly_day("date").dt.month
monthly_day("month_lbl") = monthly_day("month_num").apply(lambda m: calendar.month_abbr[m])
             month_order = list(calendar.month_abbr)[1:] # ["Jan",..., "Dec"]
             # If station metadata exists, keep it; else create a placeholder for "All"
            has_station = ("station_id" in monthly_day.columns) and ("name" in monthly_day.columns)
             if not has_station:
                monthly_day["station_id"] = "ALL"
monthly_day["name"] = "All stations"
            # ----- BOX PLOT: Daily Max Temp distribution by month (dropdown: All vs. per-station) ------ # Build one set of 12 box traces per option (All + each station)
            # Note: if you have MANY stations, this can create many traces; consider filtering top-N.
      # Build "All stations" frame
            all_df = monthly_day.copy()
            \mbox{\tt\#} If you *do* have station_id, the "All" view aggregates across stations:
            all_df["view_key"] = "All stations"
            # Build per-station frames
            station_map = (monthly_day.dropna(subset=["station_id"])
                             .drop_duplicates("station_id")
.set_index("station_id")["name"].to_dict())
            station_ids = list(station_map.keys()) if has_station else []
            box_traces = []
trace_groups = [] # which view each trace belongs to
            views = ["All stations"] + [station_map[sid] for sid in station_ids]
             def add_box_traces(src_df, view_label):
                 # add 12 boxes (one per month)
                 for m in month_order:
                     yvals = src_df.loc[src_df["month_lbl"]==m, "tmax_f"]
                      box_traces.append(go.Box(
                          x=[m]*len(yvals), y=yvals,
```

```
hover template = "\b>\fix < \b>\color=Daily Max Temp: \fi = \fi 
0
                                                                      showlegend=False
                                                     trace groups.append(view label)
                      # All stations boxes
                      add_box_traces(all_df, "All stations")
                       # Per-station boxes
                      for sid in station_ids:
                                    df_s = monthly_day.loc[monthly_day["station_id"]==sid].copy()
                                       add_box_traces(df_s, station_map[sid])
                      # Visibility: show only first view initially
                     visible = [g == "All stations" for g in trace_groups]
                      fig_box = go.Figure(data=box_traces)
                      # Dropdown buttons toggle which group's 12 traces are visible
                      buttons = []
                      for view in views:
                                      vis = [g == view for g in trace_groups]
                                    buttons.append(dict(
                                                     label=view,
```

name=m, boxpoints="outliers", jitter=0.3, whiskerwidth=0.5,

