

Approaches to Quality Control and Assurance in a Meteorological Station Network John H. Porter, Nina Groleger & Jonah Morreale

During the summer of 2021 we have been going back over 30+ years of data from a small network of meteorological stations to resolve existing quality issues caused by partial or complete sensor failures. Once detected, issues are input into a database that then uses code-generation to flag, remove or comment on issues. Here we present some of the automated or semiautomated methodologies we have found to be most effective, as well as future plans for gap filling.

Issues:

- Sensor Failure (complete)
 - Easy to detect no data recorded
- Sensor Failure (intermittent)
 - Harder to detect
- Anomalous Sensor Readings
 - Consistent "stuck" sensors
 - Intermittent Bad values
 - Longterm Drift



Single Station

Range & Spike

Tests

Multi-parameter

Anomaly

Detection

Climatological

Outliers

- For parameters that are similar over a
 broad spatial range (e.g., Temperature,
 Pressure, Humidity), inter-station
 comparisons are especially helpful at
 identifying issues
- For parameters that vary locally (e.g., summer rainfall) temporal consistency may be more useful (e.g., extended periods of no rain, or apparent rainfall every day)

Automated Detection is accomplished using R code to estimate periods of sensor failure. Depending on the severity of the apparent failure, data are flagged, or in extreme cases, set to missing.

Instructions for data processing are recorded into a database (SensorProb) and code-generation services are used to transform level 0 (raw) data into level 1 (processed) data.

For example, automated code identified over 1,600 periods out of 30+ years of data at 3 stations where air temperature represented an anomaly when compared with other stations at the same time. A web service was used to "flag" as questionable the identified data in the SensorProb database.

To aid in eventual production of gap-filled data we are developing interstation regression models so that one station can be used to predict values at another, and daily and hourly annual climatologies for each station based on 30+ years of data

Regression based on data from nearby stations Gap-Filled Data Data Ginatologies

This work was supported by NSF Grant 1832221 and a Fellowship to N. Goleger from the Environmental Data Initiative (NSF Grants 1931143 and #1931174)

