



# Assuring the quality of observations

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# Outline

- Problem definition
- Automatic quality assurance system
- Overview of Belgingur QA system
- A case study





# Overview

- Measurement devices can cause multiple types of errors
- Erroneous observations can lead to wrong assessments
- Important to identify malfunctioning gauges or periods of observations that can't be trusted





# Sources of problem

- A measurement device stopped working, battery is off, memory was full
- Improperly calibrated devices
- Sensor was stuck for some reason (icing...)
- Lightning activity





# Spotting the problem

- Are the values within their assumed range?
- Do the observations vary with time? Don't they vary too much?
- Is the data series smooth or does it have spikes? Are there any single missing values from the time series?
- Are the observations from neighbouring stations similar?
- Manual inspection





# Automatic Quality Assurance

- Shafer, M. A., C. A. Fiebrich and D. S. Arndt (1999). Quality assurance procedures in the Oklahoma network. J. Atm. Ocean. Tech., 17(4):474–494.
- Perform **four quality tests** on **daily portions** of data and assign one of **six quality indicators** for each test





# Quality checks

- Range check
- Step check
- Persistence check
- Spatial check





# QA: range check

- Compare each observation datum, separately for each variable, to typical sensor performance and the expected climatological extremes
- Do all the variables fall within their ranges?
- Detect flaws such as oversaturation, incorrectly calibrated instruments





# QA: step check

- Is the timeseries continuous?
- Aren't the jumps between consecutive observations too large?





# QA: persistence check

- Works on data from a whole day at a time
- Detect too small changes in the standard deviation of the observed variable
- Detect too short span of the observed variable within the day
- Marks the whole day for given variable according to the result





# QA: spatial check

- How does the result for this station correspond with neighbourhood ones?
- At least six stations must be analyzed concurrently





# QA flags

0 – pass

1 – suspect

2 – warning

3 – failed

8 – not tested

9 – missing





# QA results

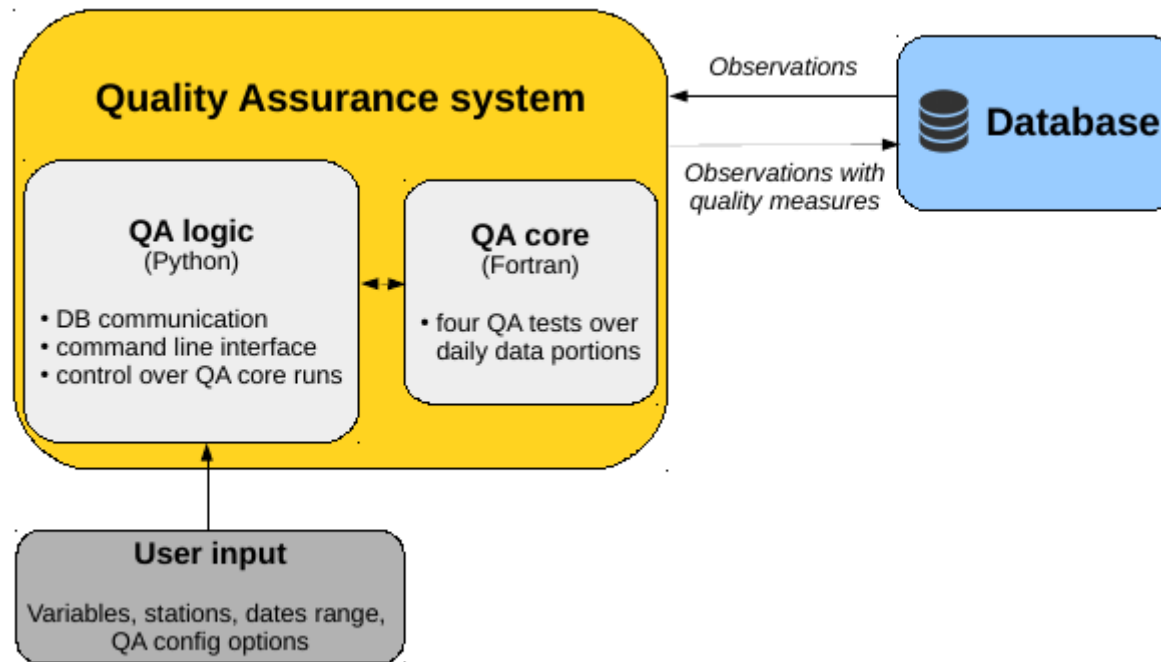
- 0 – pass, 1 – suspect, 2 – warning, 3 – failed, 8 – not tested, 9 – missing

time	station_id	variable	value	q_range	q_step	q_persistence	q_spatial
2010-07-02 11:00:00+00	908	rel_hum	106	3	8	8	8
2010-07-02 11:00:00+00	908	temp	3	0	0	0	0
2010-07-02 11:00:00+00	908	wind_dir	132	0	0	0	0
2010-07-02 11:00:00+00	908	wind_speed	3.2	0	0	0	0
2010-07-02 12:00:00+00	908	rel_hum	106	3	8	8	8
2010-07-02 12:00:00+00	908	temp	3.3	0	0	0	0
2010-07-02 12:00:00+00	908	wind_dir	183	0	0	0	0
2010-07-02 12:00:00+00	908	wind_speed	1.7	0	0	0	0
2010-07-02 13:00:00+00	908	rel_hum	103	0	8	0	0
2010-07-02 13:00:00+00	908	temp	3.4	0	0	0	0
2010-07-02 13:00:00+00	908	wind_dir	156	0	0	0	0
2010-07-02 13:00:00+00	908	wind_speed	3	0	0	0	0





# The QA system of Belgingur



- ...first, populate the database with data





# Defining metadata

- Create a setup file describing your stations and use our metadata importer to get it to a database

```
provider: {name: Vedurstofan, ref: vi.is}
stations:
- {active: true, has: 52.0, lat: 64.1275, lon: -21.9028, manual: true, name: Reykjavík, ref: vi.is.rvk, ref_provider:
- {active: false, has: 20.0, lat: 64.417, lon: -21.833, manual: true, name: Leirá, ref: vi.is.lra, ref_provider: 96,
- {active: true, has: 14.0, lat: 64.6419, lon: -21.590799999999998, manual: true, name: Stafholtsey, ref: vi.is.stey,
- {active: false, has: 46.0, lat: 64.667, lon: -21.617, manual: true, name: Hamraendar í Stafholtstungum, ref: vi.is.l
- {active: false, has: 45.0, lat: 64.817000000000001, lon: -22.267000000000003, manual: true, name: Haukatunga, ref: v
- {active: false, has: 80.0, lat: 64.800000000000001, lon: -23.033, manual: true, name: Neðri-Hóll, ref: vi.is.ndrh, r
- {active: false, has: 12.0, lat: 64.817000000000001, lon: -23.133, manual: true, name: Garðar, ref: vi.is.grdr, ref_p
- {active: true, has: 13.0, lat: 64.8393, lon: -23.3003, manual: true, name: Bláfeldur, ref: vi.is.blfl, ref_provider
- {active: false, has: 70.0, lat: 64.9, lon: -23.933, manual: true, name: Gufuskálar, ref: vi.is.gfsk, ref_provider:
- {active: false, has: 12.0, lat: 64.917, lon: -23.883000000000003, manual: true, name: Hellissandur, ref: vi.is.hsd,
```





# Importing observations to a database

- Transform your data to this format...

```
#station_ref: vi.is
#provider_ref: 96
time, temp, wind_speed, rel_hum
2000-01-01T00:00:00, 11.1, 11, 91
2000-01-01T00:10:00, 11.2, 12, 92
2000-01-01T00:20:00, 11.3, 13, 93
2000-01-01T00:30:00, 11.4, 14, 94
2000-01-01T00:40:00, 11.5, 15, 95
```

- ...or write your own customized parser providing your data to our system

```
class ExampleObservationParser(SeparatedTextObservationParser):

    def __init__(self):
        super(ExampleObservationParser, self).__init__(separator=',', headers=['^,]*[a-zA-Z]+[^,]*(*(. *[a-zA-Z]+.)*)*$')
        self.meta['provider_ref'] = 'met.sc'

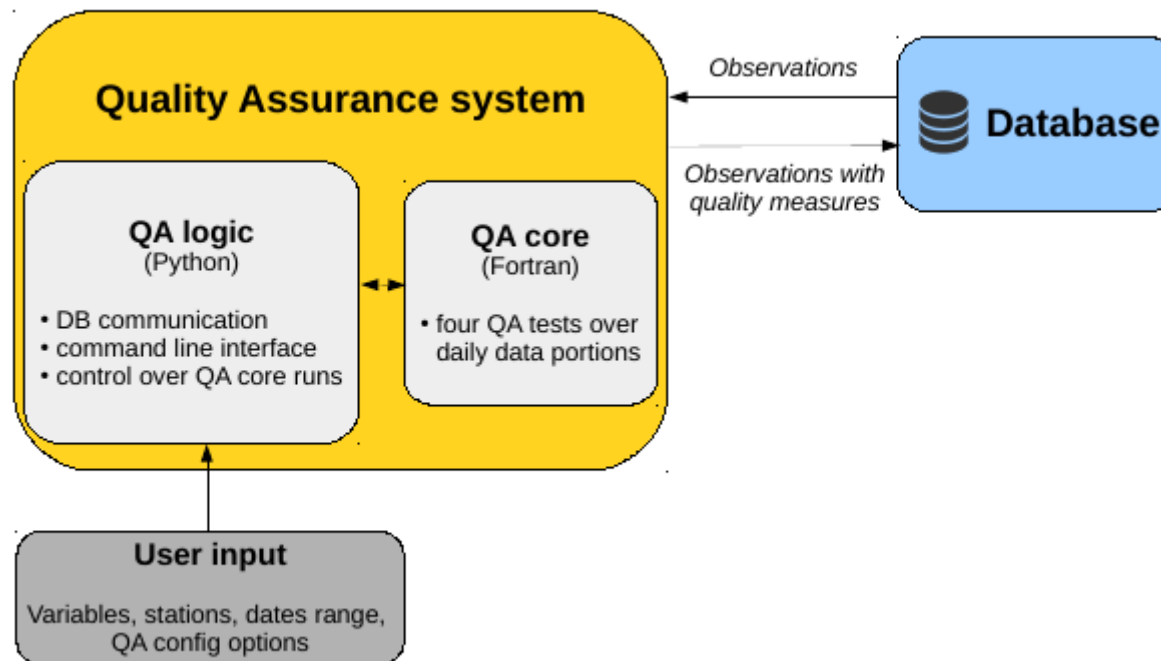
    def parse_header_line(self, line):
        if not super(ExampleObservationParser, self).parse_header_line(line):
            return False
        if line != '':
            parts = self.separator.split(line.strip())
            self.field_list = [('time', parse_time, parts.index('DD/MM/YYYY HH:MM')), ('station_ref', str, parts.index('Site'))]
            for key_in, key_out in FIELD_ALIASES.iteritems():
                if key_in in parts:
                    self.field_list.append((key_out, float_or_no_data, parts.index(key_in)))
        return True

    def parse_data_line(self, line):
        return super(ExampleObservationParser, self).parse_data_line(line.strip())
```



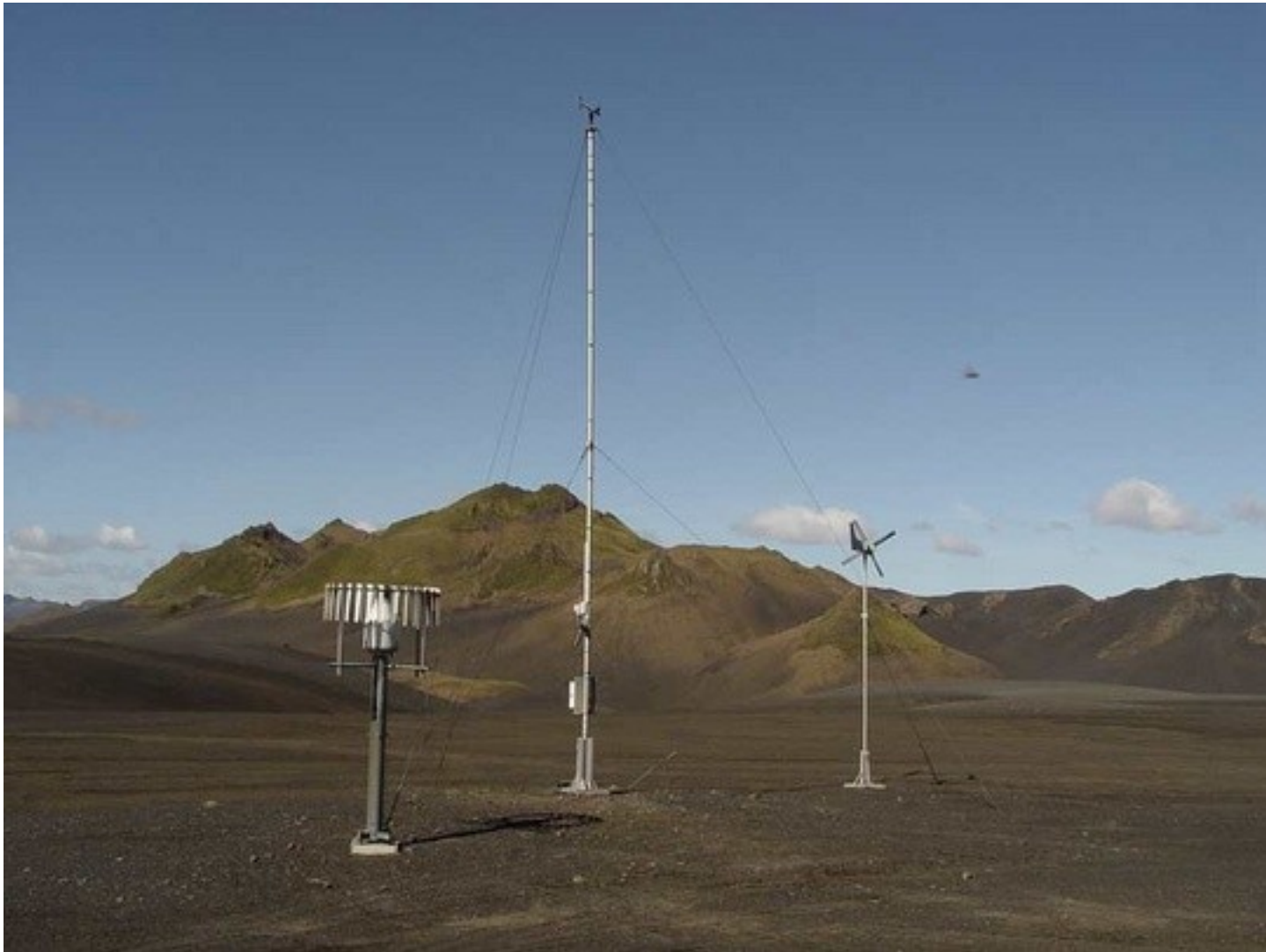


# The QA system of Belgingur





# QA system performance for observations from an automatic weather station in Iceland





# QA system performance

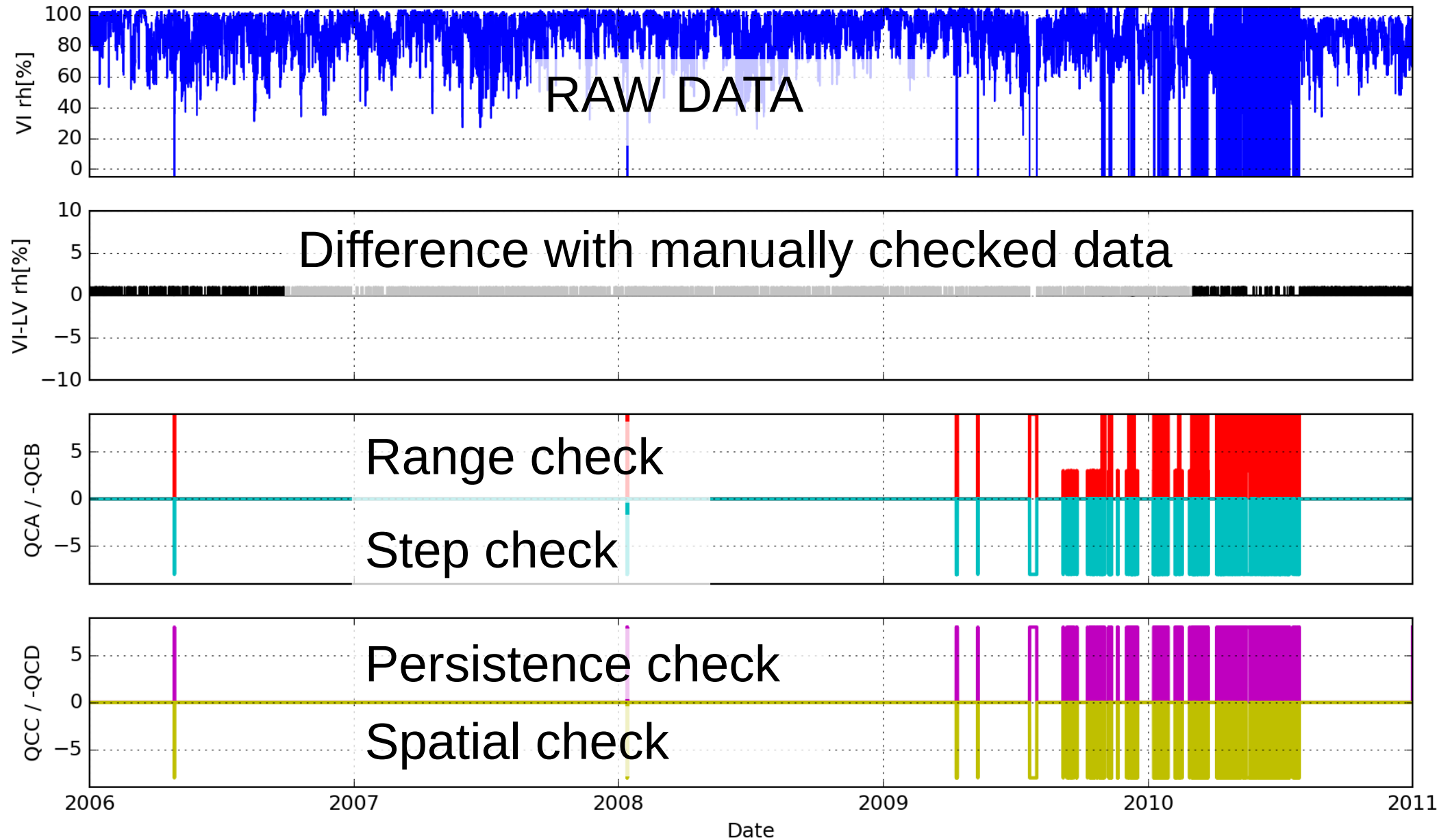
## Typical problems in observed data:

- Large spikes in the input data, i.e. in wind speed during lightning activity and bad weather.
- Icing problems on anemometers and thermometers.
- NANs in observed datums, i.e. due to incorrectly registered observations.
- Measurements which are out of the range of physical values, typically associated with poor instruments measurement technique.
- Faulty or failing instruments, causing an error or a drift in the measured variable.



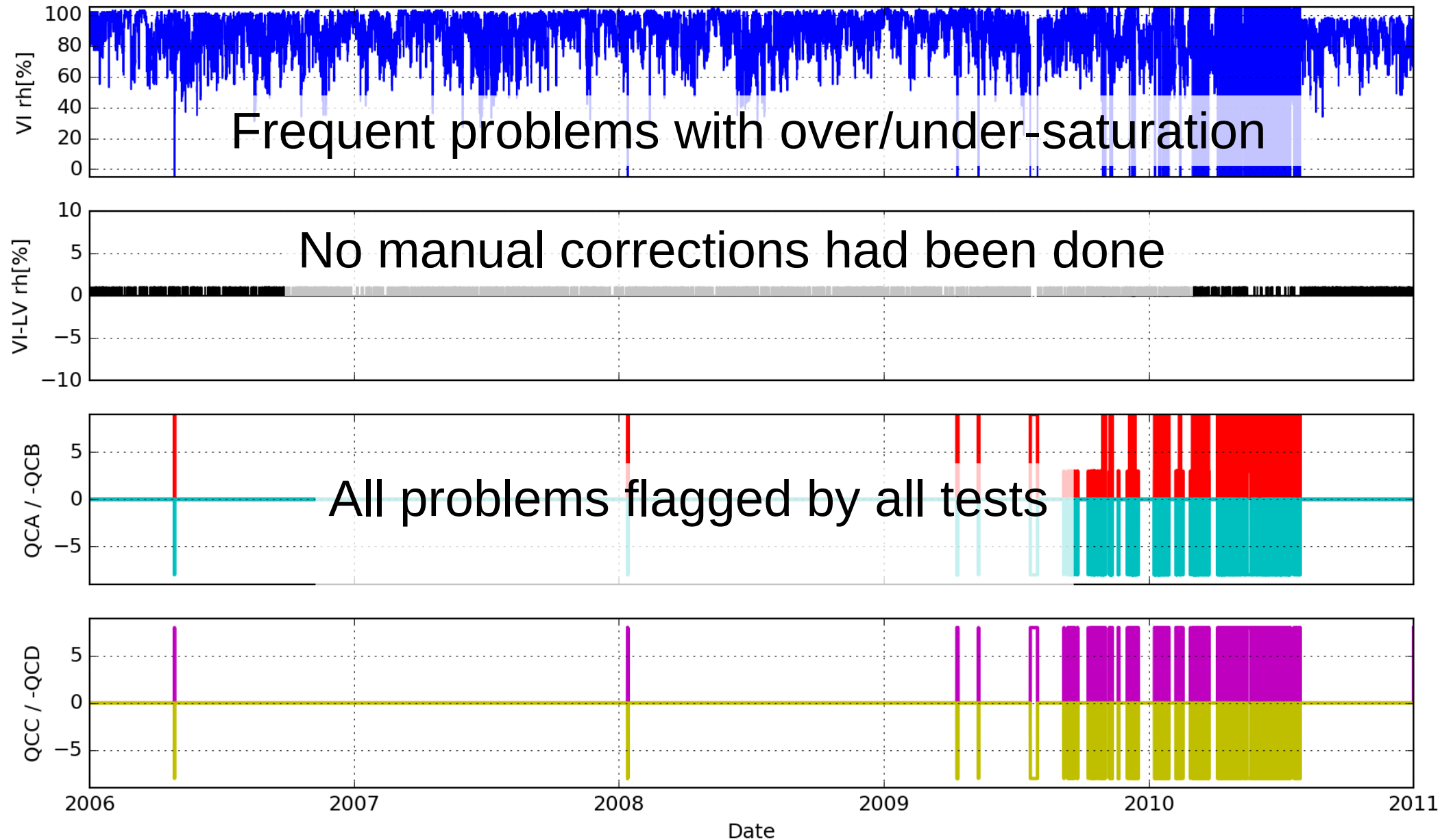


# QA system



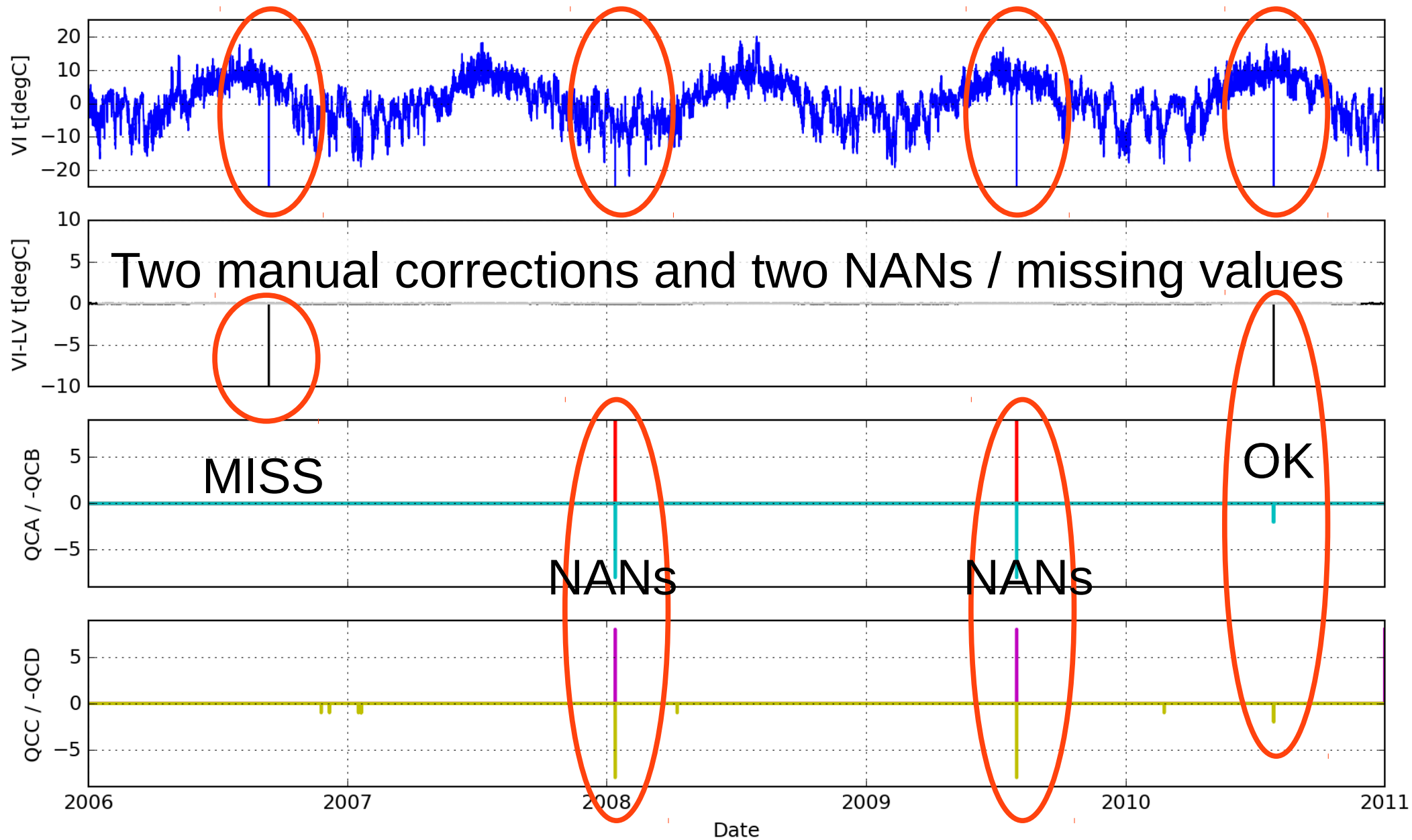


# QA system – Relative humidity





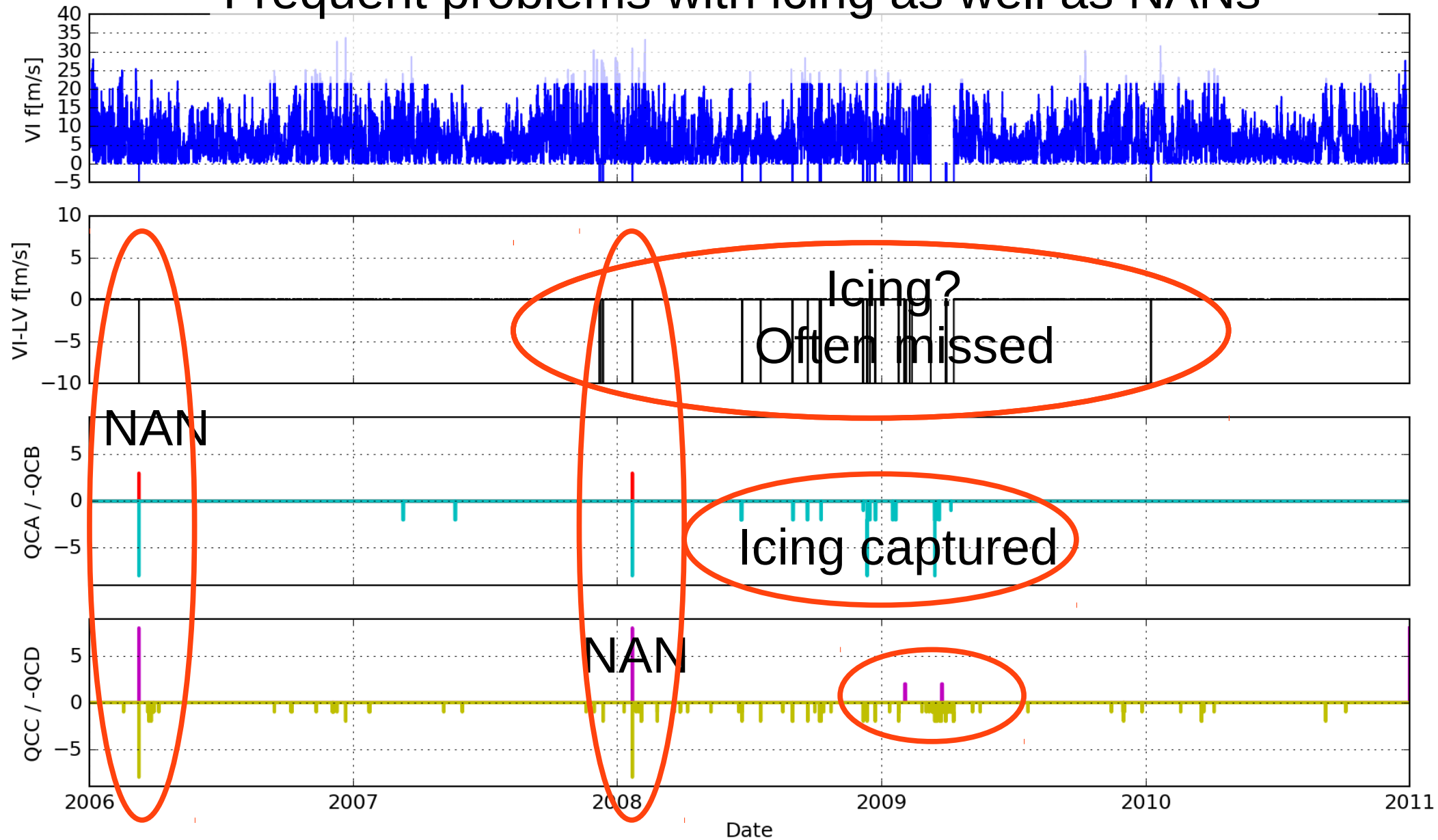
# QA system - Temperature





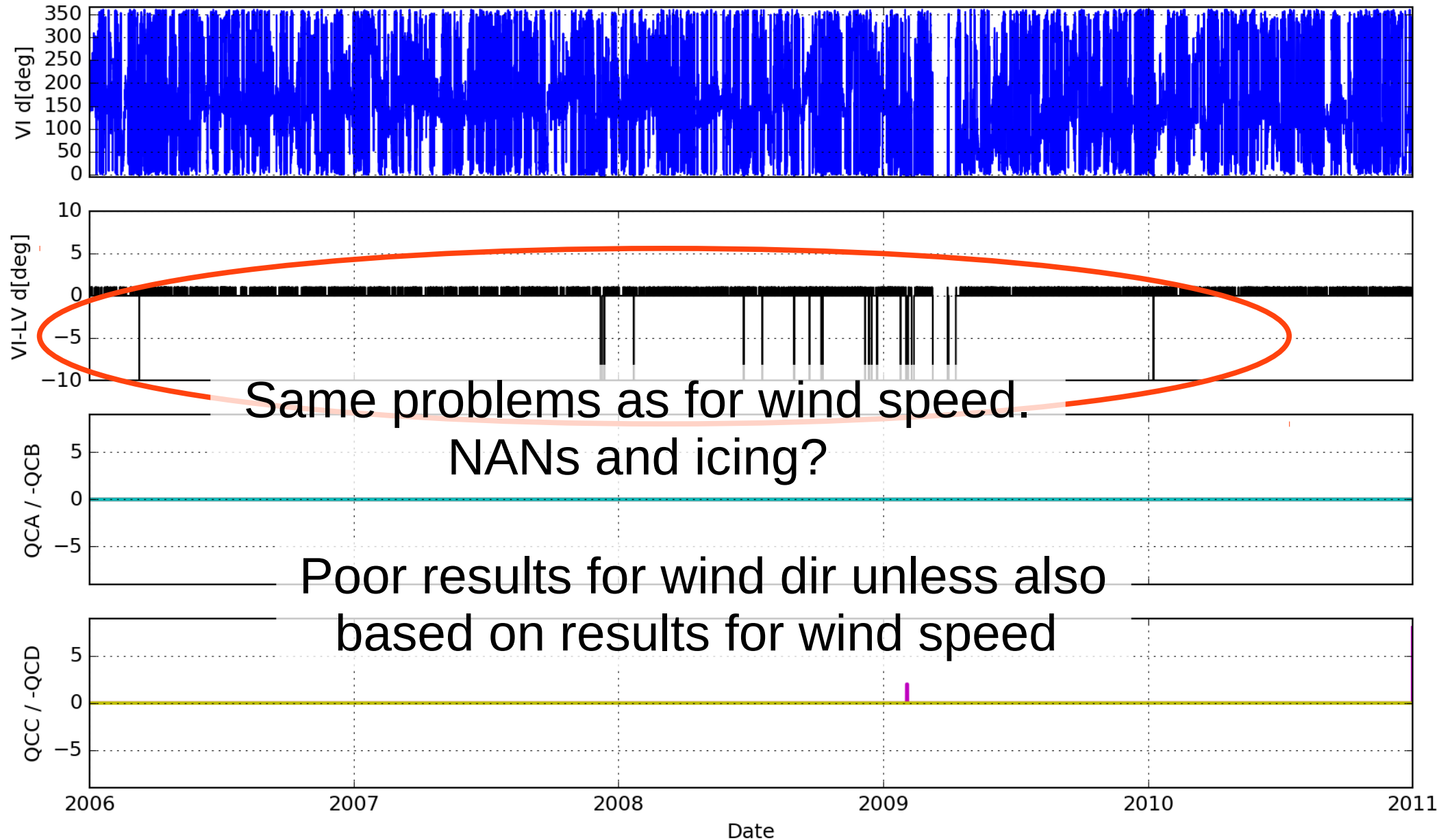
# QA system – Wind speed

Frequent problems with icing as well as NANs





# QA-system – Wind direction





# QA system in Iceland

Performance of QA-system was checked for +10 stations in Iceland by comparison of QA-results with manually checked data:

- Most potential errors are caught by QA-system.
- Necessary improvements to persistence check, with regard to anemometer problems (icing), have been identified.
- Wind direction analysis must take into account QA-results for wind speed.
- Spatial test often fails in complex terrain or in regions with sparse stations.
- Missing records (no-observation) can not be flagged.
- Regional and climatologically relevant criteria/thresholds should be found and used in tests.





# Further development

- Implement additional means of assessment:
  - analyze wind direction dependent on the wind speed result
  - more flexibility for the persistence check
  - the original system was designed mainly for hourly data, what happens if we put 10-minute data?
  - more conditions on the spatial check





# Potential use in Africa

- Speed up the manual inspection of the data
- Spot malfunctioning gauges with the spatial check
- Use in the linear regression point forecast procedure (negatively flagged observations won't be used as input for the method)







Thank you!

