Documentation for Estimation of $PM_{2.5}$ in western US: Total and Attributed to Wildfires and Prescribed Fires

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Abstract

The purpose of this document is to provide detailed information about the estimation of $PM_{2.5}$ (total and attributed to prescribed fires and wildfires) that our work could be reproduced. Figure 1 shows the study area of interest.

Contents

1 Ideas, To Do, Resources, etc

code up fires by type of land coverage

Consider using the work of Westerling et al for a comprehensive fire history (up through 2012) http://science.sciencemag.org/content/313/5789/940, http://www.pnas.org/content/108/32/13165, http://rstb.royalsocietypublishing.org/content/371/1696/20150178 Westerling (2016b,a) Also look into the fire histories referenced in Westerling Westerling (2016b,a): http://fam.nwcg.gov/fam-web/weatherfirecd/fire_files.htm and http://fam.nwcg.gov/fam-web/kcfast/mnmenu.htm See also http://www.nifc.gov

look into the Fire and Smoke Model Evaluation Experiment (FASMEE) http://www.fasmee.net

Compare our results with EPA Downscaler https://www.epa.gov/air-research/downscaler-model-predicting-

Look at Kollanus et al. (2016) again for references for PM2.5 paper, especially the introduction. Consider using NAAPS in our study.

read **?**

read **?**

see also https://www.5280.com/2018/09/can-colorado-burn-its-way-out-of-a-wildfire-crisis/

Could we use inciweb to distinguish prescribed fires?

look up Global Fire Emissions Database (GFED3) - maybe it would be useful for our study as an input to the machine learning? see Liu et al. (2016)

see **?** for potential data sources for ML project

emissions vary by temperature https://cires.colorado.edu/news/wildfire-temperatures-key-better-understand and https://www.atmos-chem-phys.net/18/9263/2018/

read Monitoring Trends in Burn Severity MTBS, 2014. Data Access: Fire Level Geospatial Data. US Department of Agriculture, Forest Service and US Department of Interior, Geological Survey. http://mtbs.gov/data/individualfiredata.html/.

Idea: look at ambulance calls and PM2.5, similar to what Salimi et al. (2016) did in Australia.

read **?**

Database of planned/proposed prescribed burns: WRAP's Fire Emissions Tracking System: http://wrapfets.org/index.cfm

See Di et al., 2016 and Johnston et al., 2012, Rappold et al., 2014 in **?** - combine modelled and monitored/satelite data to estimate PM2.5

See page 11 of **?** for discussion of discrepancies related to burned area estimates

http://www.ptep-online.com/ctan/symbols-a4.pdf

US National Atlas http://nationalmap.gov/small_scale/atlasftp.html

Thought: Using DigitalGlobe for fire data compared to NASA: would have higher spatial resolution, but not consistently viewing all areas (no cost to CU people)

Look up Openair R package

Papers/resources to look into: https://daac.ornl.gov/cgi-bin/dsviewer.pl?ds_id=1293

https://www.fs.fed.us/psw/publications/4451/psw_2009_4451-001.pdf

https://labcit.ligo.caltech.edu/~ethrane/Resources/UNIX/

https://community.tableau.com/thread/141548

According to **?**, GEOS-Chem "can be classified according to emission source", that implies that we could tag the emissions as wildfire vs prescribed fire vs urban. Would there be any advantages of this model over CAMx?

could analyze data with NAAQS and WHO PM2.5 standards

projection/datum info: https://gis.stackexchange.com/questions/664/whats-the-difference-between-a-projec
http://resources.esri.com/help/9.3/arcgisengine/dotnet/89b720a5-7339-44b0-8b58-0f5bf2843393.htm
http://grindgis.com/blog/wgs84-vs-nad83

Monitoring Trends in Burn Severity (MTBS) MTBS, 2016: Data Access: Fire Level Geospatial Data. USDA Forest Service/U.S. Geological Survey, accessed 8 October 2016, https://mtbs.gov/direct-download. Eidenshink, J., B. Schwind, K. Brewer, Z.-L. Zhu, B. Quayle, and S. Howard, 2007: A project for monitoring trends in burn severity. Fire Ecol., 3, 3–21, https://doi.org/10.4996/fireecology.0301003.

Idea: Maybe instead of just distance to closest fire, we should follow the example of [Baek2016] and do distributed lags with concentric circles with information about fires in each concentric circle... also, instead of just distance to fire, maybe we could come up with a variable that is something like [distance*size of fire] since both are important.

Fire stats/records: https://www.nifc.gov/fireInfo/fireInfo_statistics.html

See **?** for description of fire perimeter data that perhaps we could use (CA only)

See **?** for info about MTBS and Active Fire Mapping Program and NWS smoke products. See also Lassman et al **?** cited therein.

Read these papers cited in **?**: Yao and Henderson, 2014; Henderson et al 2011; Liu et al 2015; Gan et al 2017; and look at their sources of PM2.5 data to see if we could add any of those to our project.

2 PM2.5 Surface Paper Notes

Discussion of trends in anthro PM2.5: **?**

2.1 Papers published in Atmospheric Environment - use as style example

Need to go through these papers

- Brokamp et al. (2017) (partially done, done through intro)
- Sampson et al. (2013)
- Anyenda et al. (2016)
- Torvela et al. (2014)
- Whiteman et al. (2014)

Put in Brokamp et al. (2017); Larsen et al. (2017)

3 Papers to cite/discuss in Introduction and/or Discussion

Westerling (2016b,a)

try to find English version http://80.24.165.149/webproduccion/PDFs/15CAP03.PDF

For fire identification, consider using NOAA's Hazard Mapping System and BlueSky

3.1 Notes on Papers

See J. et al. (2016) for statistics about wildfires in western US, e.g., % started by humans, number of fires, etc.

4 Fire attribution paper

revisit **?**
include **?** - does a good job of summarizing the debate about more vs less prescribed burns
sources of fire data **?**, **?**
will need to compare our work to **?**
include Westerling (2016b,a) and Abatzoglou and Williams (2016)
See **?** for an alternative method of attributing PM2.5 to wildfire smoke (instead of CAMx)
See Le et al 2014 **?**
See Huff et al **?**

4.1 text written for the COPD paper - variation of this may be useful

Larsen et al., 2017 Larsen et al. (2017) found that, on average, ground-level $PM_{2.5}$ concentrations increased by 2.9 μg·m^{-3} (2.8, 3.0) when there was a visible wildfire smoke plume overhead (from satellite imagery), as well as a 2.6 ppb (2.5-2.7) increase in O_3. Satellite data provides a wealth of data and can provide information about air quality where monitors are not present. However, satellite imagery inherently comes with a substantial uncertainty in that satellite data describes the entire atmospheric column and not specifically just air pollution at the ground level, where people are breathing.

5 Data Sources for Machine Learning

For the creation of the spatiotemporal daily exposure surface via machine learning, a large number of data sets will be collected as discussed below. The dependent variable will be daily 24-hour $PM_{2.5}$ from monitoring data.

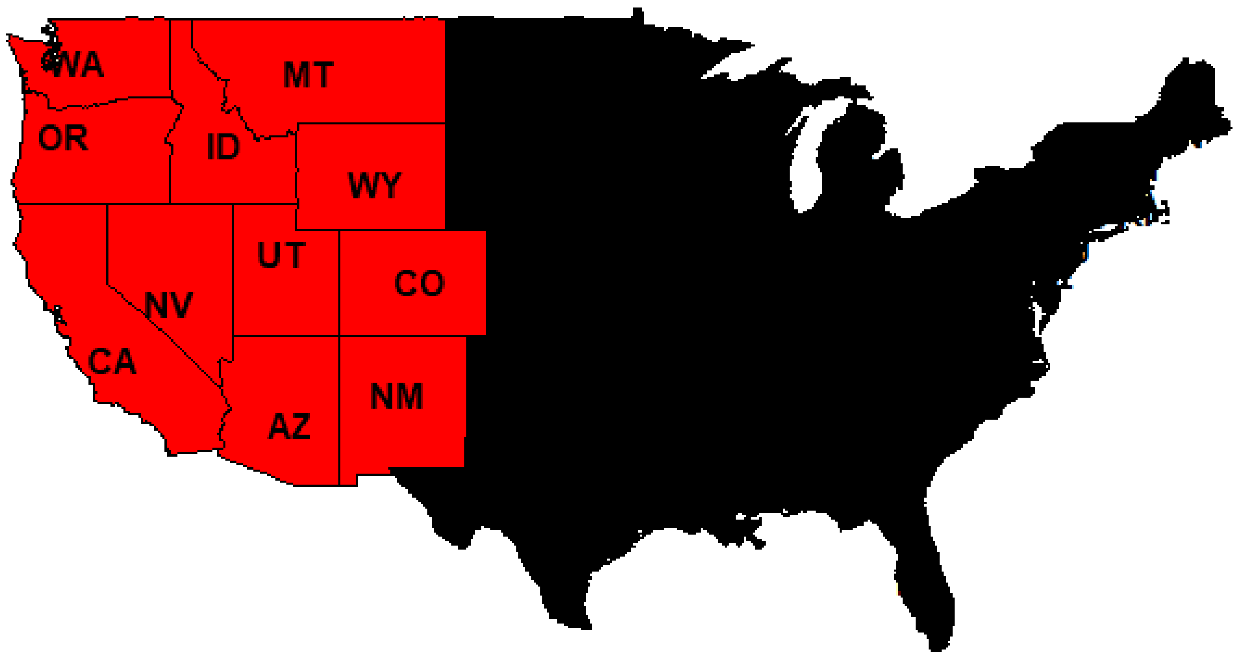

Figure 1: Map of 11-state study area.

All PM2.5 Monitor Locations

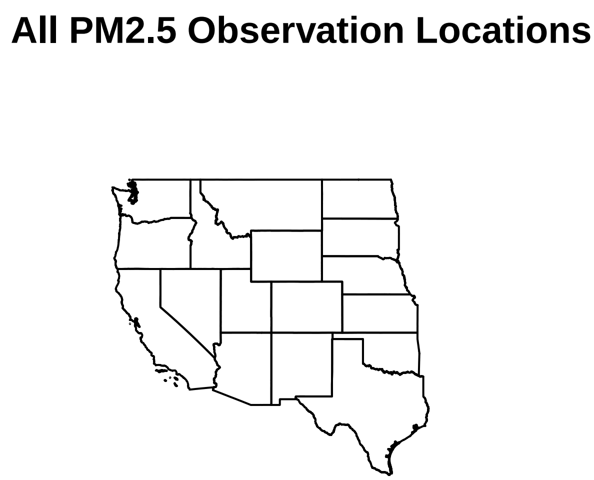

Figure 2: Map of locations of PM2.5 observations for entire study period, 2008 to 2014.

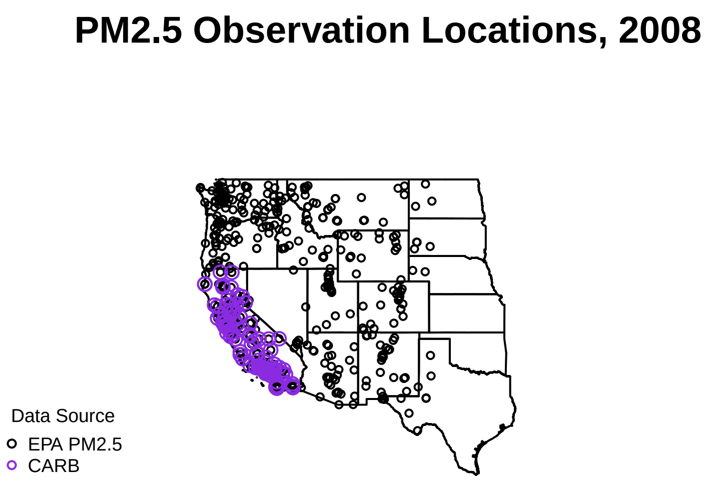

Figure 3: Map of locations of PM2.5 observations during 2008.

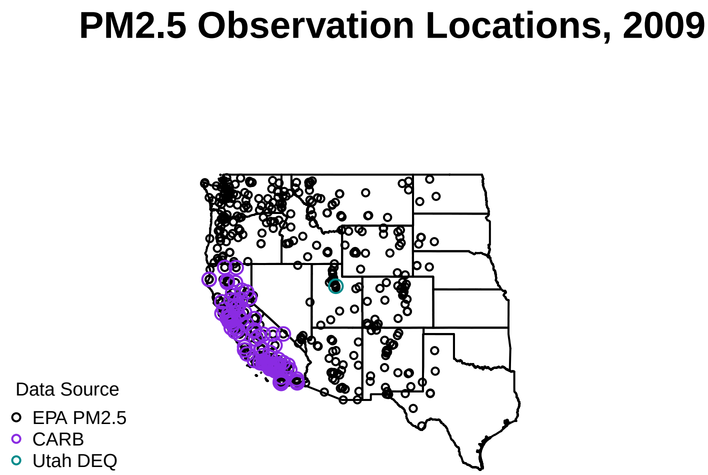

Figure 4: Map of locations of PM2.5 observations during 2009.

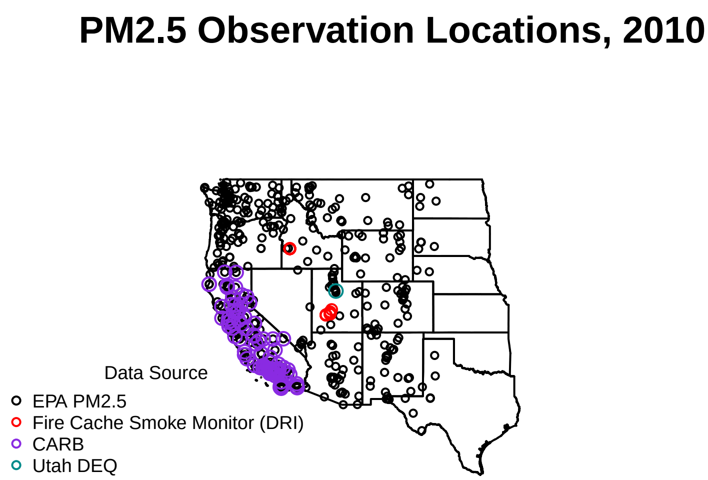

Figure 5: Map of locations of PM2.5 observations during 2010.

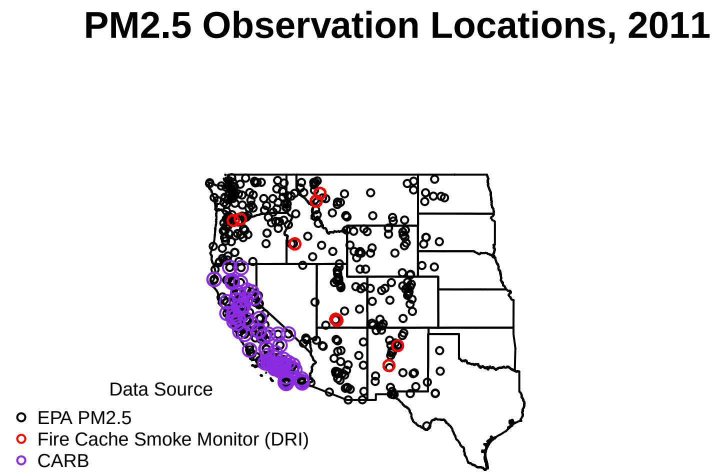

Figure 6: Map of locations of PM2.5 observations during 2011.

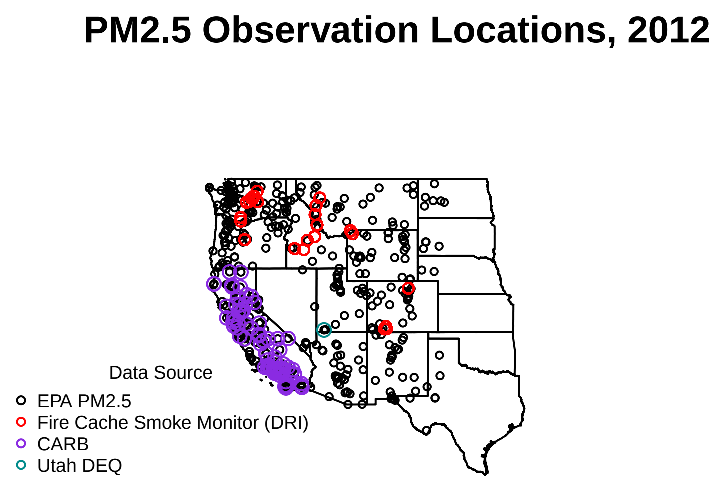

Figure 7: Map of locations of PM2.5 observations during 2012.

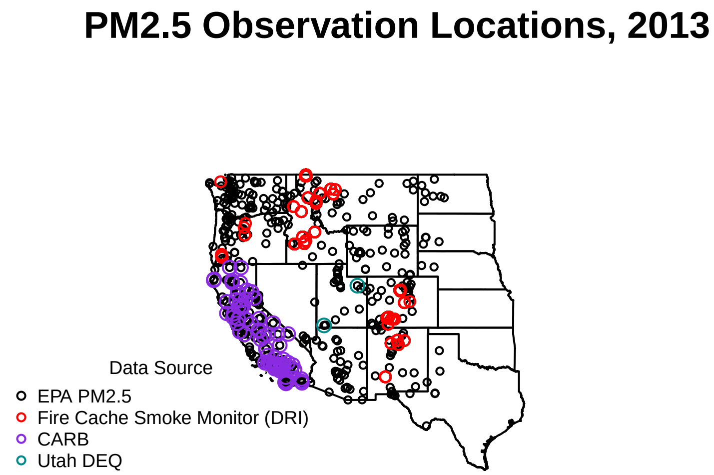

Figure 8: Map of locations of PM2.5 observations during 2013.

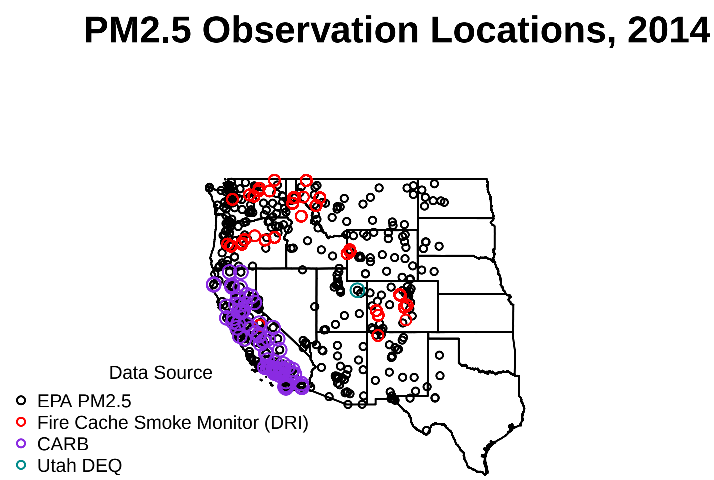

Figure 9: Map of locations of PM2.5 observations during 2014.

5.1 PM2.5 Monitor data from US EPA AQS Air Data Query Tool

Data Source

- **Contact**
  Can email the Air Quality Analysis Group (U.S. EPA Office of Air Quality Planning and Standards) on their website at https://www.epa.gov/outdoor-air-quality-data/forms/contact-us-about-outd
- **Citation/Link**
  United States Environmental Protection Agency. *Pre-Generated Data Files: Daily Summary Files, PM2.5 FRM/FEM Mass (88101) and PM2.5 non FRM/FEM Mass (88502), 2008-2014*. https://aqs.epa.gov/aqsweb/airdata/download_files.html#Daily
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent** 2008 through 2014
- **Acknowledgment**

Download spreadsheet listing all AQS monitors with datums (https://aqs.epa.gov/aqsweb/airdata/aqs_monitors.zip) from "Monitor Listing" at https://aqs.epa.gov/aqsweb/airdata/download_files.html#Meta. The file name is aqs_monitors.csv in the AQS_Daily_Summaries folder in the S3 data.

Brief Description

We will download $PM_{2.5}$ data from both the US EPA AQS Air Data Query Tool (US EPA, 2017c) and the IMPROVE monitors that capture air quality information in more rural areas (US EPA, 2017e) for the 11-state region (Figure 1) including any of the following parameter codes: 88101, 88500, 88502, 81104 (US EPA, 2017a,d,f).

Notes

File Format

Data Filtering and Processing

Final Variable(s)

Methods

1.
2.

Quality Control

Script Names

1.

Data File Names

1. daily_88101_2008.csv
2. daily_88101_2009.csv
3. daily_88101_2010.csv
4. daily_88101_2011.csv
5. daily_88101_2012.csv

6. daily_88101_2013.csv
7. daily_88101_2014.csv
8. daily_88502_2008.csv
9. daily_88502_2009.csv
10. daily_88502_2010.csv
11. daily_88502_2011.csv
12. daily_88502_2012.csv
13. daily_88502_2013.csv
14. daily_88502_2014.csv

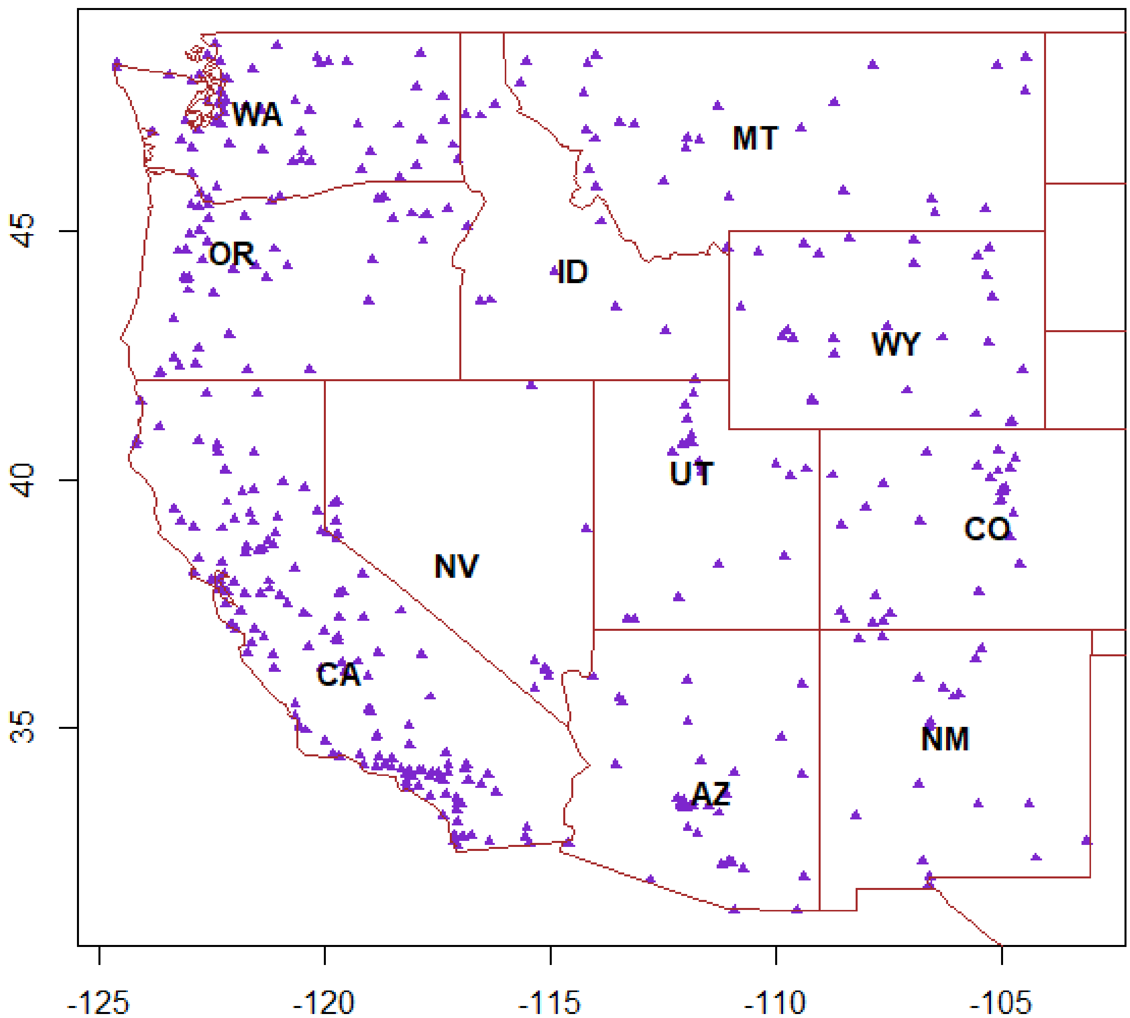

Figure 10: Map of 88101 and 88502 $PM_{2.5}$ Monitors.

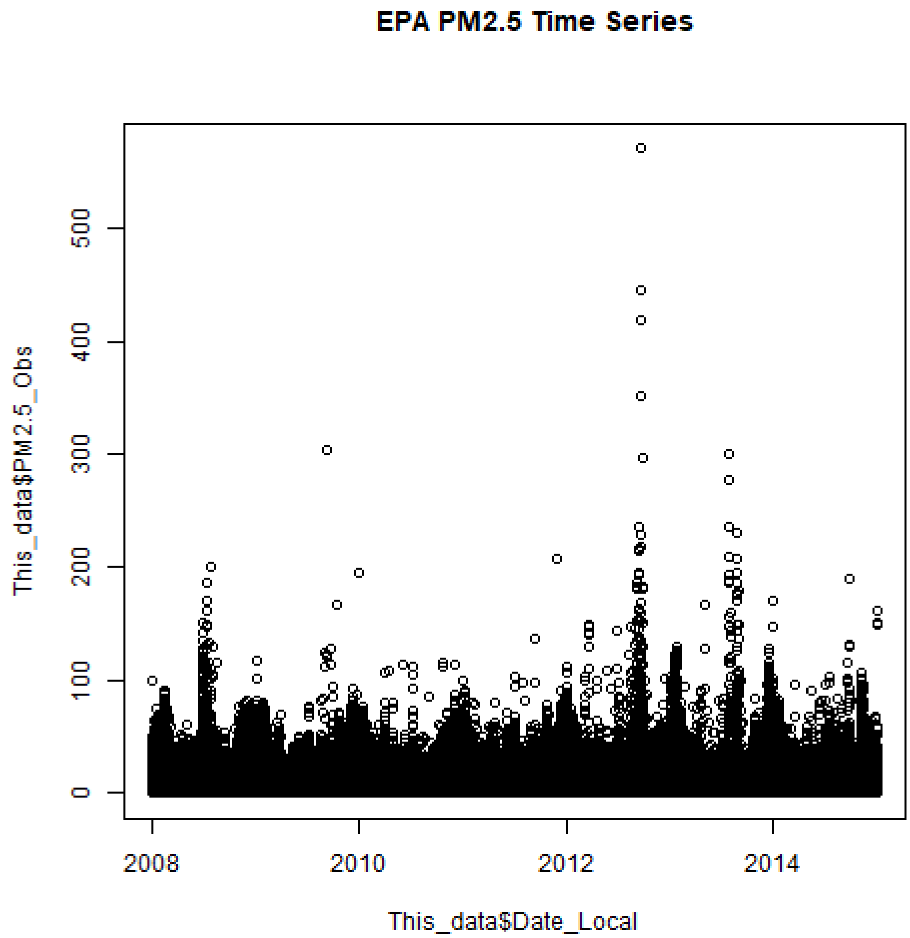

Figure 11: EPA PM2.5 time series.

5.2 EPA PM2.5 Plots

5.3 $PM_{2.5}$ data from the Federal Land Manager Environmental Database

Data Source

- **Citation/Link** http://views.cira.colostate.edu/fed/DataWizard/Default.aspx
- **Download Date** March 15, 2018
- **Data (local)** $PM_{2.5}$ data from the Federal Land Manager Environmental Database
- **Geographic Extent** Nationwide
- **Temporal Extent** January 1, 2008 - December 31, 2014
- **Acknowledgment** - need to fill in

Downloading IMPROVE Aerosol, RHR II (New Equation) data (one parameter at at time):

1. Reports: Raw data
2. Datasets:"IMPROVE Aerosol, RHR II (New Equation)"
3. Sites: select all
4. Parameters:
   (a) Mass, PM2.5 (Fine): Code MF, Type PM2.5, Units ug/m$\hat{3}$ LC AQS ID 88101
   (b) Mass, PM2.5 Reconstructed (Fine): Code RCFM, Type PM2.5 Units ug/m$\hat{3}$ LC, AQS ID 88401
5. Select Dates: By Years and Months: 2008-2014; select all months
6. Aggregations: Non-aggregated
7. Fields: Select All
8. Options: Text File; Generate one file containing all the data; Comma delimited, Standard ("wide" format); Data & Metadata, Display Column Headers, Don't Display Section Titles, String Quotes: Double Quotes, Missing Values (blank); Date Format: 3/14/2002; Display Results: In a separate browser window; Show Report Log
9. Submit

Repeat the downloading steps above, except replace step #2 with these Datasets and parameters:

1. IMPROVE Aerosol, RHR III (DRAFT - Preliminary Most Impaired Days dataset)
   (a) Mass, PM2.5 (Fine) is listed twice - these turned out to be the same data

File Formats

csv

Original Data File Names

1. Federal_Land_Manager_IMPROVE_RHR_II_88101_20183151757452922Mvw0s.csv
2. Federal_Land_Manager_IMPROVE_RHR_II_88401_20185113533660420xLwJ.csv
3. Federal_Land_Manager_RHR_III_88101_first_param_201851152033932P22My0.csv

5.4 $PM_{2.5}$ data from the Fire Cache Smoke Monitor Archive

Data Source

- **Contact** Josh Walston at 775-673-7624; Amber Ortega directed us to the website and Scott Landis suggested that a good person to contact about the page would be Mike Broughton from the US Forest Service (michaelbroughton@fs.fed.us)
- **Citation/Link** https://wrcc.dri.edu/cgi-bin/smoke.pl
- **Data (local)** $PM_{2.5}$ data from the Fire Cache Smoke Monitor Archive
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

Notes

Several of the files were password protected, so we contacted Josh and they were able to unlock most of them. As of March 20, 2018, only "Smoke NCFS E-BAM # 3 is still password protected. (Need to try calling Josh again.) Here are some comments that the system administrator passed along to us (via Josh): the data does not get quality controlled, so we should do our own qa/qc. The monitors/data were designed for the fire community to see data in real time, not for research purposes. If we want to speak with the people who ran the monitors, we should contact Josh and the director can probably put us in contact.

Update 2018-05-2018: sent email to Josh requesting the "Smoke NCFS E-BAM # 3" data with flags and with the other formatting settings we used on the other files. Also asked how to determine which datum is assoiciated with the latitude/longitude data.

These monitors were not included because the website indicated that it did not have data during our study period (January 1, 2008 - December 31, 2014):

1. Smoke E-BAM 418
2. Smoke E-BAM 591
3. Smoke E-BAM 592
4. Smoke E-BAM 882
5. Smoke E-BAM 969
6. Smoke USFS R2-922
7. Smoke USFS R2-923
8. Smoke USFS R2-924
9. Smoke USFS R8-34
10. Smoke USFS R8-35
11. Smoke USFS R8-55
12. Smoke USFS R8-56
13. Smoke USFS 3015
14. Smoke USFS 3016
15. Smoke USFS R9-3017
16. Smoke USFS R9-3018
17. RSF Smoke Monitor 1
18. Lolo NF Smoke Monitor #1
19. Lolo NF Smoke Monitor #2

File Formats

.csv

Data Filtering and Processing

Final Variable(s)

Methods

1.
2.

Quality Control

Script Names

1.

Original Data File Names

1. Fire_Cache_Smoke_DRI_Cache-NCFS-EBAM#3-with-flags.csv
2. Fire_Cache_Smoke_DRI_FWS_Smoke_N1.csv
3. Fire_Cache_Smoke_DRI_Smoke_N11.csv
4. Fire_Cache_Smoke_DRI_Smoke_N13.csv
5. Fire_Cache_Smoke_DRI_Smoke_N15.csv
6. Fire_Cache_Smoke_DRI_Smoke_N16.csv
7. Fire_Cache_Smoke_DRI_Smoke_N17.csv
8. Fire_Cache_Smoke_DRI_Smoke_N19.csv
9. Fire_Cache_Smoke_DRI_Smoke_N20.csv
10. Fire_Cache_Smoke_DRI_Smoke_N21.csv
11. Fire_Cache_Smoke_DRI_Smoke_N22.csv
12. Fire_Cache_Smoke_DRI_Smoke_N23.csv
13. Fire_Cache_Smoke_DRI_Smoke_N24.csv
14. Fire_Cache_Smoke_DRI_Smoke_N25.csv
15. Fire_Cache_Smoke_DRI_Smoke_N65.csv
16. Fire_Cache_Smoke_DRI_Smoke_N66.csv
17. Fire_Cache_Smoke_DRI_Smoke_N67.csv
18. Fire_Cache_Smoke_DRI_Smoke_N68.csv
19. Fire_Cache_Smoke_DRI_Smoke_N69.csv
20. Fire_Cache_Smoke_DRI_Smoke_N84.csv
21. Fire_Cache_Smoke_DRI_Smoke_N86.csv
22. Fire_Cache_Smoke_DRI_Smoke_N215.csv
23. Fire_Cache_Smoke_DRI_Smoke_N216.csv
24. Fire_Cache_Smoke_DRI_Smoke_N217.csv
25. Fire_Cache_Smoke_DRI_Smoke_E_BAM_52.csv
26. Fire_Cache_Smoke_DRI_Smoke_E_BAM_65.csv
27. Fire_Cache_Smoke_DRI_Smoke_E-BAM_231.csv
28. Fire_Cache_Smoke_DRI_Smoke_E-BAM_840.csv
29. Fire_Cache_Smoke_DRI_Smoke_E-BAM_866.csv
30. Fire_Cache_Smoke_DRI_Smoke_E-BAM_925.csv
31. Fire_Cache_Smoke_DRI_Smoke_NCFS_E-BAM_N1.csv

32. Fire_Cache_Smoke_DRI_Smoke_NCFS_E-BAM_N2.csv
33. Fire_Cache_Smoke_DRI_Smoke_USFS_R1-39.csv
34. Fire_Cache_Smoke_DRI_Smoke_USFS_R1-52.csv
35. Fire_Cache_Smoke_DRI_Smoke_USFS_R1-53.csv
36. Fire_Cache_Smoke_DRI_Smoke_USFS_R1-306.csv
37. Fire_Cache_Smoke_DRI_Smoke_USFS_R1-307.csv
38. Fire_Cache_Smoke_DRI_Smoke_USFS_R2-69.csv
39. Fire_Cache_Smoke_DRI_Smoke_USFS_R2-78.csv
40. Fire_Cache_Smoke_DRI_Smoke_USFS_R2-264.csv
41. Fire_Cache_Smoke_DRI_Smoke_USFS_R2-265.csv
42. Fire_Cache_Smoke_DRI_Smoke_USFS_R3-28.csv
43. Fire_Cache_Smoke_DRI_Smoke_USFS_R3-86.csv
44. Fire_Cache_Smoke_DRI_Smoke_USFS_R5-39.csv
45. Fire_Cache_Smoke_DRI_Smoke_USFS_R5-49.csv
46. Fire_Cache_Smoke_DRI_Smoke_USFS_R8-33.csv
47. Fire_Cache_Smoke_DRI_Smoke_USFS_R9-15.csv
48. Fire_Cache_Smoke_DRI_Smoke_USFS_R9-16.csv
49. Fire_Cache_Smoke_DRI_Smoke_USFS_R9-17.csv
50. Fire_Cache_Smoke_DRI_Smoke_USFS_R9-60.csv
51. Fire_Cache_Smoke_DRI_Smoke_NPS_Yosemite_01_California.csv

Processed/Cleaned Data File Names

1.
2.

Download instructions

1. https://wrcc.dri.edu/cgi-bin/smoke.pl
2. Hover over the appropriate drop-down menu and click on the monitor you want to download e.g., “Cache Monitors” then “Smoke #11”
3. On the left-side menu, click on “Data Details”
4. Set the starting date: January 1, 2008 (or as far back as it goes if it doesn’t go back to 2008)
5. Set the ending date: December 31, 2014 (or the last date possible if it ends before 2014)
6. Elements (ignore - default is to include all elements)
7. Options
8. Excel (.xls) (It had html code in the file if I chose other options.)
9. Data Source: Original
10. Represent missing data as: -9999.
11. Include data flags: Yes
12. Date format: MM/DD/YYYY hh:mm
13. Time format: LST 0-23
14. Table Header: Column header short descriptions
15. Field Delimiter: comma (,)
16. Select the Units: Metric
17. Leave Sub interval windows set to: January 01, December 31, Hours: 00 and 24
18. Submit Info

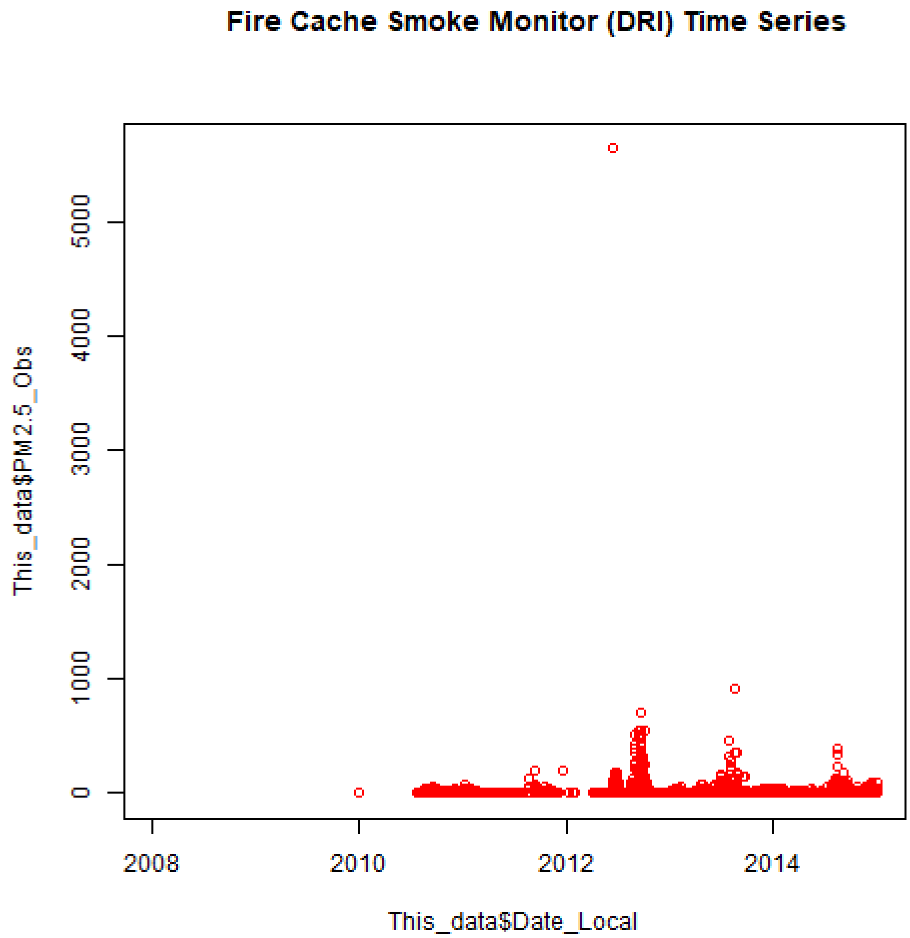

Figure 12: Fire Cache Smoke Monitor (DRI) time series.

19. Open in excel
20. Save as: Fire_Cache_Smoke_DRI_*.csv Where * is the monitor name with spaces replaced with underscore and # symbols replaced with the letter N, e.g., the file name for monitor "Smoke #11" is "Fire_Cache_Smoke_DRI_Smoke_N11.csv"
21. Upload file to S3 bucket: https://732215511434.signin.aws.amazon.com/console
22. Click on S3
23. Earthlab-reid-group
24. Fire_Cache_Smoke_DRI (folder)
25. Check the following:
26. The name of the monitor is in cell A1
27. The header is spread across rows 2-4
28. There are 34 columns of data (goes through columns "AH" in excel)
29. Concentration in the 11th ("K") columns
30. List the file names in the overleaf documentation (PM25_Fire_Cache_Smoke_Monitor_Archive.tex)

5.5 Fire Cache Smoke Monitor (DRI) Plots

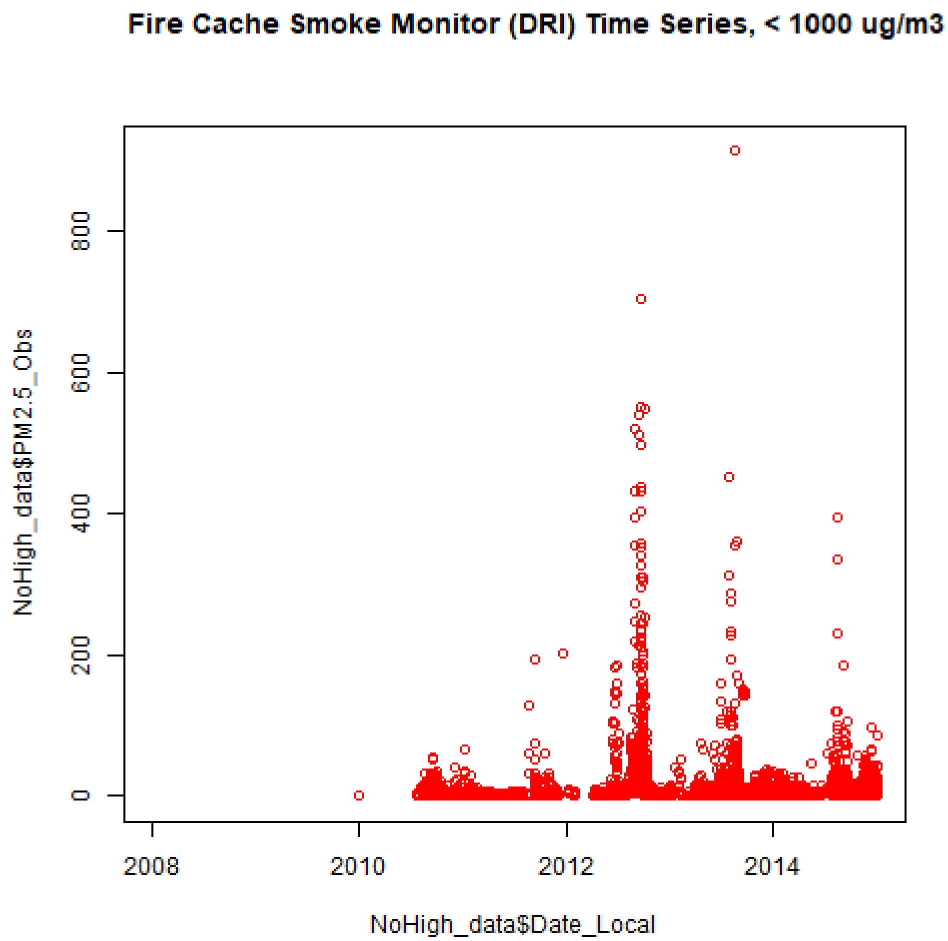

Figure 13: Fire Cache Smoke Monitor (DRI) time series without data above 1000 ug/m3 so that the majority of data can be seen.

5.6 California State Air Quality and Meteorological Information System (AQMIS)

Data Source

- **Contact** Denise Odenwalder, Denise.Odenwalder@arb.ca.gov
- **Citation/Link** To AQMIS: https://www.arb.ca.gov/aqmis2/aqmis2.php
- **Data (local)**
- **Geographic Extent** Whole state of California, wherever there are monitors
- **Temporal Extent** 2008-2014, daily averages
- **Acknowledgment** California Air Resources Board was very helpful in gathering and sending us this data.

Brief Description

- PM2.5 measurements at all monitoring stations in CA
- Some entries are 24-hour measurements while others are the average of hourly measurements
- One entry per 3 days

Notes

Reached out to aqmis@arb.ca.gov after determining that there was data being collected in CA that is not published on the EPA AQS website. They emailed us within a week, with a file of the data we requested.

File Formats

xlsx spreadsheet

Data Filtering and Processing

Final Variable(s)

Methods

1.
2.

Quality Control

Script Names

1.

Original Data File Names

1.
2.

Processed/Cleaned Data File Names

1.
2.

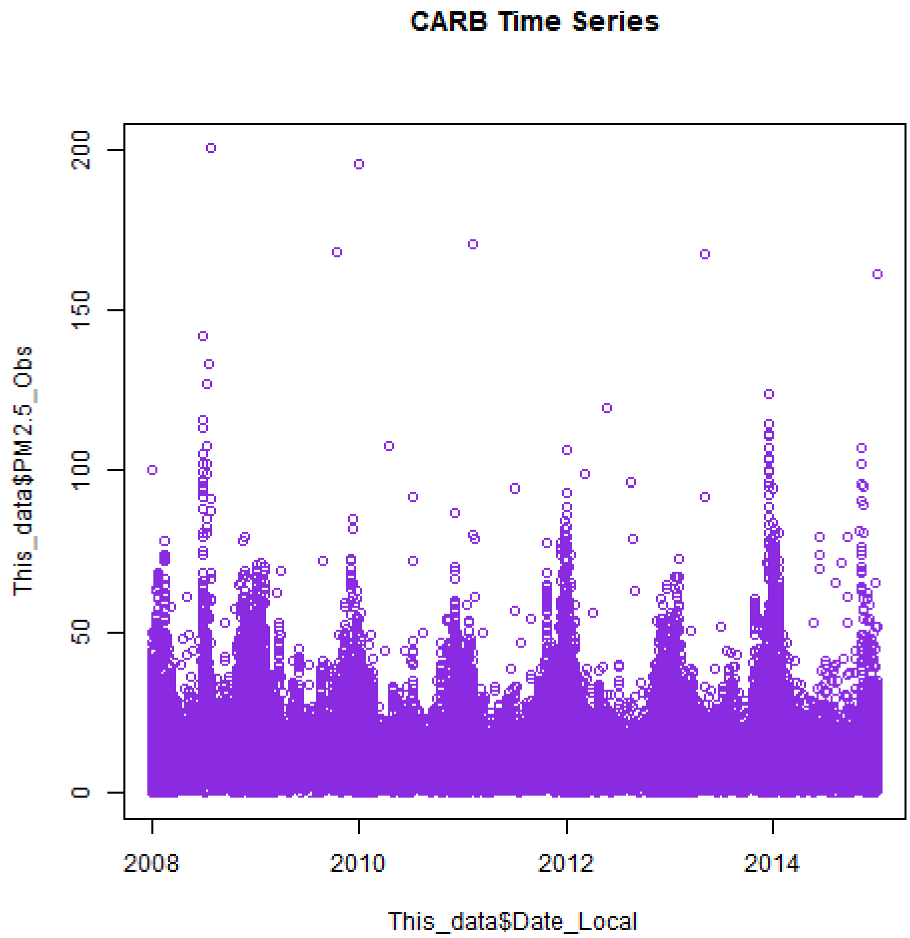

Figure 14: CARB time series.

5.7 CARB Plots

5.8 $PM_{2.5}$ Monitor data from Uintah Basin

Data Source

- **Contact** Seth Lyman
- **Citation/Link** seth.lyman@usu.edu
- **Data (local)** $PM_{2.5}$ measurements from 10 sites in Uintah Basin, Utah
- **Geographic Extent** Uintah Basin, Utah
- **Temporal Extent** October 2009 - March 2017
- **Acknowledgment** $PM_{2.5}$ data from the Uintah Basin were provided by Seth Lyman at Utah State University.

Brief Description

$PM_{2.5}$ data were provided by Seth Lyman at Utah State University via email on January 16, 2018. The .xlsx file has $PM_{2.5}$ data from 10 stations during 2009-2017. The .png file has the longitude and latitude of each site.

Notes

Additional information from Seth's email:
"I've attached most of the PM2.5 observations that have ever been collected in the Uintah Basin. What are in the Excel file are 24-hr average data. Data from Roosevelt, Vernal, Ouray, Red Wash, Myton, and Rangely are from the EPA AQS database.
Data from Horsepool are from a BAM 1020 monitor that we operate every winter. Data in Ft. Duchesne and Randlett are 24-hr filter samples that were analyzed gravimetrically. Data from Rabbit Mountain are from a BAM 1020, and data through mid-2013 are in the AQS database.

I have hourly data from Horsepool and Rabbit Mountain if you'd rather have that.

Site locations are given in the list of monitoring stations for the Basin below."

The .png file is easier to read in some programs than others, e.g., it looks fine in "Paint," but not "Photos."

File Formats

Excel and png

Data Filtering and Processing

FinalPM2.5_multiyear_thruwint2017_sheet1.csv is the first sheet of FinalPM2.5_multiyear_thruwint2017.xlsx converted to .csv, and the second row of the header was merged into the first (24hr avg PM2.5).

FinalPM2.5_multiyear_thruwint2017_GISsheet.csv is the third sheet of FinalPM2.5_multiyear_thruwint2017. converted to .csv and gives the latitude and longitude of each site. This sheet originally did not have location information from the Rangely site, so this was filled in by hand with the numbers form UintahBasinSiteLocations.png.

Final Variable(s)

Methods

1.

2.

Quality Control

Script Names

1.

Original Data File Names

1. FinalPM2.5_multiyear_thruwint2017.xlsx
2. UintahBasinSiteLocations.png

Processed/Cleaned Data File Names

1. FinalPM2.5_multiyear_thruwint2017_sheet1.csv
2. UintahBasinSiteLocations.png

5.9 $PM_{2.5}$ data from PCAPS in the Salt Lake Valley

Data Source

- **Contact** Dr. Geoff Silcox in Chemical Engineering at the University of Utah (geoff@chemeng.utah.edu)
- **Citation/Link** Publication: https://www.sciencedirect.com/science/article/pii/S1352231011011204 (Data was received from Dr. Silcox via email on February 6, 2018.)
- **Data (local)** $PM_{2.5}$ data from the Persistent Cold Air Pool Study (PCAPS)
- **Geographic Extent** Salt Lake Valley
- **Temporal Extent** January - February, 2011
- **Acknowledgment** Dr. Geoff Silcox

Brief Description

Notes

File Formats

.xlsx

Data Filtering and Processing

PCAPS_Site_Locations.csv is the same data as Table 1 of final_publication.pdf, and has the site locations and elevation.

Final Variable(s)

Methods

1.
2.

Quality Control

Script Names

1.

Original Data File Names

1. final_publication.pdf (Publication of paper)
2. MiniVol_data.xlsx

Processed/Cleaned Data File Names

1. MiniVol_data.csv
2. PCAPS_Site_Locations.csv

5.10 Utah Department of Environmental Quality

Data Source

- **Contact**
- **Citation/Link** http://www.airmonitoring.utah.gov/dataarchive/archpm25.htm
- **Data (local)**
- **Geographic Extent** Varies...
- **Temporal Extent** Hourly Value CSVs
- **Acknowledgment**

Brief Description

PM2.5 data from all monitoring stations in Utah

Notes

There was a lot of overlap with the EPA AQS data, so we took data only from the PM2.5 stations not reported by the EPA. This ended up being one or more of three stations (NP, HC, and RS) for 2009, 2010, 2012, and 2013.

Information about the monitoring stations: http://www.airmonitoring.utah.gov/network/Counties.htm

Meta information about monitors obtained from http://www.airmonitoring.utah.gov/dataarchive/2016DailyMaxPM25.pdf

File Formats

Data Filtering and Processing

Final Variable(s)

Methods

1.

2.

Quality Control

Script Names

1.

Original Data File Names

1.

2.

Processed/Cleaned Data File Names

1.

2.

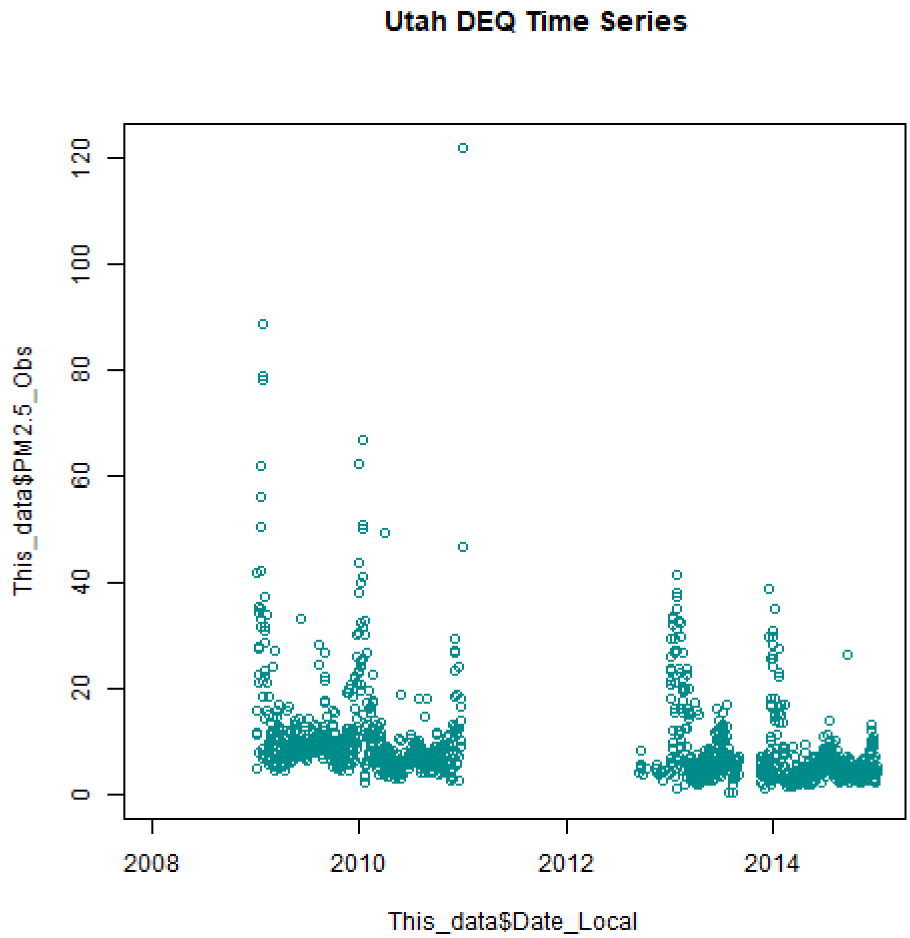

Figure 15: Utah DEQ time series.

5.11 Utah DEQ Plots

5.12 MODIS AOD

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

We will use AOD estimates from the Deep Blue retrieval algorithm for AOD from the MODIS instrument on the NASA Terra and Aqua satellites (MOD04_L2 and MYD04_L2) (Sayer et al., 2013). The MODIS product is available twice daily at a 10 km spatial resolution for cloud-free scenes and is available longer than our 2008-2014 study period (NASA LAADS DAAC, 2017a,b).

AOD products use cloud filtering algorithms that often remove pixels in the center of the smoke plumes because they are assumed to be clouds due to high reflectivity (Kondragunta and Seybold, 2009). Given that these can be in the middle of smoke plumes, often the locations most heavily impacted by smoke have missing data for a key variable, AOD. In our previous work in summer in California when rain clouds are incredibly rare, we could be confident that missing values not along the coast were not clouds. However, for this larger study region and time period, this will be a bigger challenge. We will attempt to isolate smoke plumes from true clouds using satellite imagery and smoke plume polygons from NOAA's Hazard Mapping System Fire Smoke Product (NOAA OSPO, 2017). We will then estimate missing values within validated smoke plumes, but not within clouds, using radial basis functions as was done in our previous work (Reid et al., 2015). Radial basis functions are exact interpolation functions that will return observed AOD values where they exist but can interpolate higher values than nearby observations in missing locations, which is needed since the missing values were removed due to their high reflectivity (Reid et al., 2015).

Notes

File Format

.hdf

Data Filtering and Processing

Final Variable(s)

Methods

1. Step 1: Download the MODIS AOD data sets from both Terra and Aqua sensors:

   Using the NASA EarthData online search tool, search for the 'MOD04' (Terra) data set. Set temporal extent by drawing polygon and set spatial extent by adjusting the appropriate filter on the web interface. Select the collection and proceed to download data. For data download options, specify "Stage for Delivery" through the "FTPPull" distribution option. Specify the email address for orders to be sent to. Orders will be sent to your email with instructions on how to connect to the FTP server and pull the ordered data into your local workspace through the command line. Because the amount of data being requested is large, the orders will come through several separate emails. Repeat this step for the 'MYD04'

(Aqua) data set. All of the raw downloaded data from this step will be in .hdf file format.

2. Step 2: Set up file system for data processing:

   Create a directory locally named ‘collected_data‘. In this directory, make two child directories named "MOD04_terra" and "MYD04_aqua". Follow instructions in email to download data through FTP into the appropriate MODIS directory (‘MOD04_terra‘ or ‘MYD04_aqua‘) depending on whether the order is from the Terra or Aqua sensor.

3. Step 3: Extract lat, long, and aod values from .hdf files and save into .csv files

   Run script ‘modis_aod_create_csv_file.py‘. This script will take all the .hdf files that you have downloaded and store the lat, long and aod value for non-null pixels from the ‘Deep_Blue_Aerosol_Opt
   SDS. A .csv file will be created for each corresponding .hdf file.

4. Step 4: Create .shp file for each .csv file

   Run ‘modis_aod_convert_csv_to_shapefile.py‘. This script will read in the .csv files and convert them to .shp files using multiprocessing, which speeds up the process.

5. Step 5: Project .shp files to US Albers Equal Area Conic

   Run ‘modis_aod_project_to_albers.py‘. This script will reproject the .shp files to be US Albers Equal Area Conic (ESRI:102003).

6. Step 6: Combine .shp files for same date and convert to raster with 10km resolution

   Run ‘modis_aod_create_daily_averages.py‘. This will combine all .shp files from the same date and then produce a raster for each with a 10km resolution. Then, the interpolated grids are clipped to the 11 western states (our study area) with a 100km buffer.

7. Step 7: Extract MODIS AOD value at EPA monitor locations

   Using ExtractValuesToPoints tool in ArcGIS.

Quality Control

Script Names

1. modis_aod_create_csv_file.py
2. modis_aod_convert_csv_to_shapefile.py
3. modis_aod_project_to_albers.py
4. modis_aod_create_daily_averages.py

Data File Names

1. western_states_merge.shp

5.13 GASP-West AOD

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

We will use AOD estimates from the Geostationary Operational Environmental Satellite West (GOES-West) Aerosol Smoke Product (GASP-West AOD). The GASP product is available at a 4 km resolution at nadir with retrievals every 30 minutes during daylight hours and is available from 2006 onward (NOAA NCEI, 2017).

AOD products use cloud filtering algorithms that often remove pixels in the center of the smoke plumes because they are assumed to be clouds due to high reflectivity (Kondragunta and Seybold, 2009). Given that these can be in the middle of smoke plumes, often the locations most heavily impacted by smoke have missing data for a key variable, AOD. In our previous work in summer in California when rain clouds are incredibly rare, we could be confident that missing values not along the coast were not clouds. However, for this larger study region and time period, this will be a bigger challenge. We will attempt to isolate smoke plumes from true clouds using satellite imagery and smoke plume polygons from NOAA's Hazard Mapping System Fire Smoke Product (NOAA OSPO, 2017). We will then estimate missing values within validated smoke plumes, but not within clouds, using radial basis functions as was done in our previous work (Reid et al., 2015). Radial basis functions are exact interpolation functions that will return observed AOD values where they exist but can interpolate higher values than nearby observations in missing locations, which is needed since the missing values were removed due to their high reflectivity (Reid et al., 2015).

Notes

websites: https://www.ncdc.noaa.gov/data-access/satellite-data/satellite-data-access-datasets
https://www.ncdc.noaa.gov/data-access/satellite-data/satellite-data-access-datasets
Order form for data: https://www.ncdc.noaa.gov/has/has.dsselect
https://www.ncdc.noaa.gov/doclib/index.php?choice=dsi&searchstring=3635&submitted=1&submitted=Search

File Format

Data Filtering and Processing

Final Variable(s)

Methods

1. Download Data

   Navigate to NCEI's Archive Information Request System (AIRS). Scroll down and click on 'Satellite' to expand menu. Click on 'Goes Products' to expand menu. Click on 'Order Data'. Select GOES-West for satellite ID, GASP-AOD-GZ for data type, and appropriate start and end date. Select "Yes" for Submit Batch. Enter email address and submit order. You will get emails later on with FTP links to your data. Run ‘Generic_FTP_download_to_S3.py‘ on an EC2 instance passing in the FTP url as the argument. This will download the data and upload it to S3 (and then delete it off the EC2 instance).

Quality Control

Script Names

1. Generic_FTP_download_to_S3.py

Data File Names

1.

5.14 MERRA-2

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

Notes

File Formats

Data Filtering and Processing

Final Variable(s)

Methods

1.

2.

Quality Control

Script Names

1.

Original Data File Names

1.

2.

Processed/Cleaned Data File Names

1.

2.

5.15 MAIAC

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

Notes

File Format

Data Filtering and Processing

Final Variable(s)

Methods

1.

Quality Control

Script Names

1. Contacted NASA DeepBlue team via email and was given the FTP site for their research data output. Public data set not yet available. But should be in several months under the name ‘MCD19‘.

Data File Names

1. n/a

5.16 MODIS Thermal Anomalies/Fire Daily L3 Global 1km (MCD14DL)

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

We will collect data about fire detection locations, size, and fire radiative power from the MODIS Thermal Anomalies/Fire Daily L3 Global 1km (MOD14 and MYD14) (Giglio et al., 2006; Hawbaker et al., 2017). Using GIS techniques, we will create daily clusters of fire points and use these to calculate: (1) the distance to the nearest fire cluster by day and (2) the sum of Fire Radiative Power (FRP) of the nearest clusters of fires by day as it is likely that smoke levels are higher closer to fires. The MODIS product spans longer than our study period (2008-2014) at daily temporal resolution and has a spatial resolution of 1 km.

Notes

File Format

.shp

Data Filtering and Processing

Final Variable(s)

Methods

1. Navigate to the NASA EarthData FIRMS Archive Download site
2. Select "Create new Request"
3. In the dropdown for region, select "Custom Region" and draw a bounding box around study area
4. In the dropdown for fire data source, select "MODIS C6"
5. Select dates for study time period
6. In the dropdown for file type, select "Shapefile (.shp)"
7. Enter your email address
8. You will get an email with a download link containing a zipfile with the data

Quality Control

Script Names

1. n/a

Data File Names

1. n/a

5.17 Landsat-derived burned area essential climate variable (BAECV) fire activity data

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

We will collect data about fire detection locations, size, and fire radiative power from the Landsat-derived burned area essential climate variable (BAECV) fire activity data, (LP DAAC, 2017). Using GIS techniques, we will create daily clusters of fire points and use these to calculate: (1) the distance to the nearest fire cluster by day and (2) the sum of Fire Radiative Power (FRP) of the nearest clusters of fires by day as it is likely that smoke levels are higher closer to fires. The BAECV can detect fires larger than 4 km^2 and provides an estimate of the date of the fire and is available from 1984-2015.

Notes

File Format

.shp

Data Filtering and Processing

Final Variable(s)

Methods

1. BAECV data set already downloaded by EarthLab fire group. Navigate to the 'earthlab-ls-fire' S3 bucket, then the v1.1 subdirectory. Here you will find yearly .tar.gz files. Have not spent time decompressing files and exploring data yet but my guess is that within each yearly file, we will find more detailed, daily burn data.

Quality Control

Script Names

1. n/a

Data File Names

1. n/a

5.18 MODIS/Terra and Aqua Burned Area Monthly L3 Global 500 m SIN Grid V006 (MCD64A1)

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

We will collect data about fire detection locations, size, and fire radiative power from MODIS/Terra and Aqua Burned Area Monthly L3 Global 500 m SIN Grid V006 (MCD64A1) (Schroeder et al., 2014). Using GIS techniques, we will create daily clusters of fire points and use these to calculate: (1) the distance to the nearest fire cluster by day and (2) the sum of Fire Radiative Power (FRP) of the nearest clusters of fires by day as it is likely that smoke levels are higher closer to fires.

Notes

File Format

.hdf

Data Filtering and Processing

Final Variable(s)

Methods

1. Run script 'MODIS_FTP_download.py' and pass two arguments: the first is the data set name and the second is the local directory path to save files to (i.e. "MCD64A1" "C:/Users/User/MCD64A1_
   Update: 'MODIS_FTP_download.py' is obsolete because NASA LAADS decomissioned their FTP site in favor of HTTPS. So, a new all-purpose script will need to be written to do this download that does HTTPS retrievals instead.

Quality Control

Script Names

1. MODIS_FTP_Download.py

Data File Names

1.

5.19 Visible Infrared Imaging Radiometer Suite (VIIRS) (VNP14IMGTDL_NRT)

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

We will collect data about fire detection locations, size, and fire radiative power from the Visible Infrared Imaging Radiometer Suite (VIIRS) (VNP14IMGTDL_NRT) (Schroeder et al., 2014). Using GIS techniques, we will create daily clusters of fire points and use these to calculate: (1) the distance to the nearest fire cluster by day and (2) the sum of Fire Radiative Power (FRP) of the nearest clusters of fires by day as it is likely that smoke levels are higher closer to fires. The MODIS product spans longer than our study period (2008-2014) at daily temporal resolution and has a spatial resolution of 1 km. VIIRS was launched in 2011 and has 12 h temporal resolution with 750 m resolution. The BAECV can detect fires larger than 4 km^2 and provides an estimate of the date of the fire and is available from 1984-2015.

Notes

File Format

.csv

Data Filtering and Processing

Final Variable(s)

Methods

1. Navigate to the NASA EarthData FIRMS Archive Download site
2. Select "Create new Request"
3. In the dropdown for region, select "Custom Region" and draw a bounding box around study area
4. In the dropdown for fire data source, select "VIIRS"
5. Select dates for study time period
6. In the dropdown for file type, select "Shapefile (.shp)"
7. Enter your email address
8. You will get an email with a download link containing a zipfile with the data

Quality Control

Script Names

1. n/a

Data File Names

1. fire_archive_V1_2770.csv

5.20 Classified land cover information from the Landsat-derived NLCD 2011

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

Classified land cover information from the Landsat-derived NLCD 2011 (Homer et al., 2017) will be used to calculate estimates of the percentage of urban development (codes 22, 23, and 24), agriculture (codes 81 and 82), and vegetated area other than agricultural land (codes 21, 41, 42, 43, 52, and 71) within buffer radii of 100 m, 250 m, 500 m, and 1000 m around each monitor. The buffer distance that is most highly correlated with $PM_{2.5}$ will be entered into each model. NLCD 2011 has a spatial resolution of 30 m and uses circa 2011 Landsat satellite data.

Notes

File Format

.shp

Data Filtering and Processing

Final Variable(s)

Methods

1. Navigate to the National Map Viewer and find products for "National Land Cover Database (NLCD)" at the National extent. From the search results, download "NLCD 2011 Land Cover (2011 Edition, amended 2014)". This will download a zipfile with the data.

Quality Control

Script Names

1. n/a

Data File Names

1.

5.21 MODIS Snow Cover Daily L3 Global 500m Grid, Version 6 (MOD10A1 and MYD10A1)

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

We will use snow cover data from the MODIS Snow Cover Daily L3 Global 500m Grid, Version 6 (MOD10A1 and MYD10A1) (Hall and Riggs, 2016) because snow coverage is a known contributor to wintertime $PM_{2.5}$ concentrations in mountain valleys (Whiteman et al., 2014). Daily MOD10A1 and MYD10A1 data are available since 2002 and have 500 m spatial resolution.

Notes

File Format

Data Filtering and Processing

Final Variable(s)

Methods

1. Step 1: Download the MODIS AOD data sets from both Terra and Aqua sensors:

   Using the NASA EarthData online search tool, search for the 'MOD10A1' (Terra) data set. Set temporal extent by drawing polygon and set spatial extent by adjusting the appropriate filter on the web interface. Select the collection and proceed to download data. For data download options, specify "Stage for Delivery" through the "FTPPull" distribution option. Specify the email address for orders to be sent to. Orders will be sent to your email with instructions on how to connect to the FTP server and pull the ordered data into your local workspace through the command line. Because the amount of data being requested is large, the orders will come through several separate emails. Repeat this step for the 'MYD10A1' (Aqua) data set. All of the raw downloaded data from this step will be in .hdf file format.

Quality Control

Script Names

1.

Data File Names

1.

5.22 Elevation

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

Elevation can influence $PM_{2.5}$ concentrations; for example, $PM_{2.5}$ can accumulate in mountain valleys during persistent cold air pools (commonly referred to as inversions) during winter (Whiteman et al., 2014). We will get elevation data from the 3D Elevation Program, which has resolution of 1 arc-second. This resolution is approximately 10 m north/south and varies east/west with latitude (USGS, 2017).

Notes

File Format

Data Filtering and Processing

Final Variable(s)

Methods

1. Navigate to the National Map Viewer site and find products for Elevation Products (3DEP), 1 arc-second DEM, IMG file format. Once results are returned, select "Save as Text", which will download a text file containing server links to each NED tile.

2. Download the data using the download_tiles.py script, which will access the text file that you just downloaded.

3. Extract the elevation values using the extract_values_to_points.py script.

Quality Control

Script Names

1. download_tiles.py

2. extract_values_to_points.py

Data File Names

1. n/a

5.23 Meteorological Data

Data Source

North American Mesoscale, Analysis (NAM)

- **Contact**
- **Citation/Link** https://www.ncdc.noaa.gov/data-access/model-data/model-datasets/north-american-meso https://nomads.ncdc.noaa.gov/data/namanl/
- **Geographic Extent** North America
- **Temporal Extent** Available March, 2004 - present with slight delay
- **Acknowledgment**

Brief Description

We will obtain meteorological data from the North American Mesoscale, Analysis (NAM) because it includes all of the standard meteorological variables, including planetary boundary layer height, which has proved to be an important variable for converting AOD to $PM_{2.5}$ (Liu et al., 2005). We will calculate 24-hour averages from 6-hourly data for temperature, relative humidity, sea livel pressure, surface pressure, planetary boundary layer height, dew point temperature, precipitation, snow coverage, and the U and V components of wind speed. NAM has 12 km resolution and is available 2004 onward.

Notes

File Format

Prior to 2018, the files are in *.grb ("grib1") format, while 2018 data is in *.grb2 ("grib2") format.

Resources about this file type:

- rNOMADS is an R package for accessing grb* files. It is mostly geared for grib2 files. https://cran.r-project.org/web/packages/rNOMADS/rNOMADS.pdf

- Explanation of what grib files are: http://www.cpc.ncep.noaa.gov/products/wesley/reading_grib.html,

- wgrib program information: http://www.cpc.ncep.noaa.gov/products/wesley/wgrib.html

Data Filtering and Processing

1. Use the earthlab/r-reidgroup docker image, which has wgrib and wgrib http://www.cpc.ncep.noaa.gov/products/wesley/wgrib.html and wgrib2 http://www.cpc.ncep.noaa.gov/products/wesley/wgrib2/ installed on it.

2. Run "Define_directories.R" before running each of the steps below. (Need to figure out how to automate this.)

3. Process_NAM_data_step1.R reads in Locations_Dates_of_PM25_Obs_DeDuplicate.csv and outputs Locations_Dates_of_PM25_Obs_DeDuplicate_wNextDay.csv, which includes the next day for each location/day listed in the first file. The purpose of this is so all of the necessary NAM files can be processed. UTC dates can go into the next day for western US time zones. This step uses these input files and R packages and functions:

(a) Locations_Dates_of_PM25_Obs_DeDuplicate.csv

(b) add_next_day_date_loc_function.R.

4. Process_NAM_data_step2_parallel.R downloads each NAM file, extracts relevant data, and deletes the original NAM data. (All of the NAM files together would be about 1.6 Tb.) This file operates in parallel, and will use n-1 cores, where n is the number of cores on the computer. The output is 1 csv with all locations of interest for a given date and time step. The time steps for the NAM are 0Z, 6Z, 12Z, and 18Z. The output files have the format Locations_Dates_of_PM25_Obs_DeDuplicate_YYYY_MM_DD_XXUTC.csv where XX refers the to the timestep. Change the study start and stop dates for the dates to be processed. This step uses these input files and R packages and functions:

   (a) Locations_Dates_of_PM25_Obs_DeDuplicate.csv - Data file with dates (local) and locations where you want the NAM data

   (b) MeteoVariablesNAM.csv - listing of meteorological variables to be extracted from NAM data

   (c) rNOMADS R package (which calls wgrib and wgrib2) https://cran.r-project.org/web/packages/rNOMADS/rNOMADS.pdf

   (d) parallel R package

   (e) grb1to2_conversion_prep_function.R - This script downloads the files that will be necessary to run grb1to2.pl, created by the Climate Prediction Center http://www.cpc.ncep.noaa.gov/products/wesley/grb1to2.html

   (f) loop_NAM_run_times.parallel_function.R - this function loops through the time steps on a given day and calls function (listed below) to extract meteo data at locations of interest

   (g) define_project_bounds_function.R - the bounding box for the study area is defined in this function. The scripts can run faster if the entire NAM domain does not need to be loaded into memory.

   (h) extract_NAM_data_parallel_function.R - this function extracts the NAM data at points of interest

   (i) which_type_of_grib_file_function.R - this function determines whether the data for a given time step are grib1 or grib2 format

   (j) convert_grib1to2_function.R - convert file type from grib1 to grib2, unless it's already a grib2 file. This is essentially a wrapper for grb1to2.pl created by the Climate Prediction Center http://www.cpc.ncep.noaa.gov/products/wesley/grb1to2.html

5. Process_NAM_data_step3.R merges all of the files from step 2 into a single file and adds a column at the beginning giving the UTC time stamp.

6. Process_NAM_data_step4.R To Do: deal with time zones and compiling 24-hr summaries

7. To Do: Merge_NAM_times.R will merge the 4 time steps to give a 24-hr summary. Min, max, mean, etc. is set in MeteoVariabesNAM.csv.

Final Variable(s)

See MeteoVariablesNAM.csv

Quality Control

5.24 Dust Storms

Data Source

- **Contact**
- **Citation/Link**
- **Data (local)**
- **Geographic Extent**
- **Temporal Extent**
- **Acknowledgment**

Brief Description

Dust storm records will be included in the machine learning algorithm because they can be a significant indicator of airborne particulate matter from sources other than fires. Dust storm records are available from 1993-2017. The spatial resolution varies, but includes either forecast zone or county (US National Weather Service, 2017b,c,a).

Notes

File Format

Data Filtering and Processing

Final Variable(s)

Methods

1.

2.

Quality Control

Script Names

1.

Data File Names

1.

6 Data Sources for CAMx Modeling of Source-Attributed Air Quality Modeling

For meteorological inputs, the CAMx modeling will use archived daily 27-km Advanced Research Weather Research and Forecasting (WRF-ARW) grids available via NOAA Real-time Environmental Applications and Display sYstem (READY) servers for the entire study area and time period (Wang et al., 2007; Rolph et al., 2017). For the study years 2008-2012 and 2014, we will use fire emissions datasets prepared by the Western Regional Air Partnership (WRAP) and the National Emissions Inventory (NEI) (US EPA, 2017b) based on aggregated source-tagged fire occurrence data sources, the FCCS (Ottmar et al., 2007), and Consume (Prichard et al., 2009) modeling. For the study year 2013, we will prepare a fire emissions dataset using the same aggregated source-tagged fire occurrence data sources and FCCS/Consume modeling framework in the NASA-funded Wildland Fire Emissions Information System (WFEIS) (MTRI, 2017) developed by Co-I's French and Billmire (French et al., 2014). Fire occurrence datasets include MODIS (MOD14/MYD14 and MCD64A1) and VIIRS (VNP14IMGTDL_NRT) fire data products (Giglio et al., 2006; LP DAAC, 2017; Schroeder et al., 2014). For non-fire emissions during the entire study period, we will use the dataset prepared by WRAP for year 2008.

Look into using spot forecasts to help distinguish between wild and prescribed fires: http://www.weather.gov/spot/monitor/

7 CAMx Modeling

8 Compiling Data

8.1 Processing PM2.5 data

These are the scripts that process and compile the PM2.5 data:

1. Script1_Install_Pkgs.R » install packages
2. Define_directories.R » clears all variables and defines directories. Needs to be ran between each of the following scripts. (Want to automate this eventually.) When processing a new batch of data, iterate "processed_data_version" by one letter and create a new subfolder in /home/Processed_Data/ named PM25_data_part_* where * is the new processed_data_version.
3. Process_PM25_data_step1.R » compiles the various PM2.5 data sources into a single data frame. The only eliminations of data are geographic, to remove states that are neither in our study area.Update time frame of study if necessary. The output from this script is a csv file and sink .txt for each $PM_{2.5}$ data source as well as a file with all of the PM2.5 data sources merged together ("PM25_Step1_part_*.csv).
   (a) For DRI data, put in flags for voltage data outside the range 11-17 V. (These thresholds are somewhat arbitrary, but it was noticed that when the voltage was outside this range, the $PM_{2.5}$ concentrations were often absurdly high, e.g., greater than 24,000 ug/m3.
4. Process_PM25_data_step2.R » cleans the data. The following is a list of the quality cuts and changes made to the data:
   (a) Replace "UNKNOWN" datum in EPA data with "NAD27" per Colleen's advice.
   (b) Remove negative and NA $PM_{2.5}$ concentrations. This includes removing all data for a monitor on a given day if any of the hourly observations were negative.
   (c) For the hourly data, remove monitor-days that do not have at least 18/24 observations.
   (d) For DRI data, remove data with voltage flags (which includes flags that came with the data and flags that were put in because the battery voltage was outside the range 11-17 V.
   (e) For DRI data, remove data at or below 0 L/min for flow. Think about whether a minimum value of flow should be set (higher than zero).
   (f) June 6, 2014 24-hr average $PM_{2.5}$ concentration from monitor "Smoke NCFS E-BAM #1" (Fire_Cache_Smoke_DRI_Smoke_NCFS_E_BAM_N1.csv) is 24,203 ug/m3. There's nothing apparent wrong with the hourly data, however, this is the only day of data that made it through the other quality checks from this data file. This suggests that this monitor is suspect, and will be removed.
   (g) Remove data points with lat/lon outside this box: (50,-126) to (25,-101)
   (h) **To Do** think about making cuts on any unrealistic air temperatures for DRI data
   (i) **To Do** need to convert missing values that have a -9999 etc to NA value
   (j) **To Do** merge "24-HR BLK AVG" and "24 HOUR" data together in Sample Duration variable
   (k) **To Do** figure out why Observation percent has a max value of 200%
   (l) **To Do** figure out if max AQI value of 546 is reasonable
   (m) **To Do** Some DRI files looked like they had hour 20:00 data shifted a couple of columns - look into this and fix it.
   (n) **To Do** Finish filling in Year, month, day information based on date

(o) **To Do** look over summary() output and plots of every variable and determine if any other cuts are necessary

Process_PM25_data_step3.R » convert all PM2.5 data to the same datum (NAD83) and project coordinates. Take the reprojected location info and put it into the data frame with the daily $PM_{2.5}$ data.

Process_PM25_data_step4.R » combine locations and location/date files for parts b and c. (Part a should be disregarded.) **To Do** write code to take difference between parts bc and d to get a list of just the dates/locations that are new in part d. **To Do** process part d in this file

5. Process_PM25_data_step5.R » composite replicate data - in process. **To Do** finish colocated version of code to go with aves version of code. Calls these functions:
   (a) Combine_true_replicates_R_function.R
   (b) fill_in_aves_coloc_unique_PC_POC_MN_function.R
      i. concatinate_within_column_function.R
   (c) set_data_types_by_column_R_function.R
6. Plot_ML_Input_File.R » create plots, maps, and statistical summary - needs to be changed to take input from De-duplicate code instead of file from Create_ML_Input_File.R
7. **to be written** » merge with satelite and other data

8.1.1 Notes about very high data points

June 15, 2012 24-hr average $PM_{2.5}$ concentration from monitor "Smoke #22" (Fire_Cache_Smoke_DRI_Smok is 5,638 ug/m3 - can't find any reason, so far, to remove this data point, though it's very odd that the concentrations were low single-digits except for two hours which were extremely high (123,000 and 1000 ug/m3).

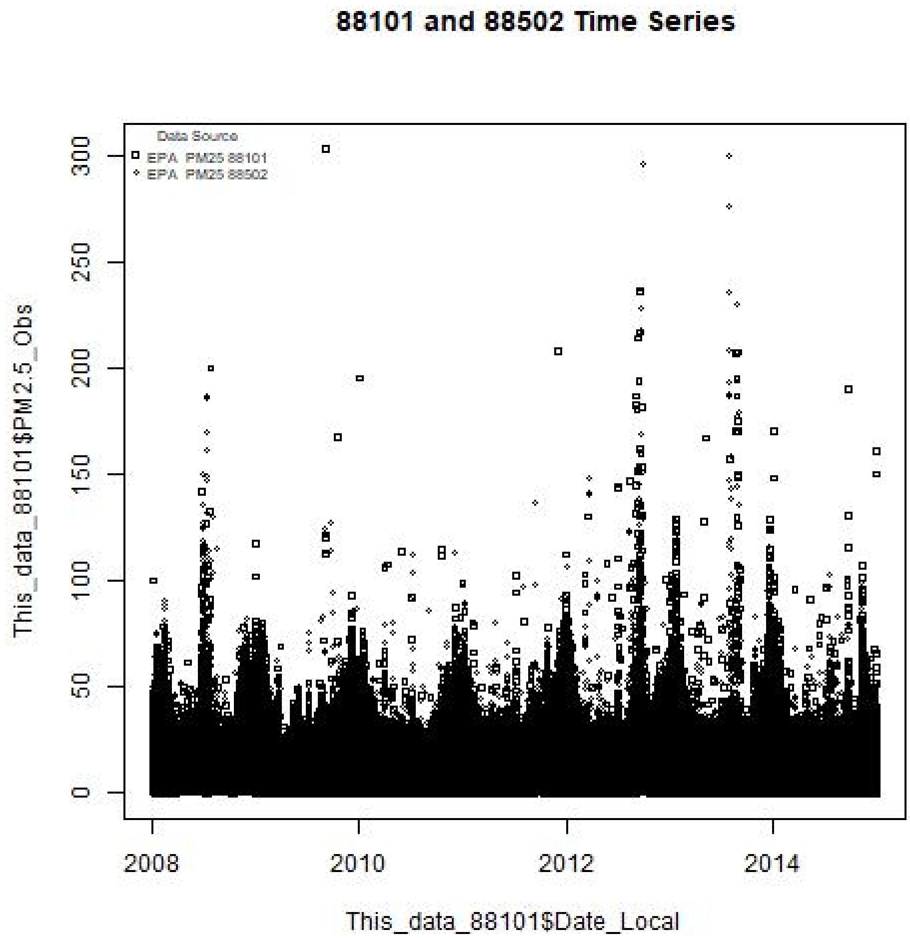

Figure 16: Time series of 88101 and 88502 PM2.5 data.

8.2 Compare 88101 to 88502 PM2.5

9 Machine Learning Methods

setting aside a portion of the PM2.5 data set and then doing 10-fold cross validation on the rest of the data

see http://www.cvent.com/events/nasa-aist-machine-learning-workshop/event-summary-1f5144a5d1734ca aspx and particularly the very end of https://global.gotomeeting.com/public/recording-player.html?id=owZDmUustOjaW9sJGQ5u9cUG2pBa4D for list of resources and papers to read.

9.1 ML Scripts

1. ML_PM25_estimation_step0.R » Merge the various predictor variables together with the monitor data or dates/locations of interest

2. ML_PM25_estimation_step0a.R » Plot the training input file

   (a) predictor variables vs date

   (b) predictor variables vs $PM_{2.5}$

3. ML_PM25_estimation_step1.R » ML training algorithms

4. ML_PM25_estimation_step1.R » create data frame of the dates/locations for which we want to predict PM2.5

10 Machine Learning Results

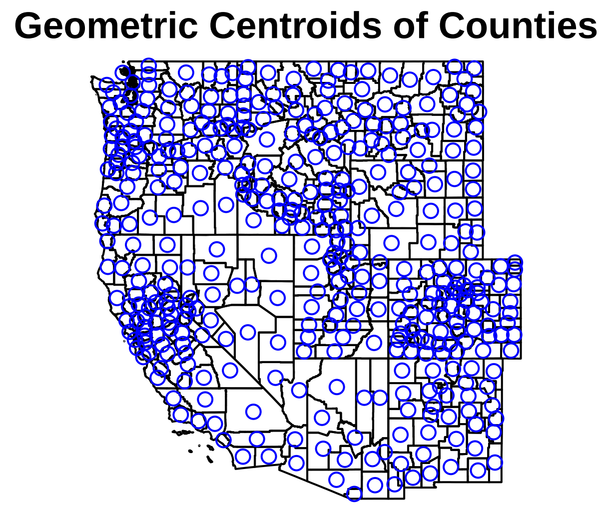

Figure 17: Geometric Centroids of Counties

10.1 Geometric Centroids of Counties Images

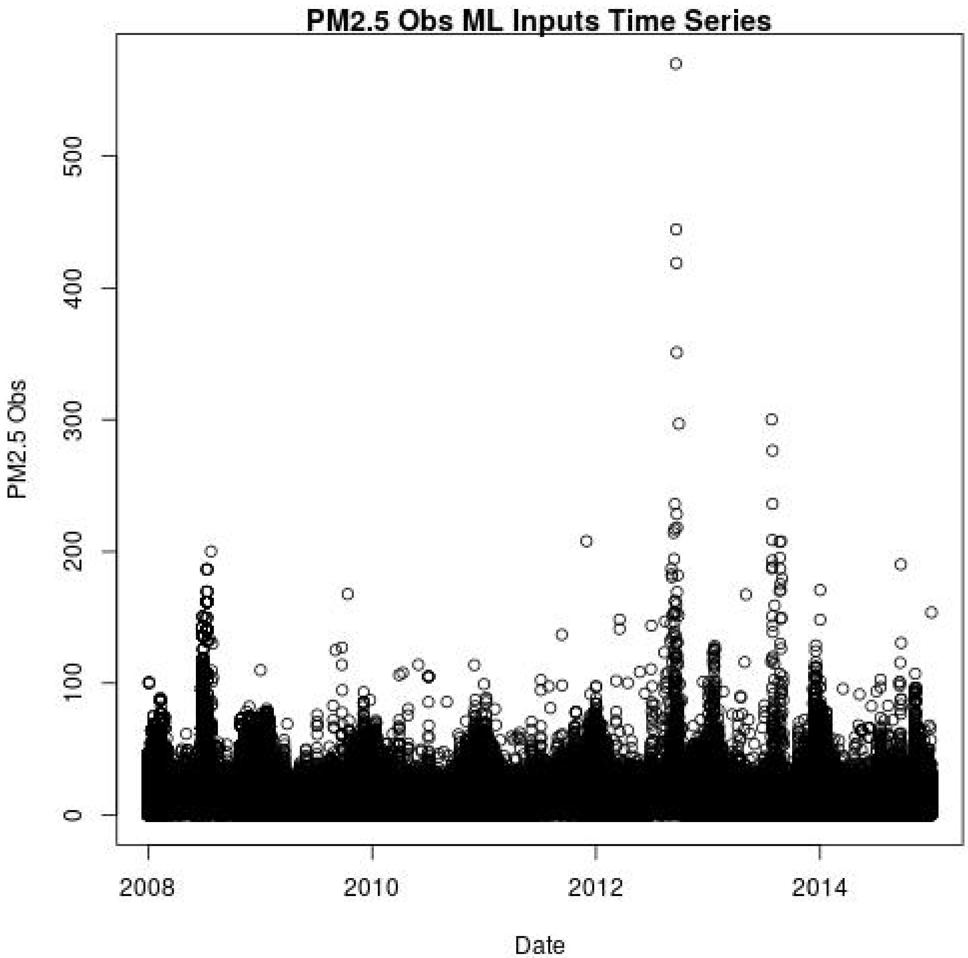

Figure 18: PM2.5 Obs ML Inputs Time Series

10.2 ML Inputs Time Series Images

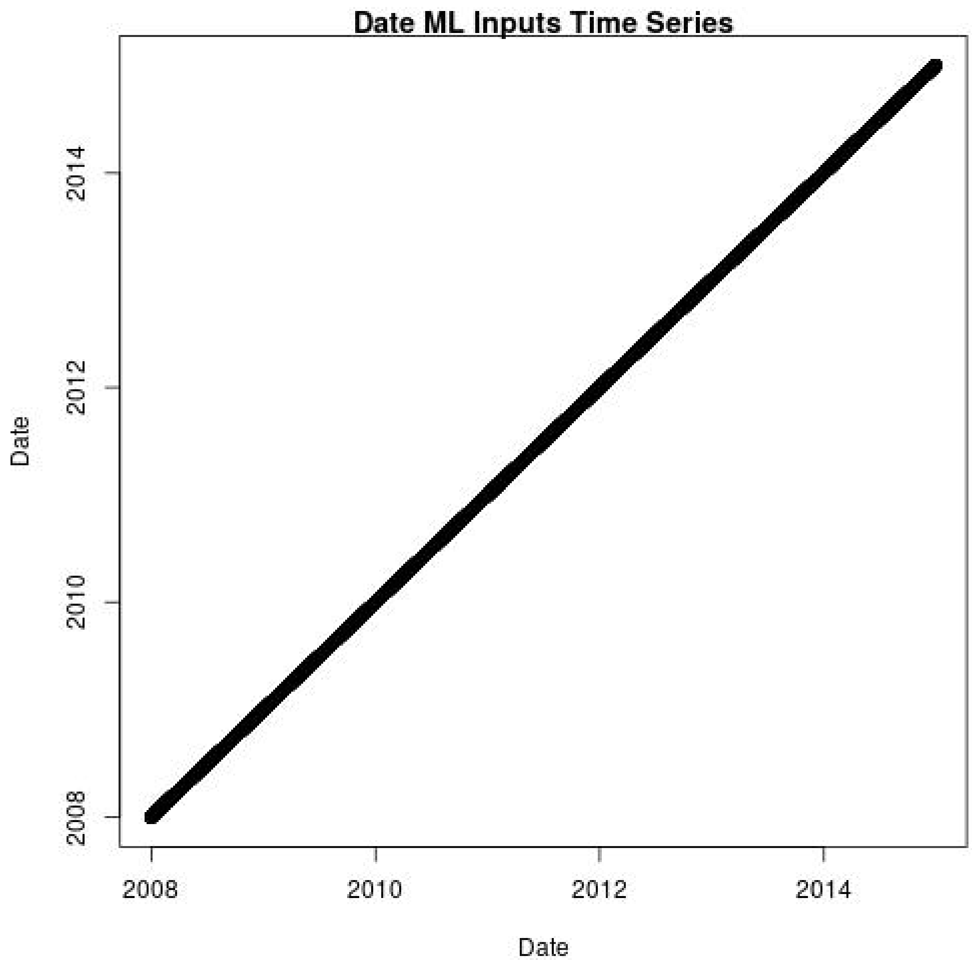

Figure 19: Date ML Inputs Time Series

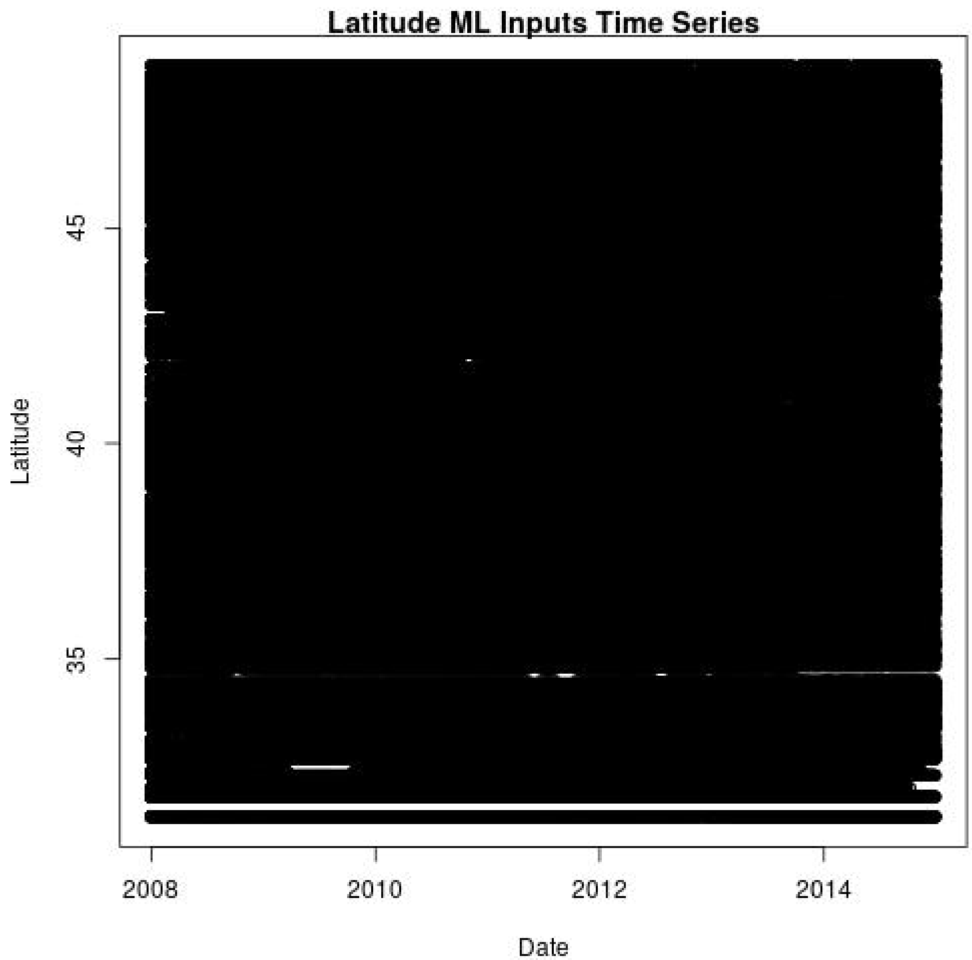

Figure 20: Latitude ML Inputs Time Series

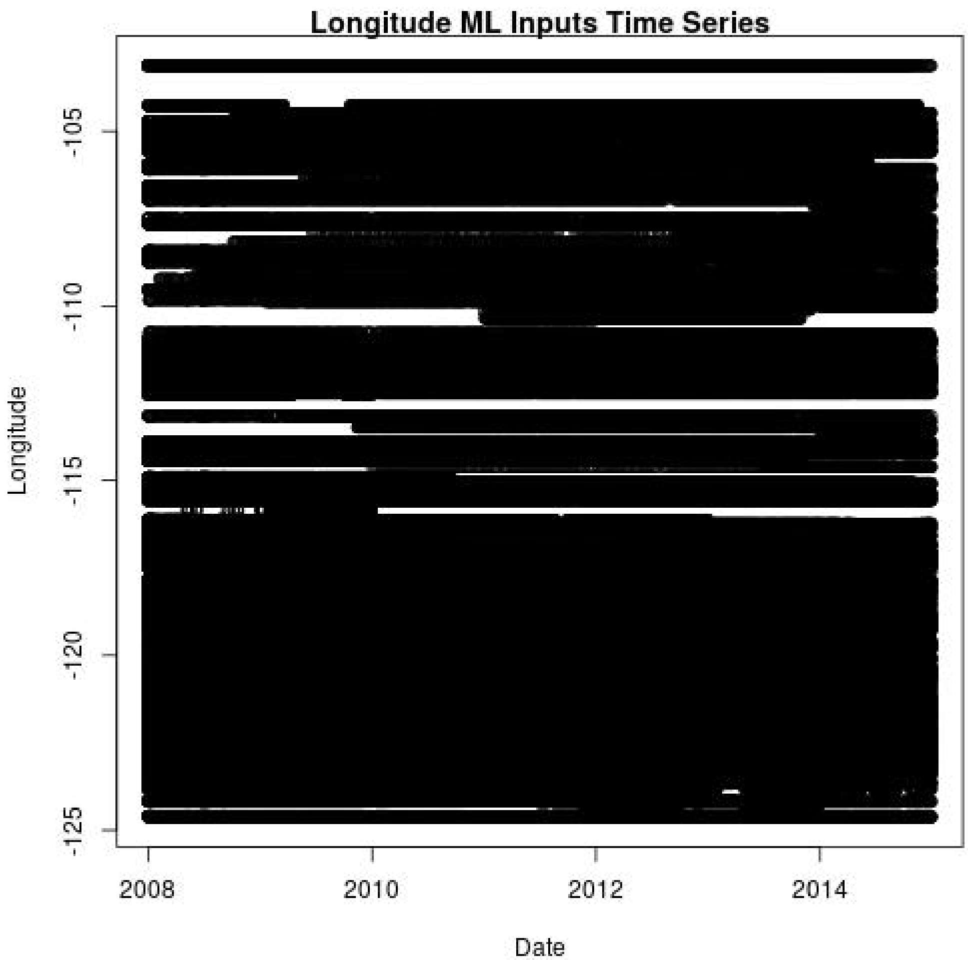

Figure 21: Longitude ML Inputs Time Series

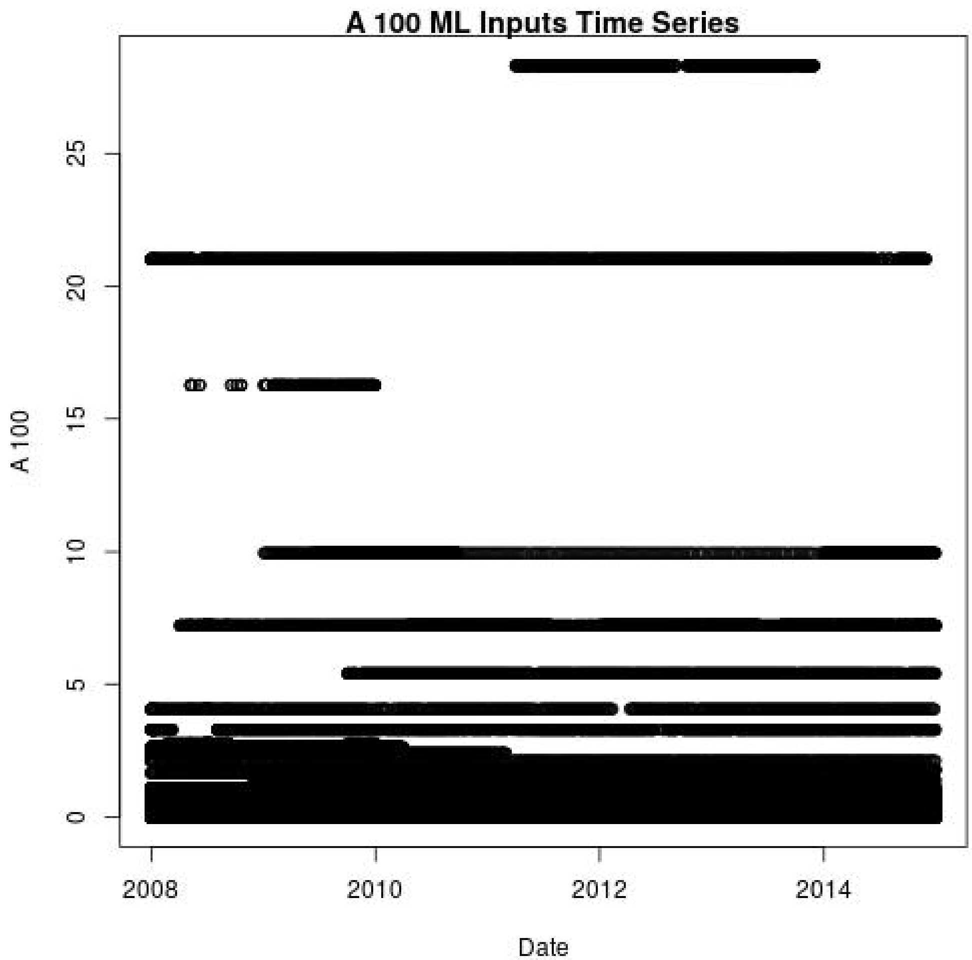

Figure 22: A 100 ML Inputs Time Series

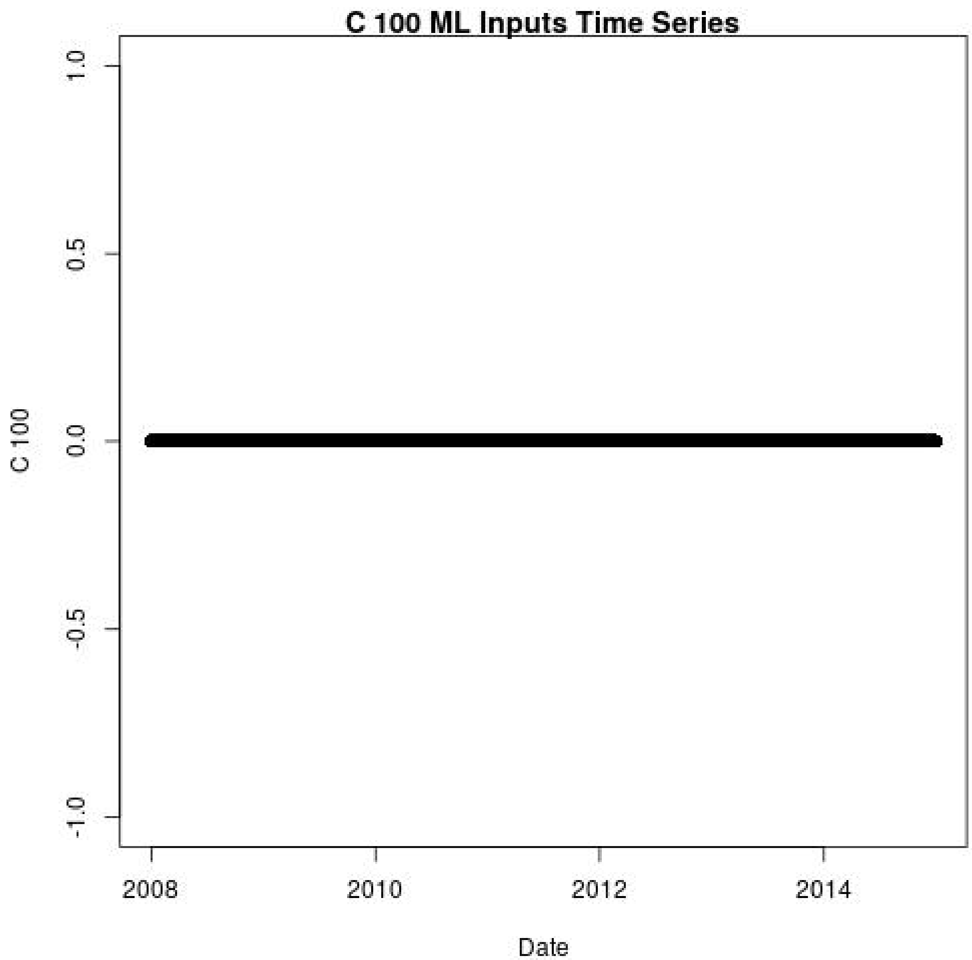

Figure 23: C 100 ML Inputs Time Series

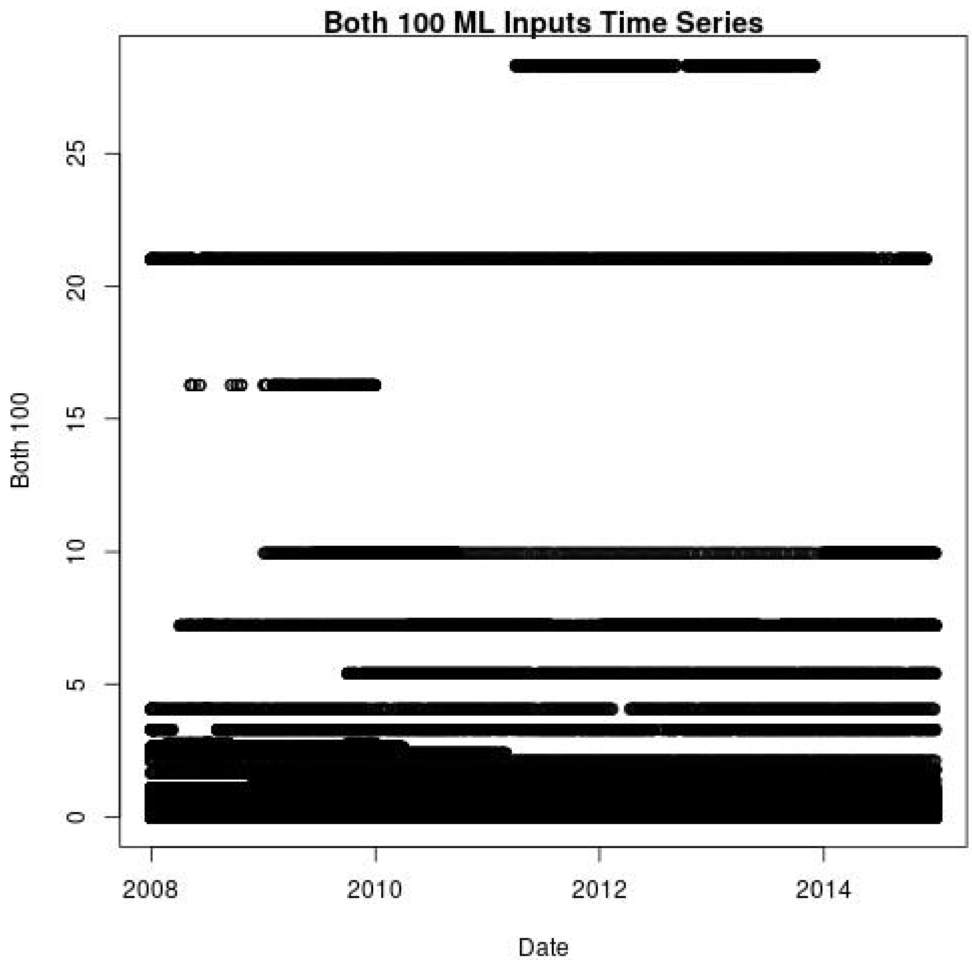

Figure 24: Both 100 ML Inputs Time Series

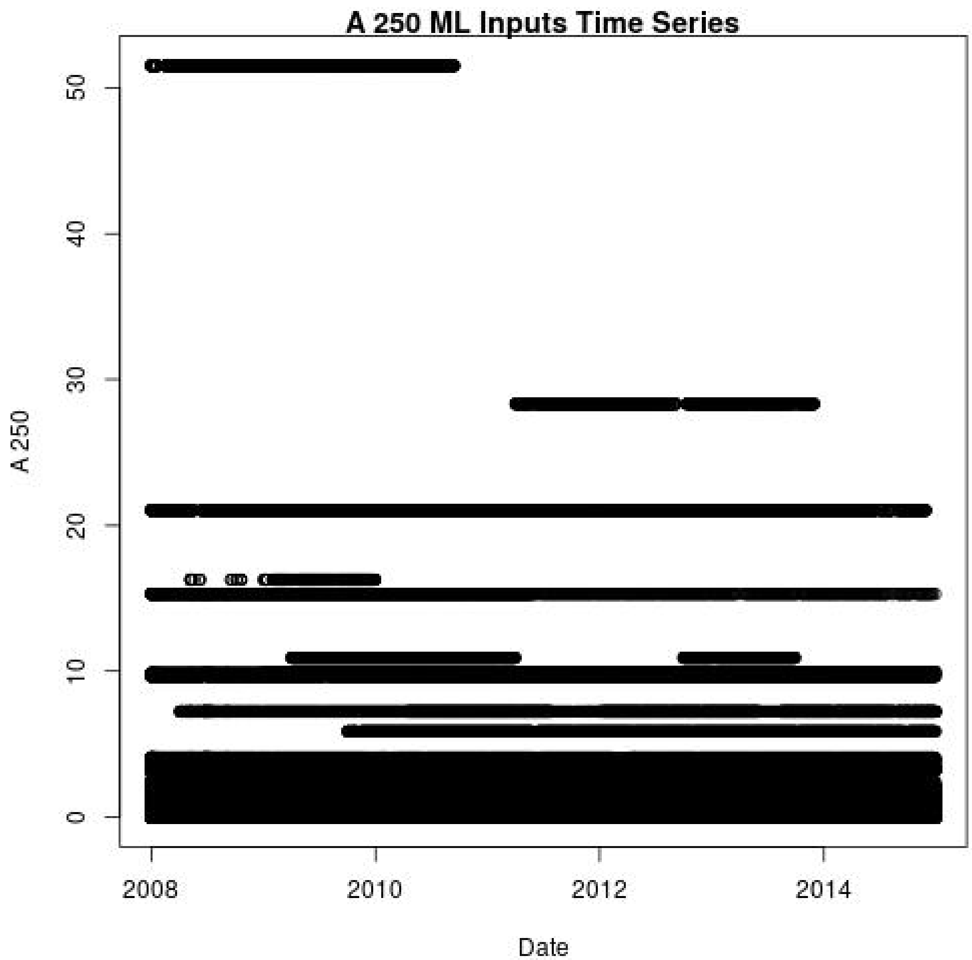

Figure 25: A 250 ML Inputs Time Series

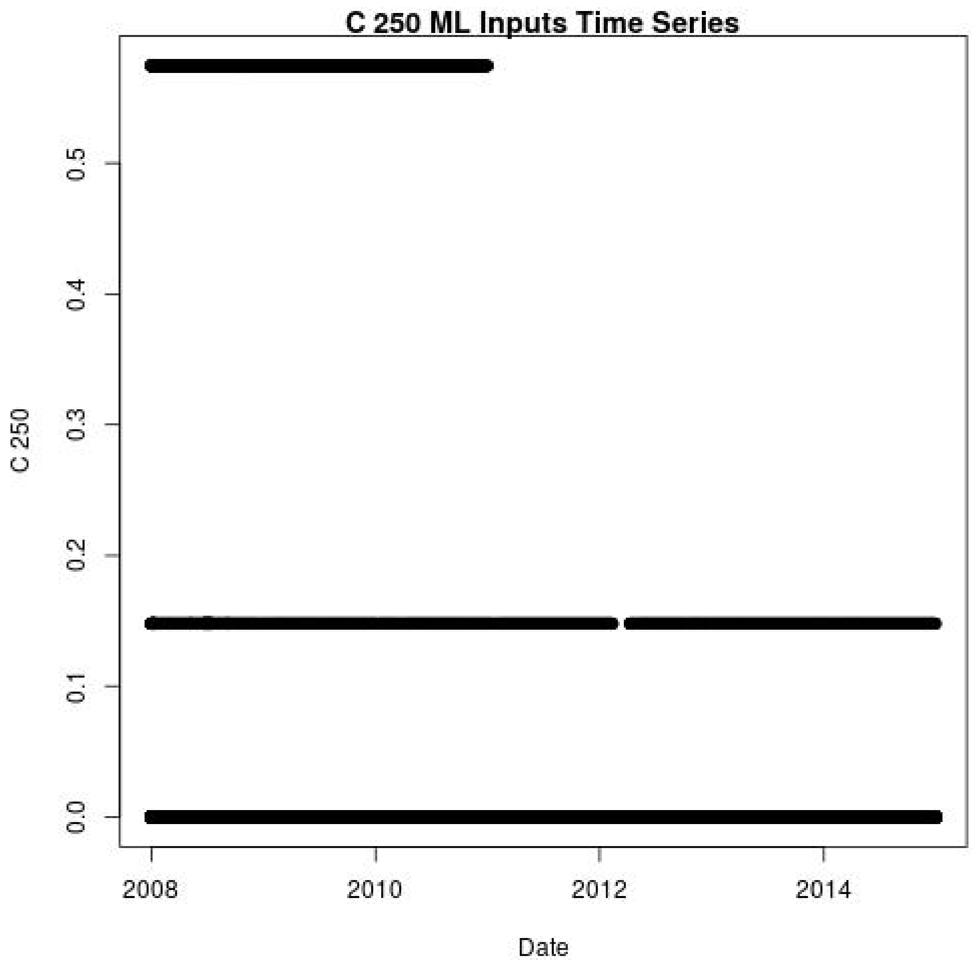

Figure 26: C 250 ML Inputs Time Series

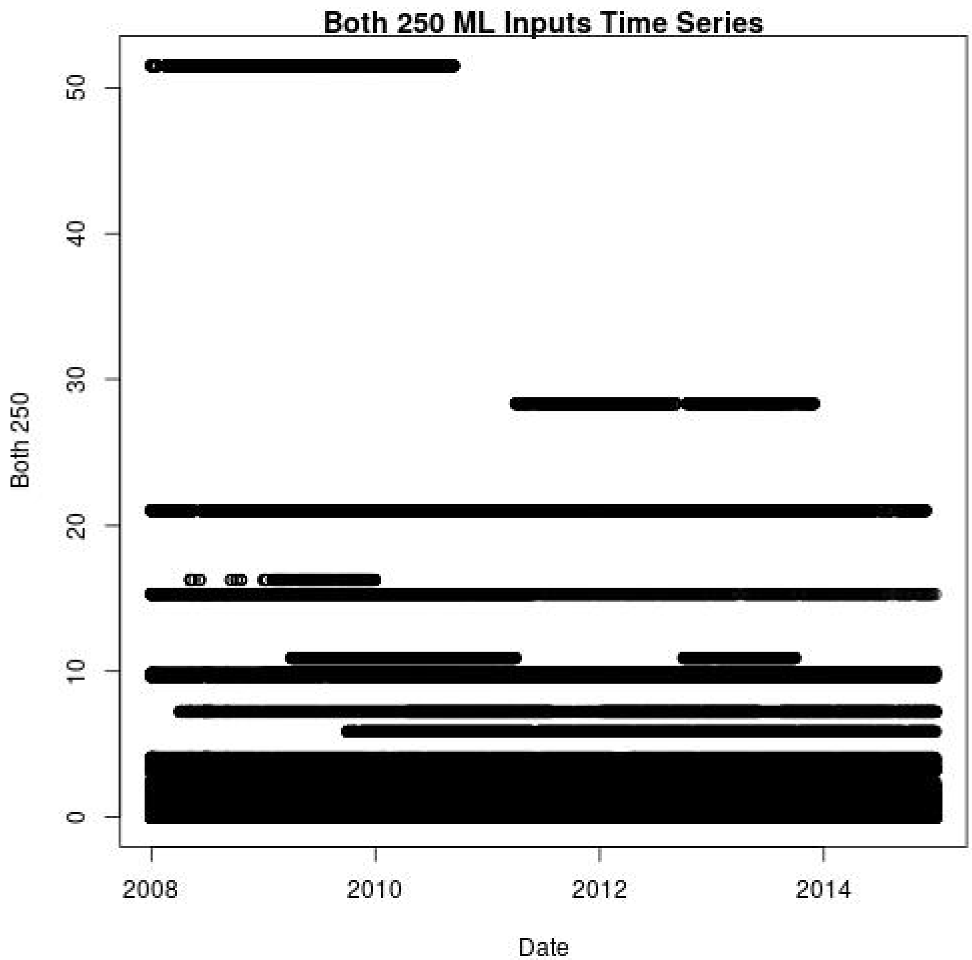

Figure 27: Both 250 ML Inputs Time Series

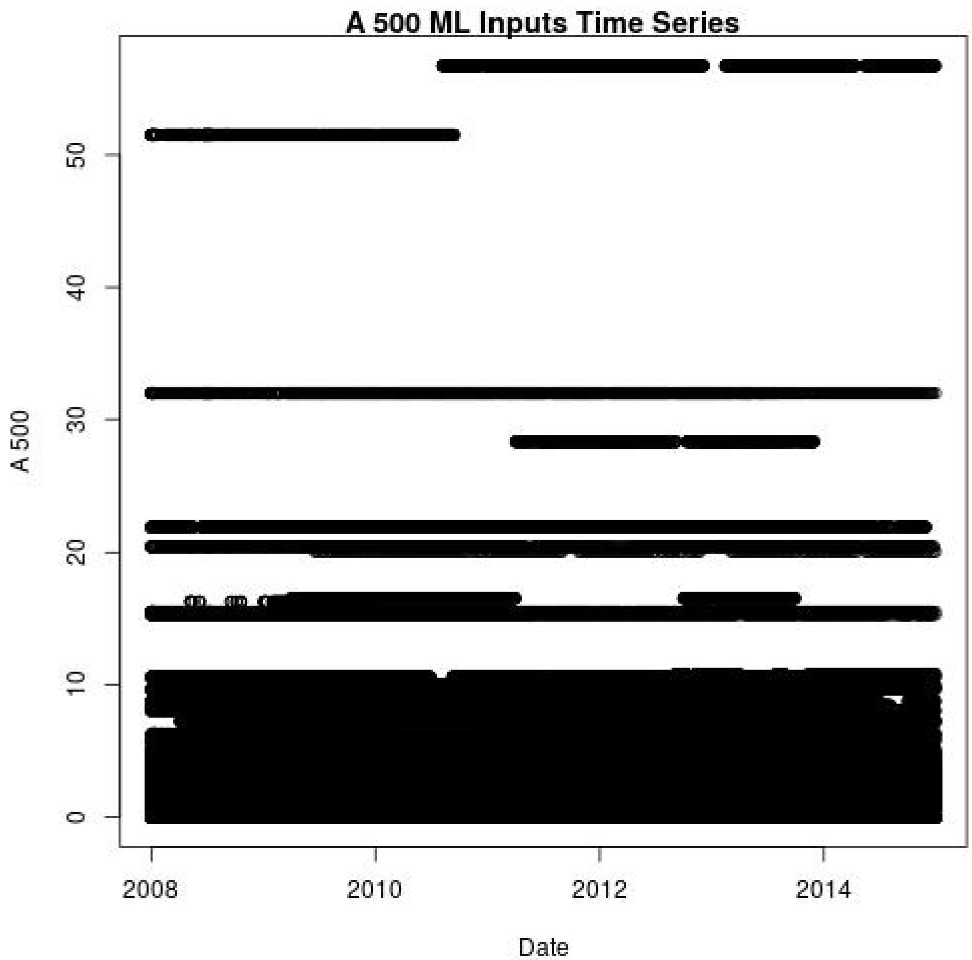

Figure 28: A 500 ML Inputs Time Series

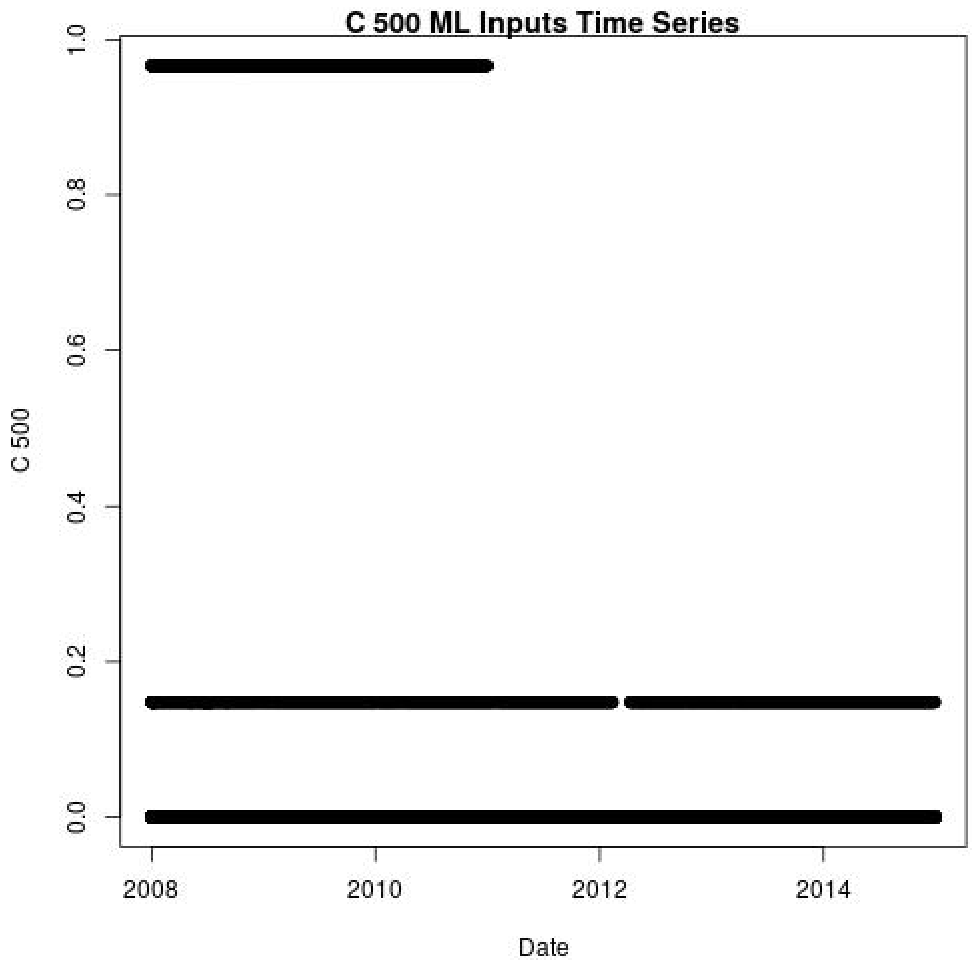

Figure 29: C 500 ML Inputs Time Series

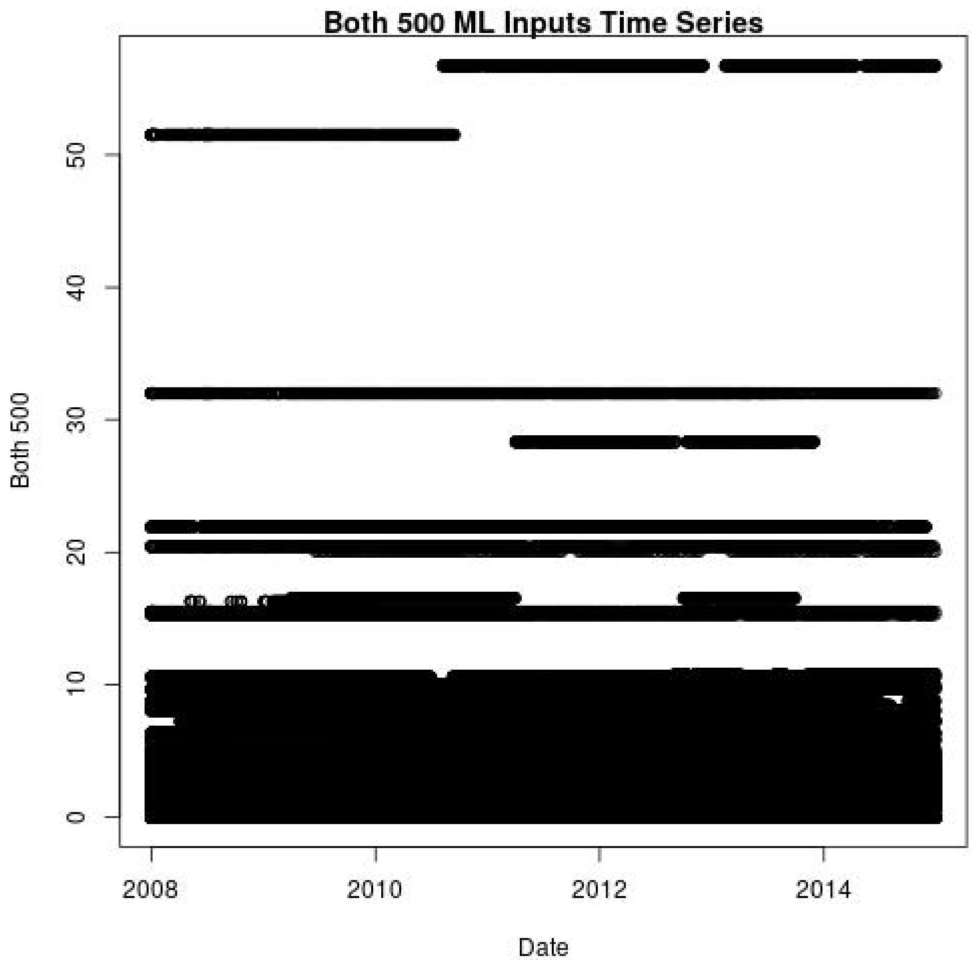

Figure 30: Both 500 ML Inputs Time Series

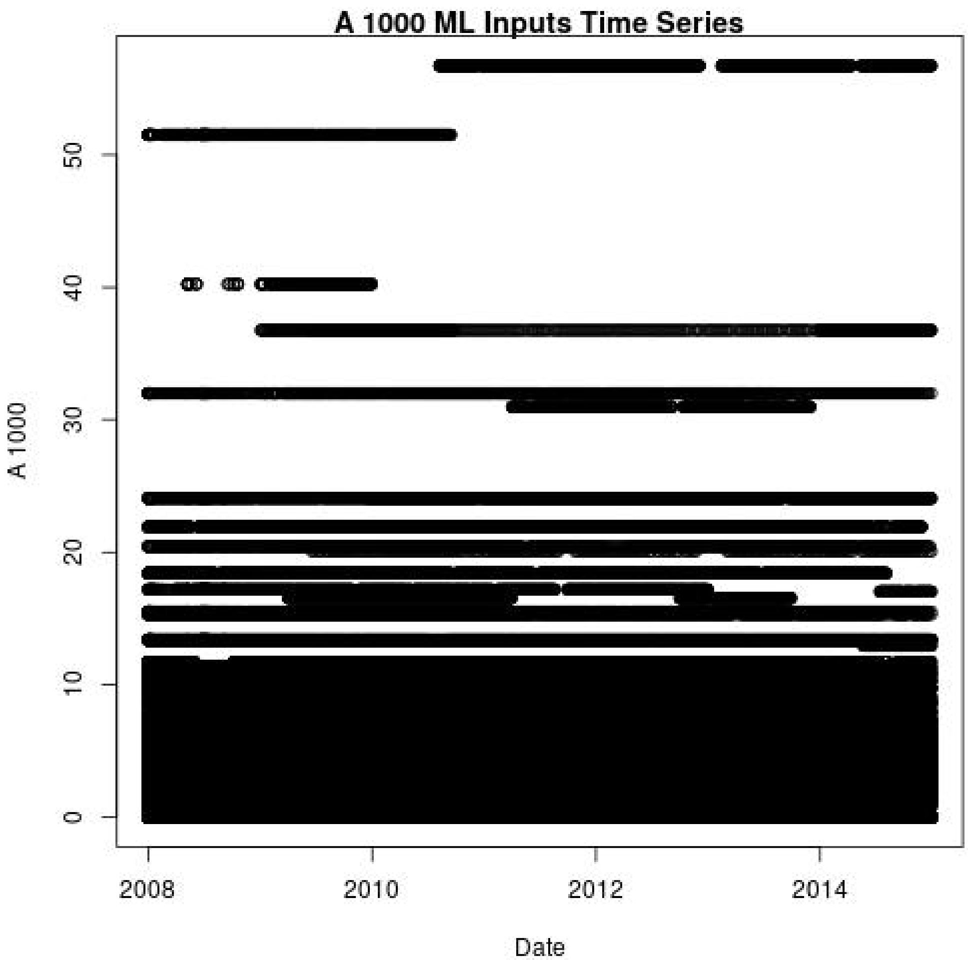

Figure 31: A 1000 ML Inputs Time Series

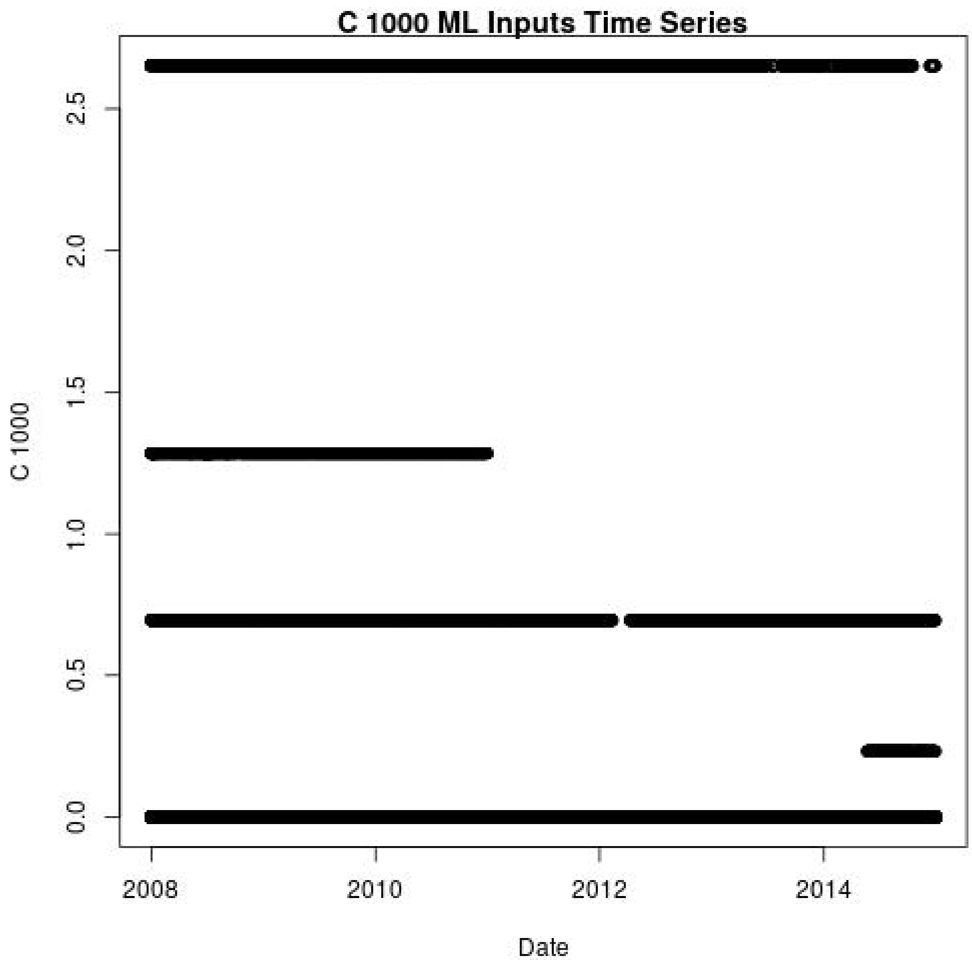

Figure 32: C 1000 ML Inputs Time Series

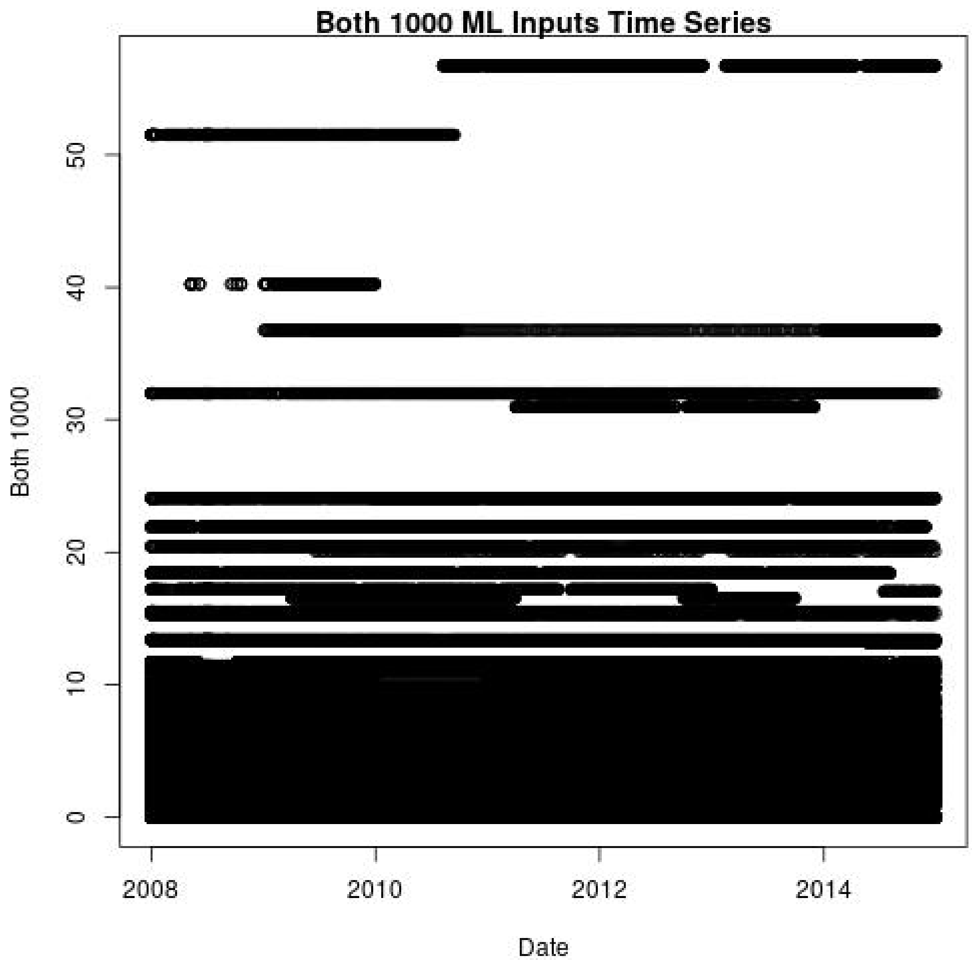

Figure 33: Both 1000 ML Inputs Time Series

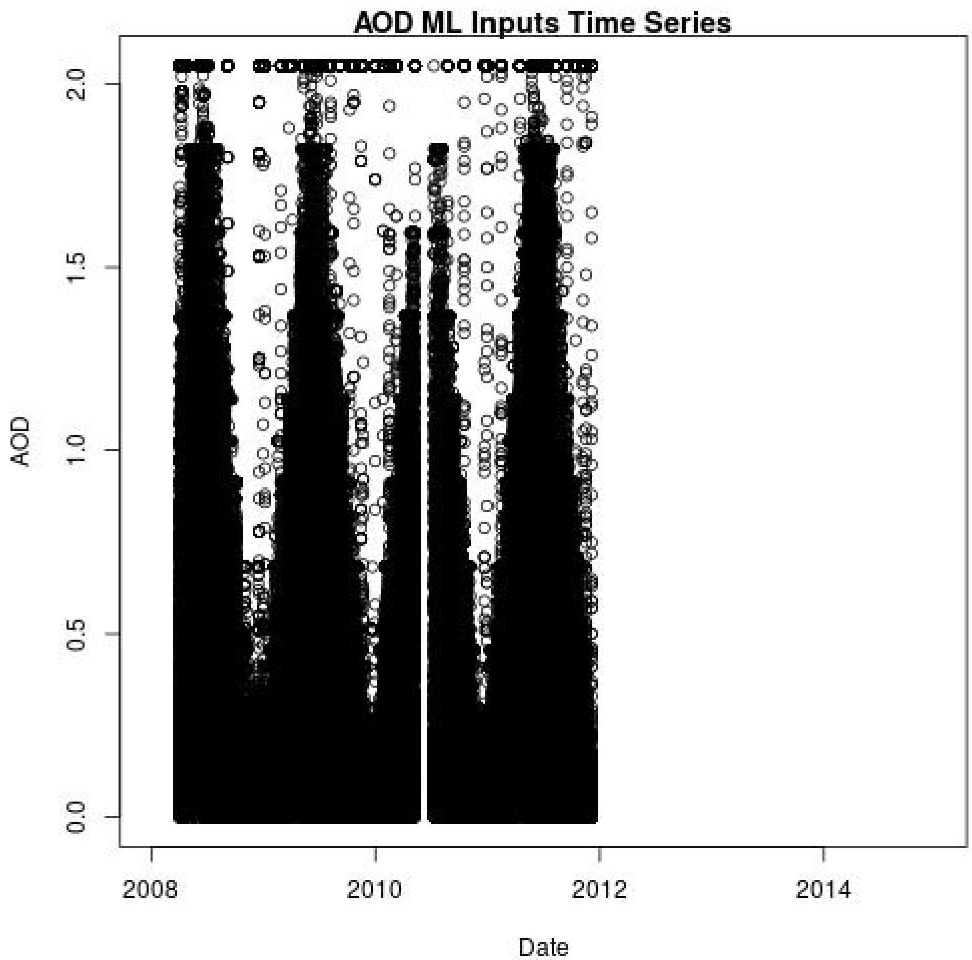

Figure 34: AOD ML Inputs Time Series

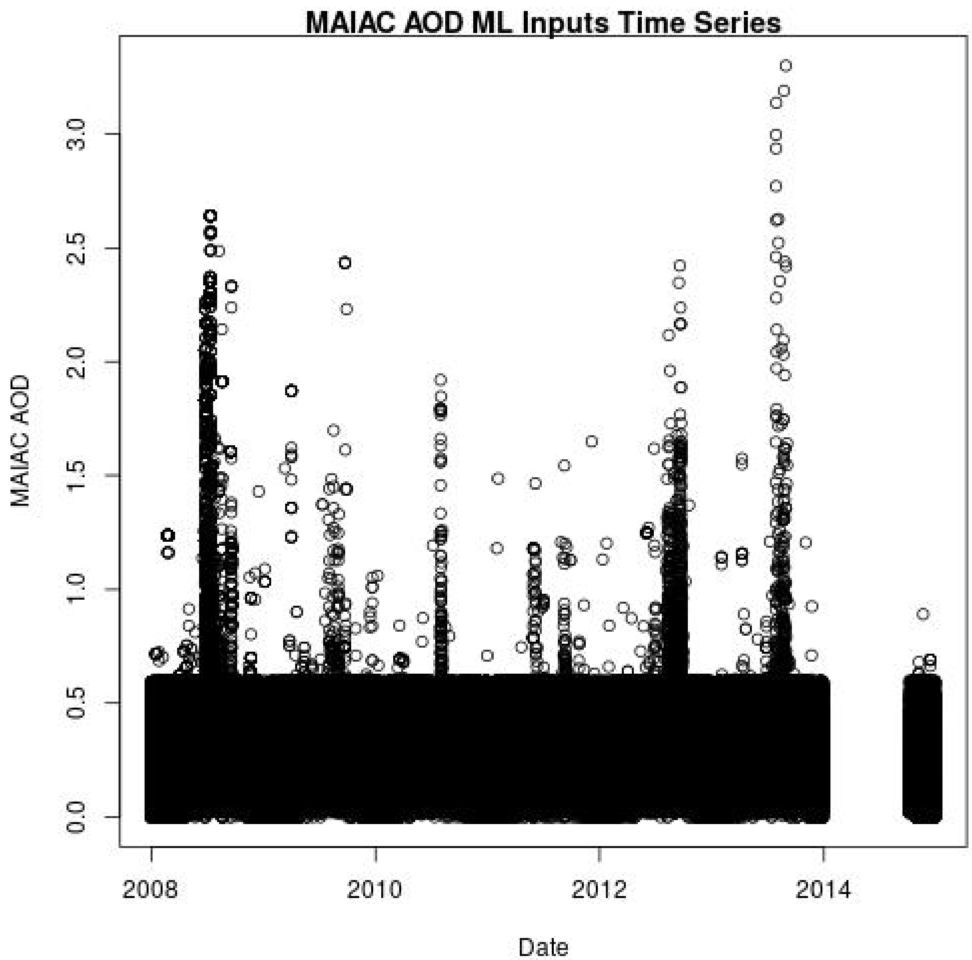

Figure 35: MAIAC AOD ML Inputs Time Series

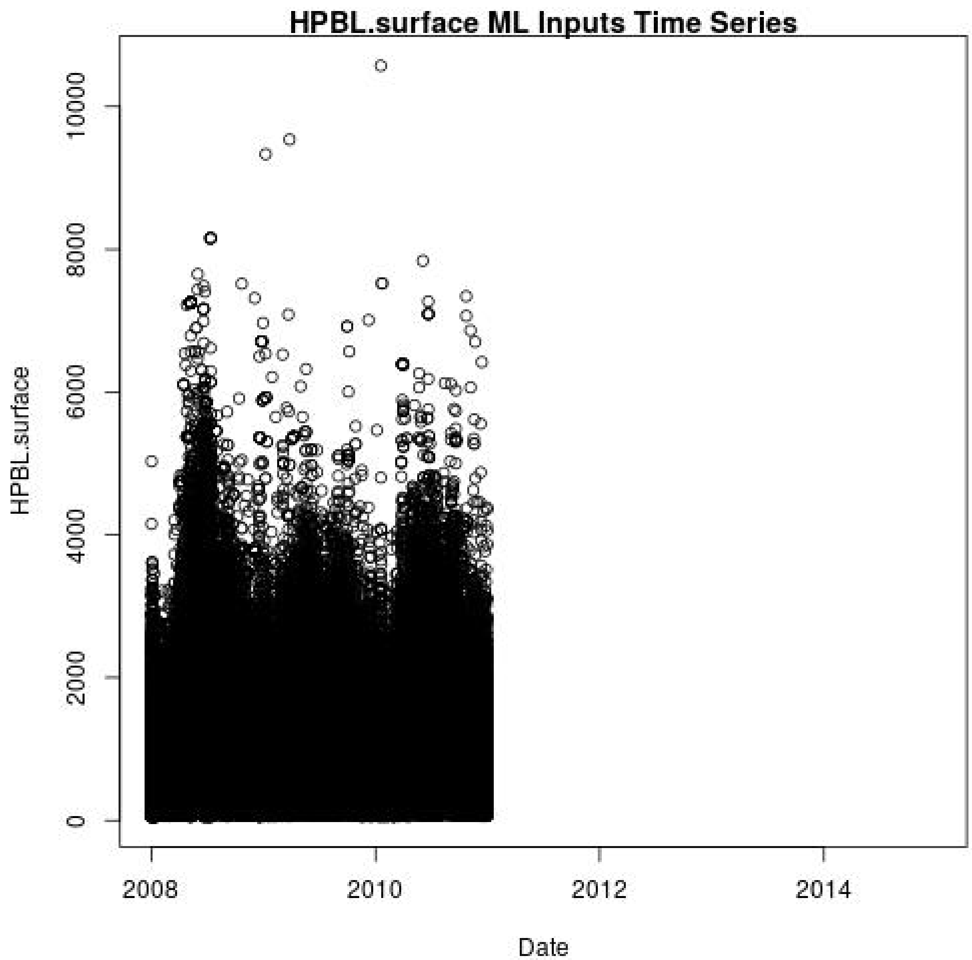

Figure 36: HPBL.surface ML Inputs Time Series

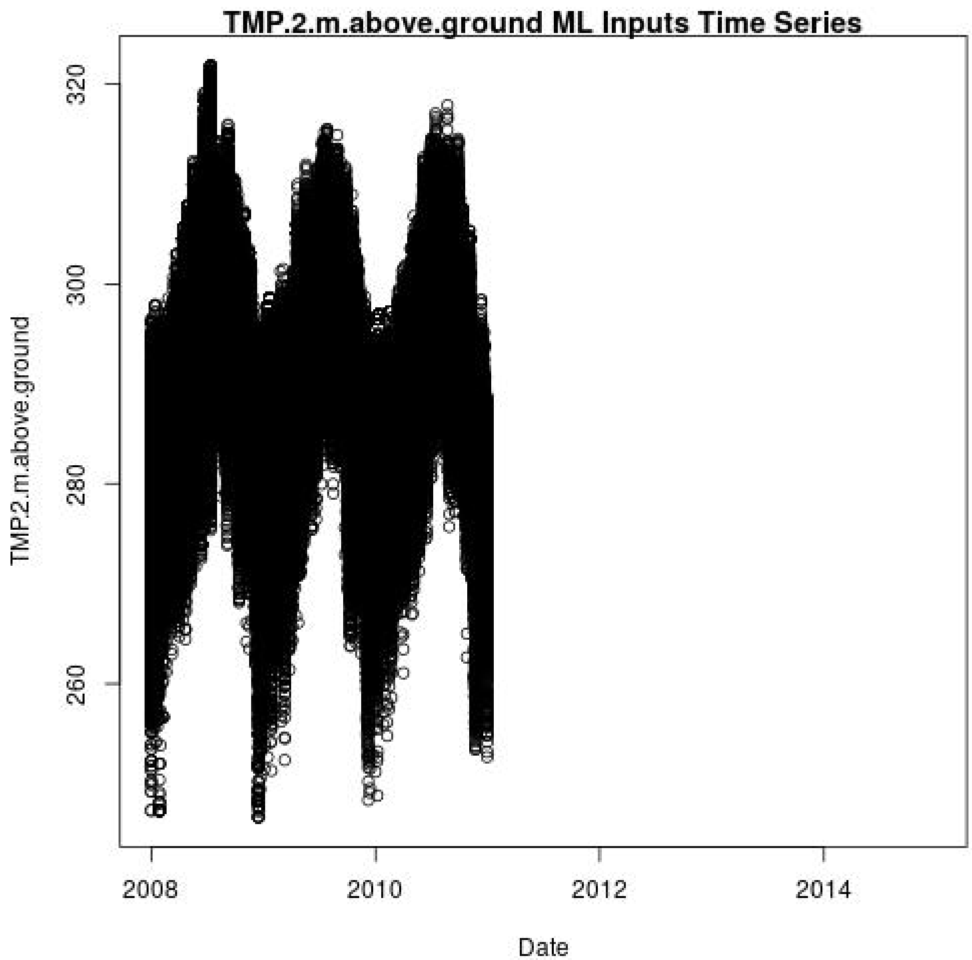

Figure 37: TMP.2.m.above.ground ML Inputs Time Series

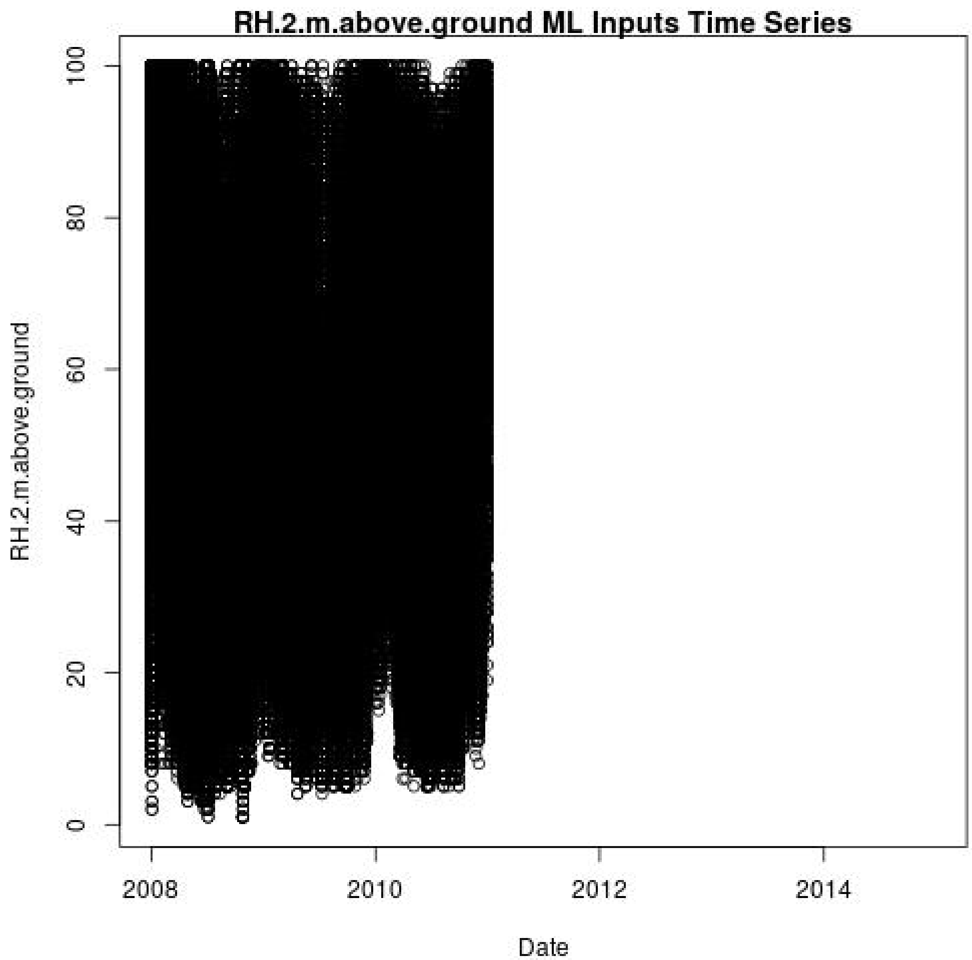

Figure 38: RH.2.m.above.ground ML Inputs Time Series

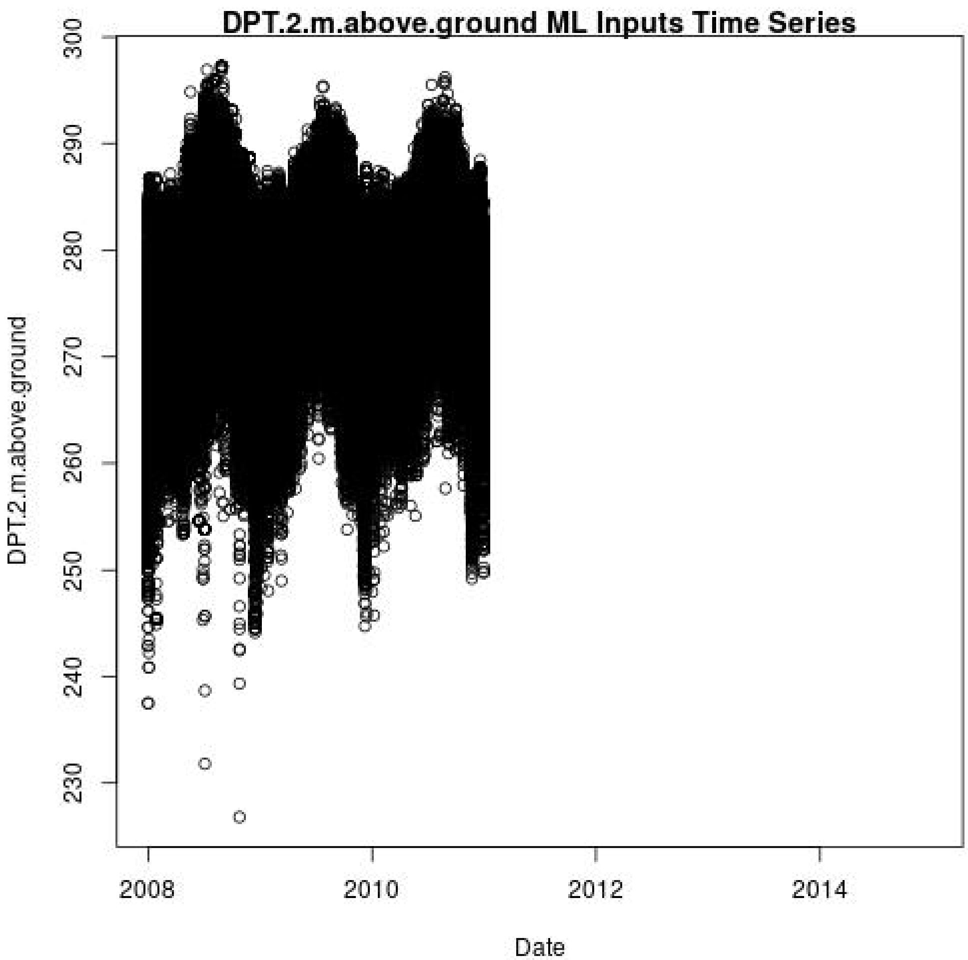

Figure 39: DPT.2.m.above.ground ML Inputs Time Series

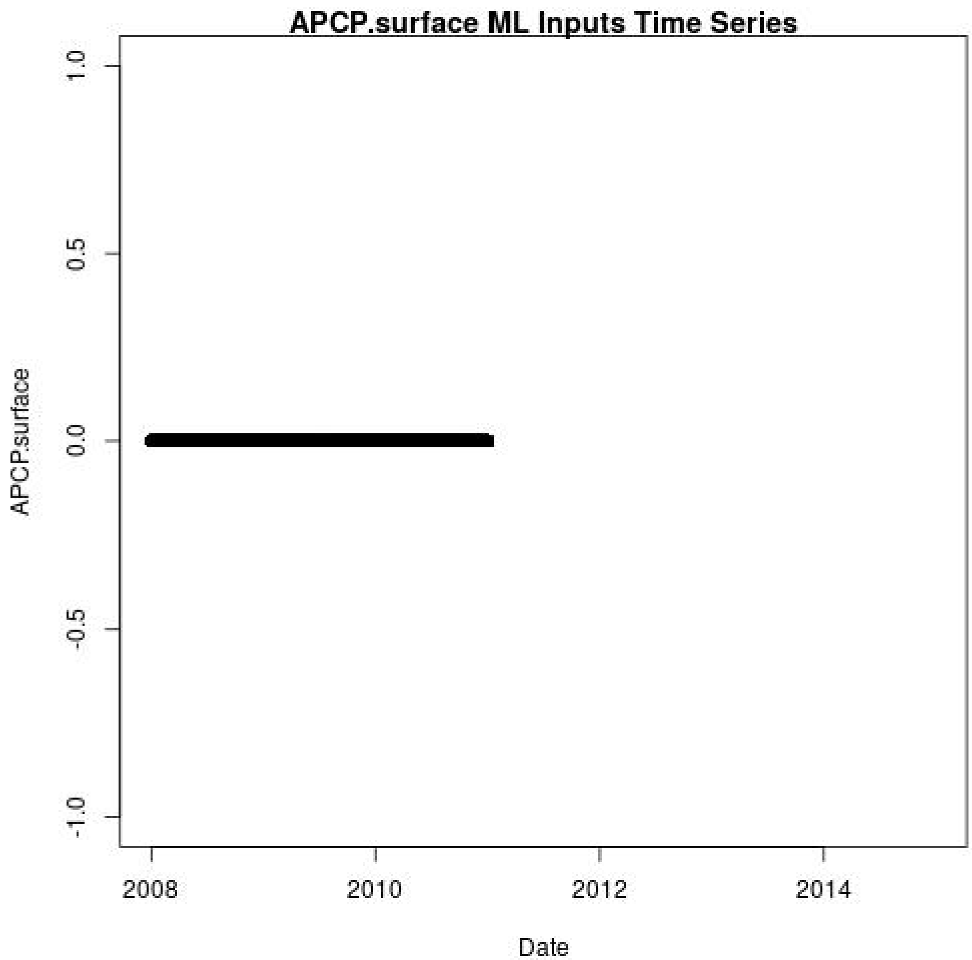

Figure 40: APCP.surface ML Inputs Time Series

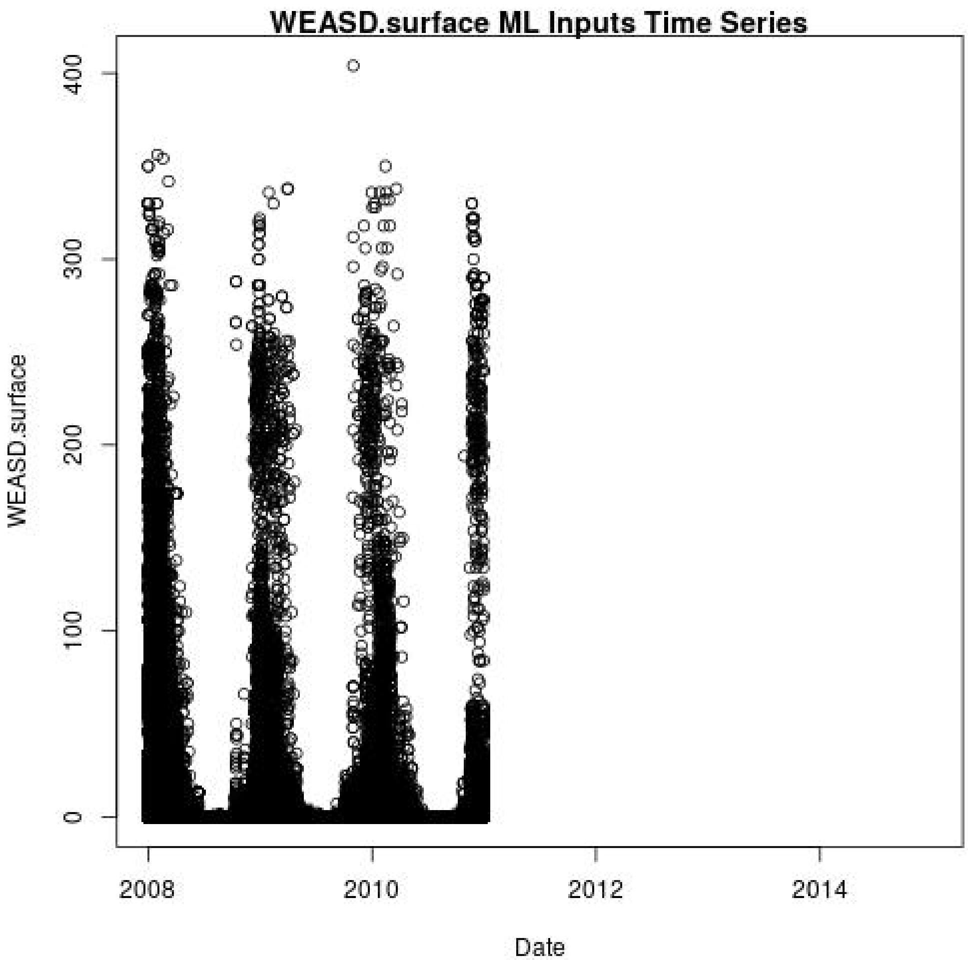

Figure 41: WEASD.surface ML Inputs Time Series

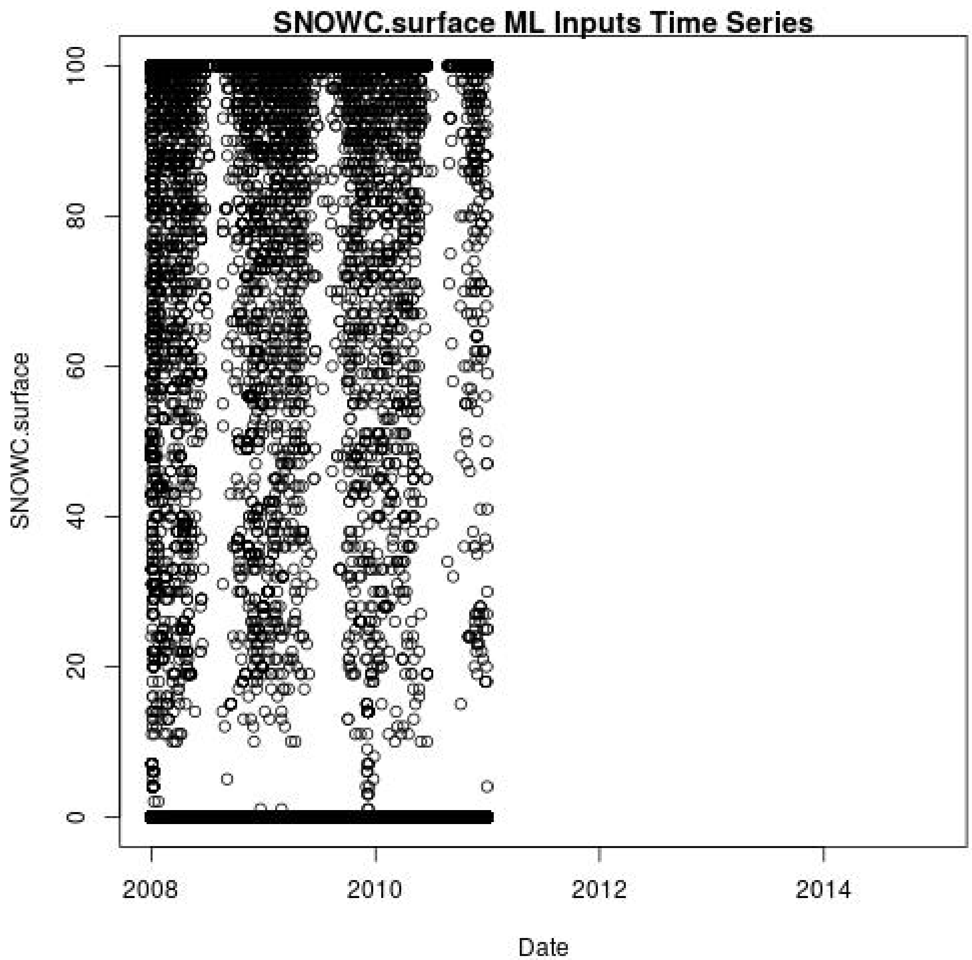

Figure 42: SNOWC.surface ML Inputs Time Series

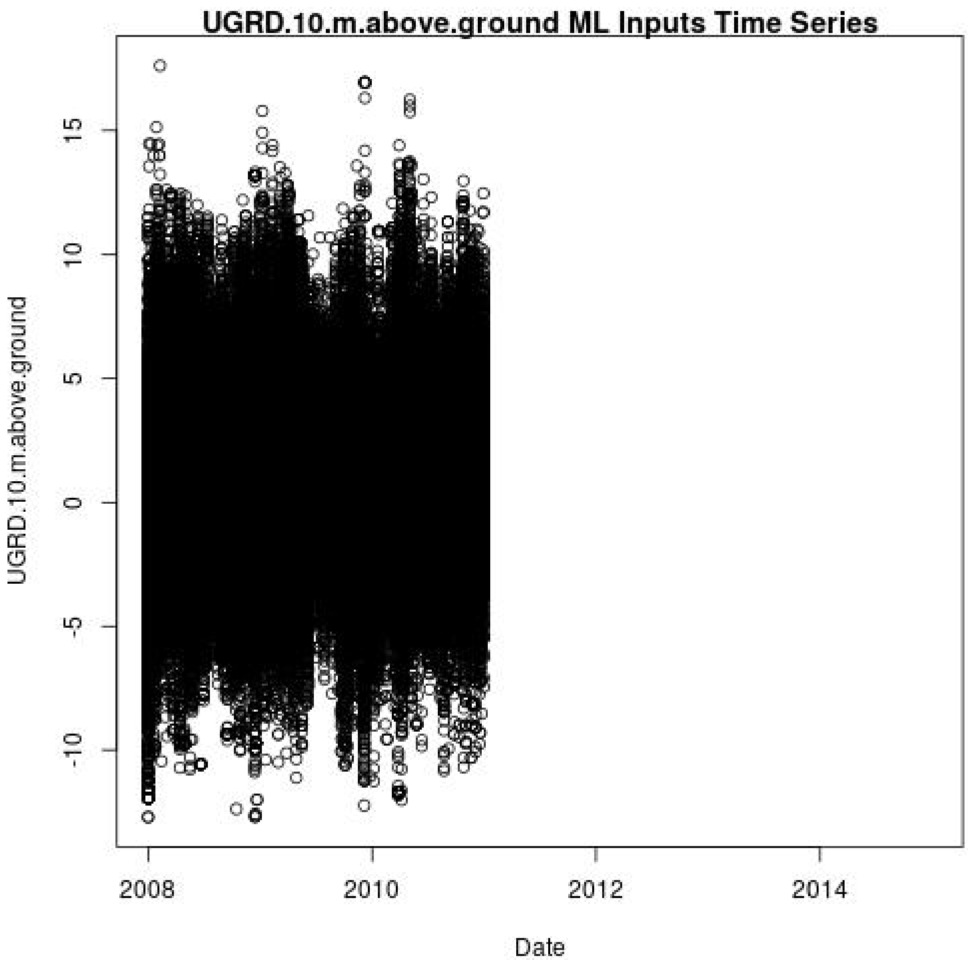

Figure 43: UGRD.10.m.above.ground ML Inputs Time Series

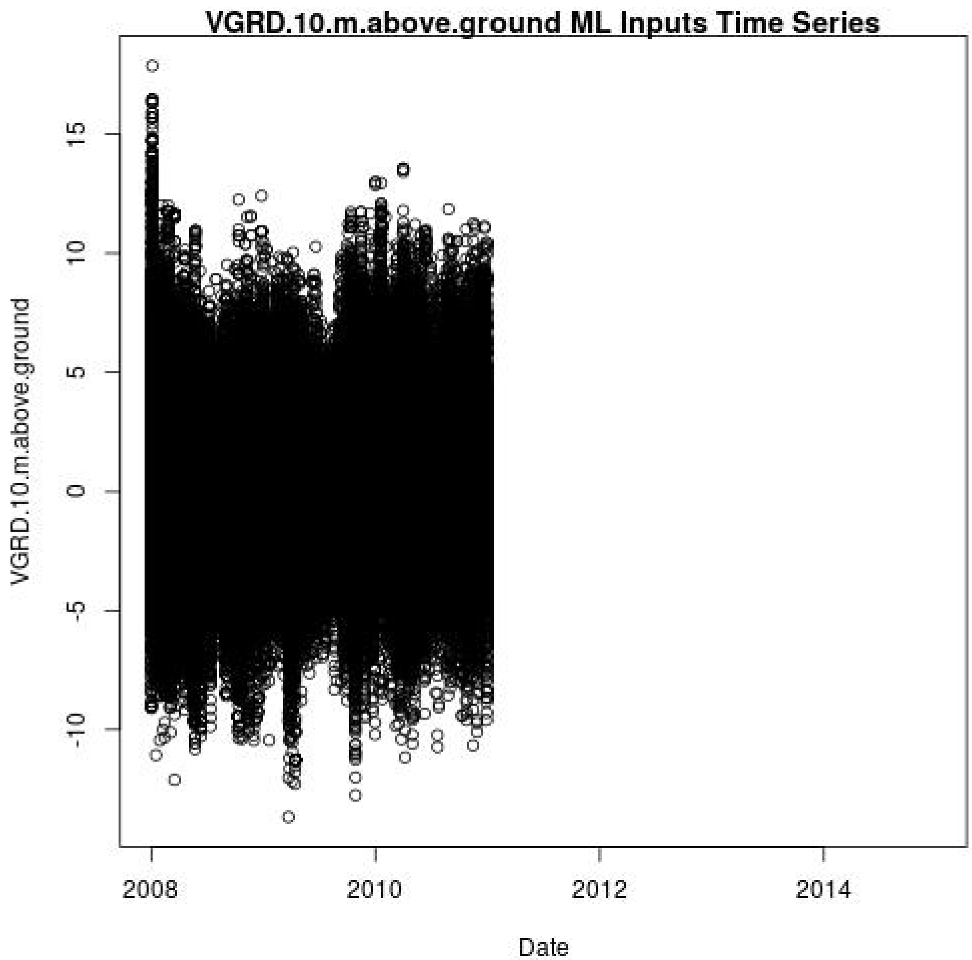

Figure 44: VGRD.10.m.above.ground ML Inputs Time Series

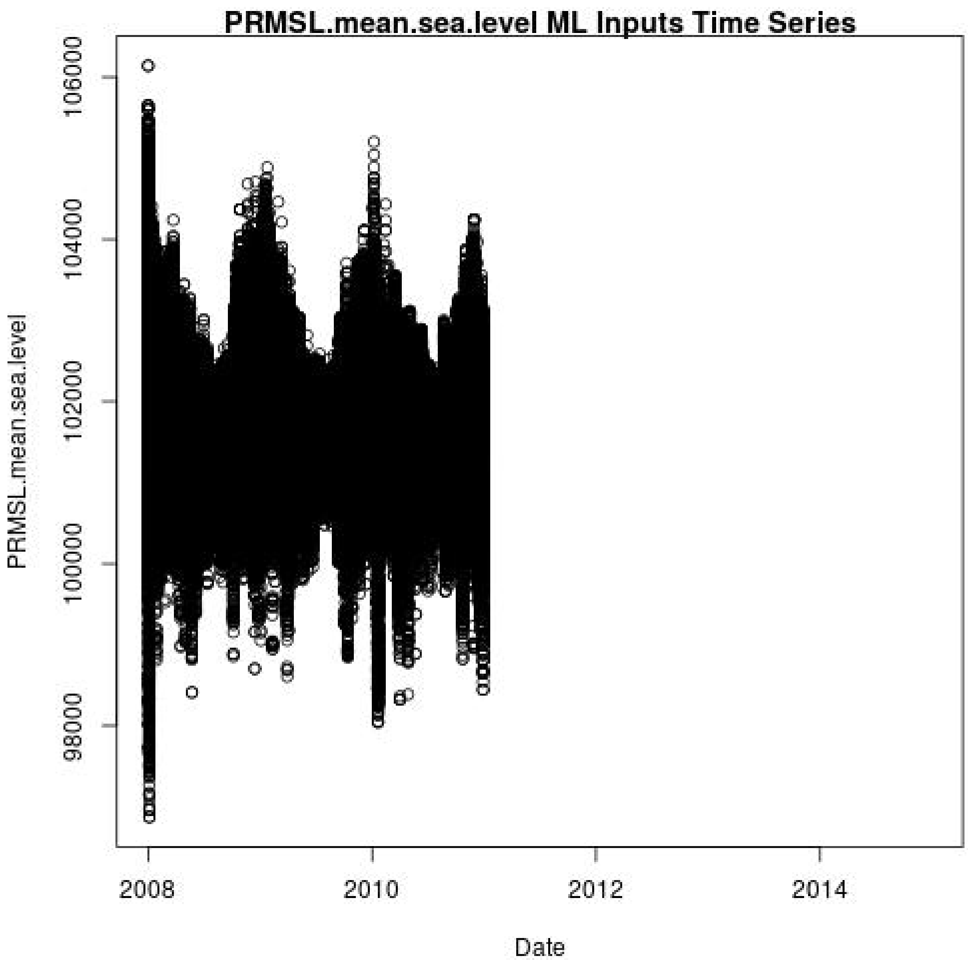

Figure 45: PRMSL.mean.sea.level ML Inputs Time Series

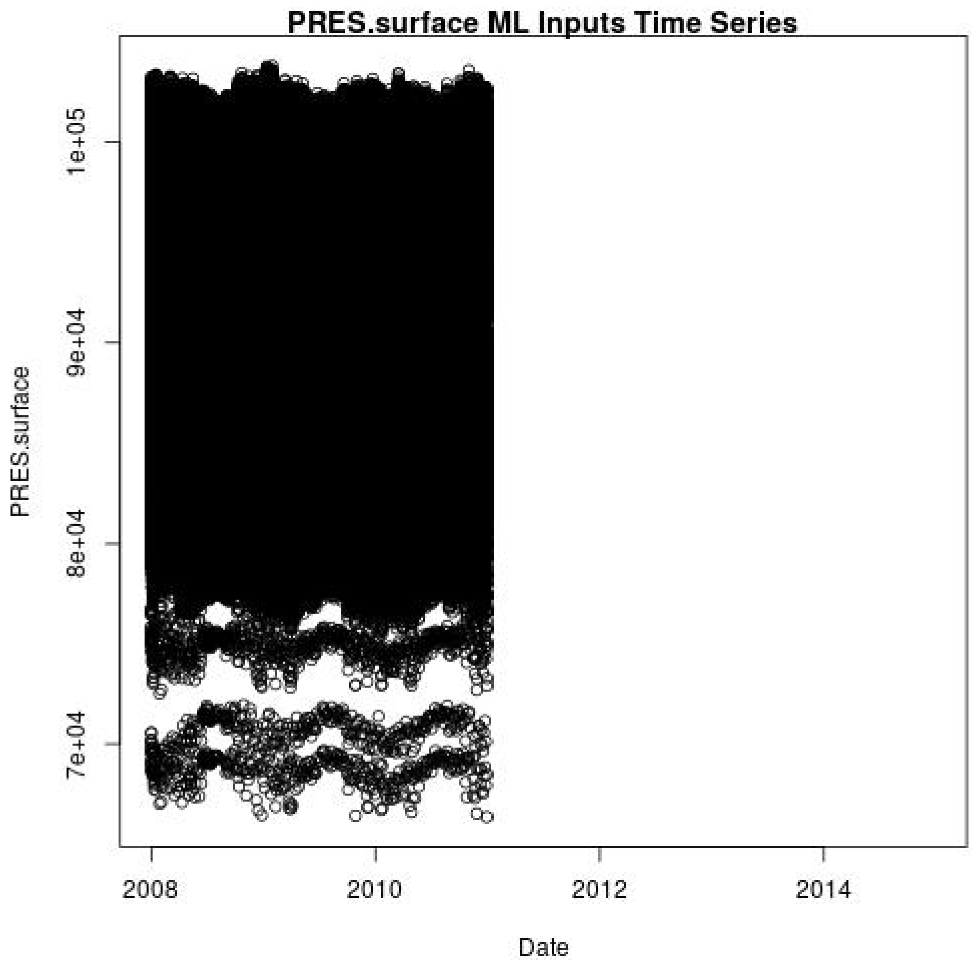

Figure 46: PRES.surface ML Inputs Time Series

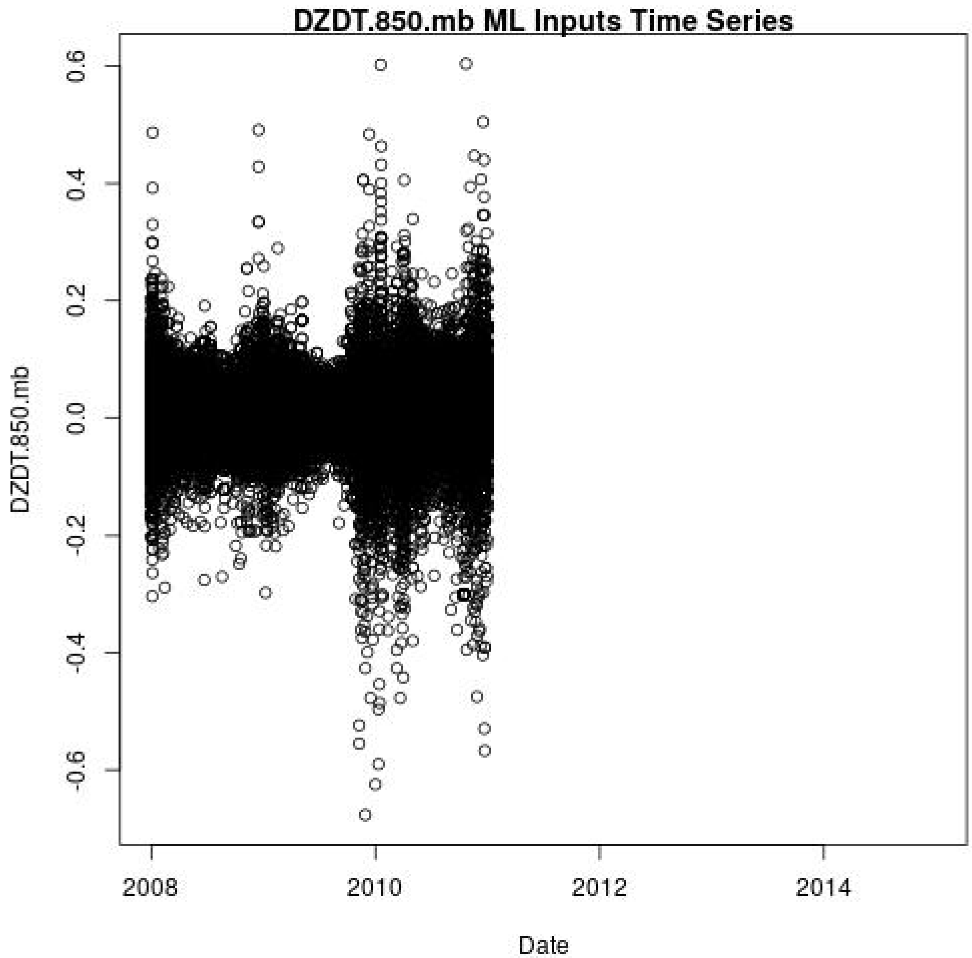

Figure 47: DZDT.850.mb ML Inputs Time Series

10.3 ML Inputs Plot against PM2.5 Images

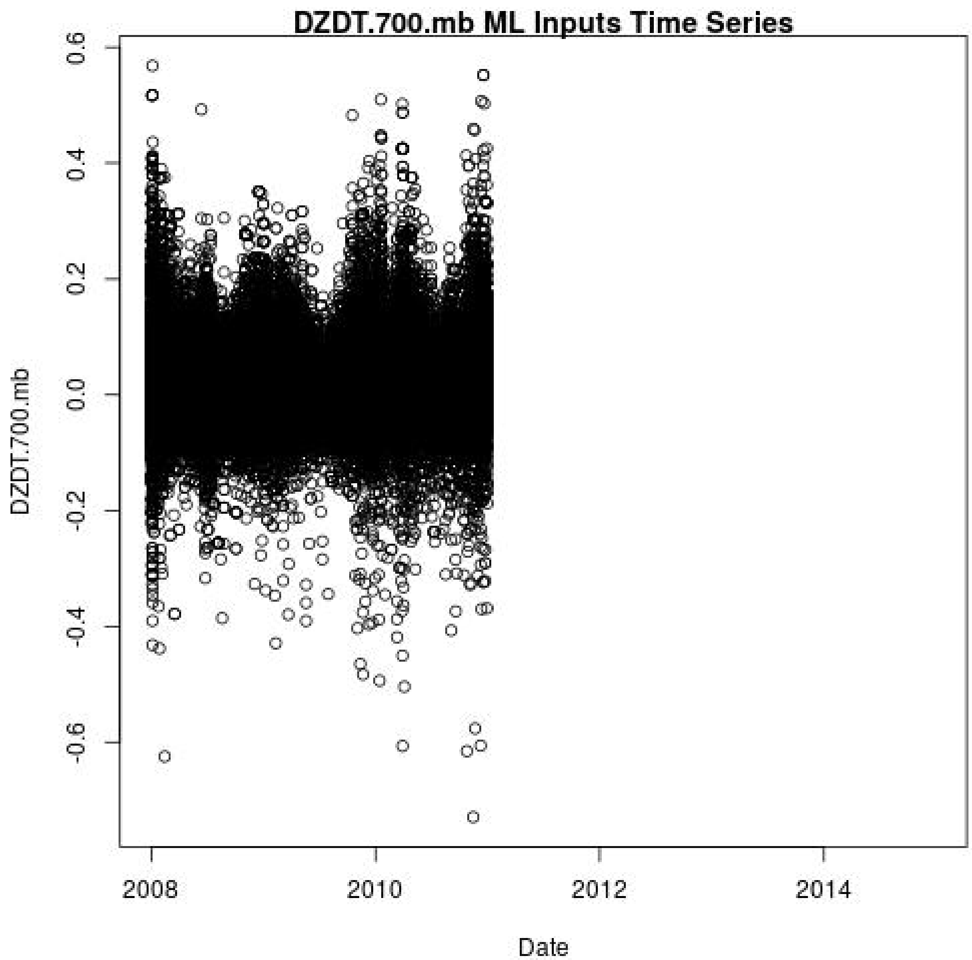

Figure 48: DZDT.700.mb ML Inputs Time Series

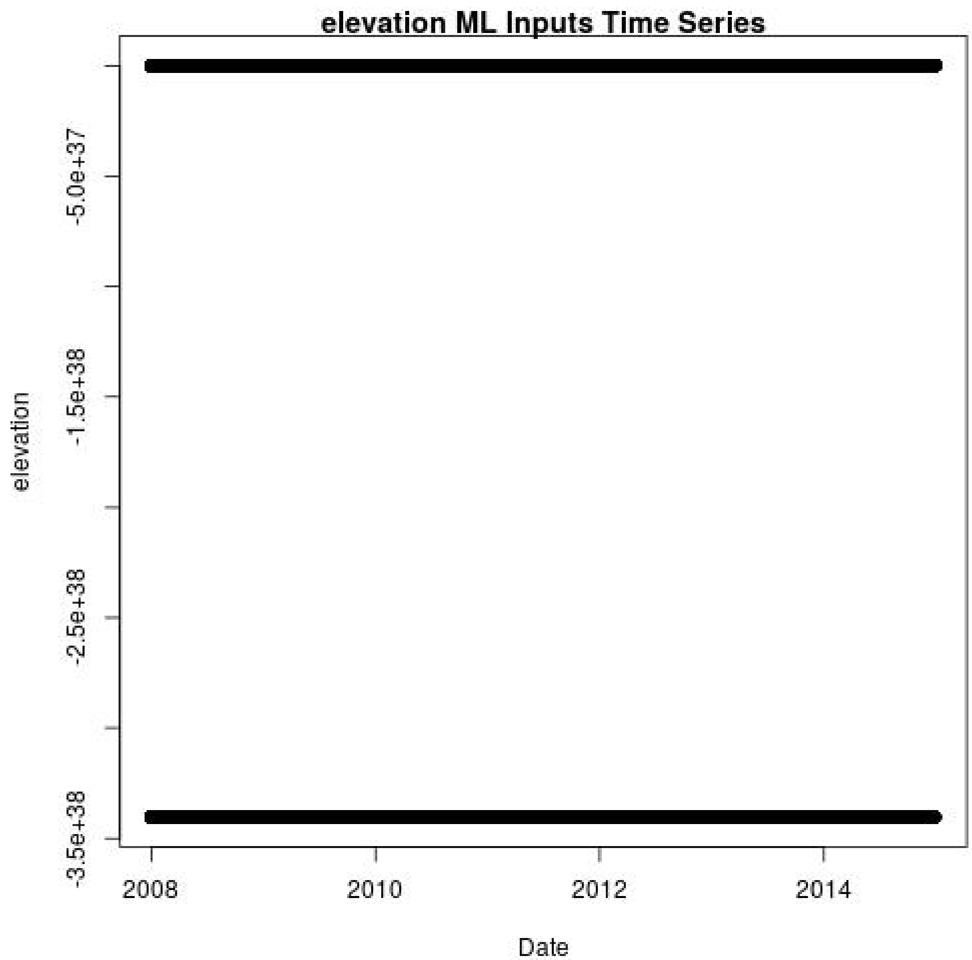

Figure 49: elevation ML Inputs Time Series

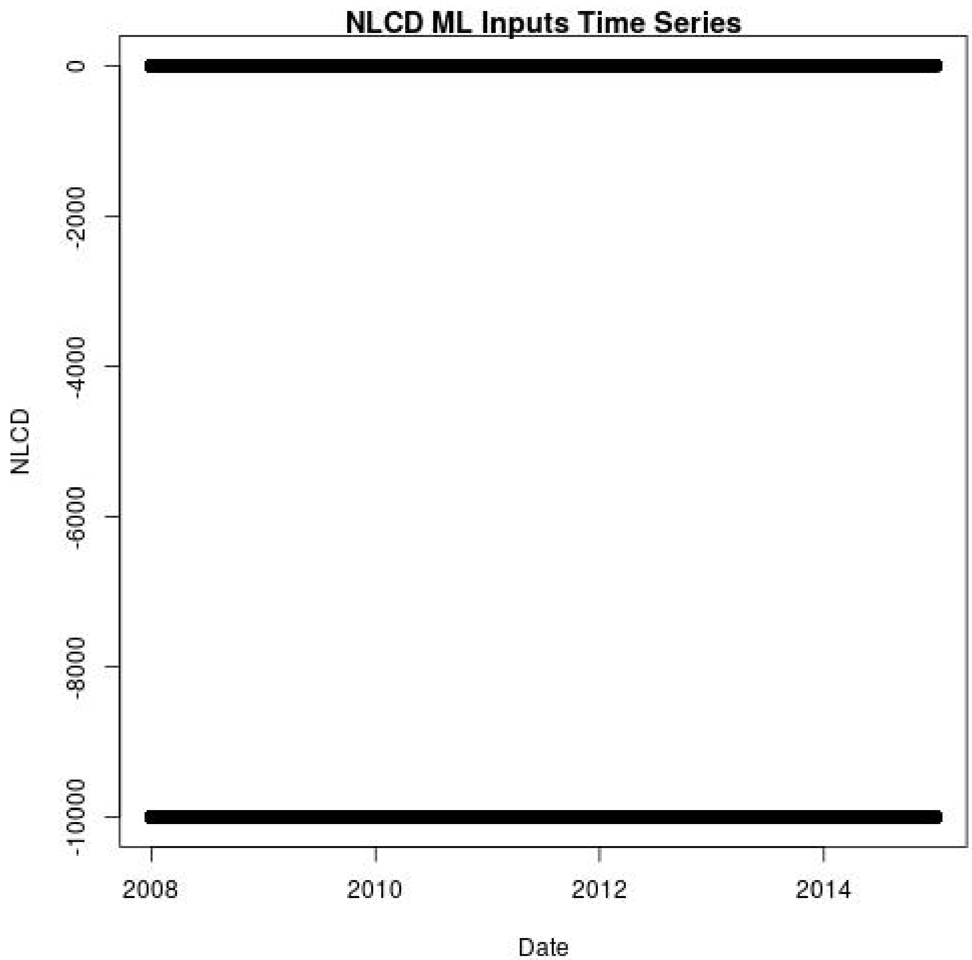

Figure 50: NLCD ML Inputs Time Series

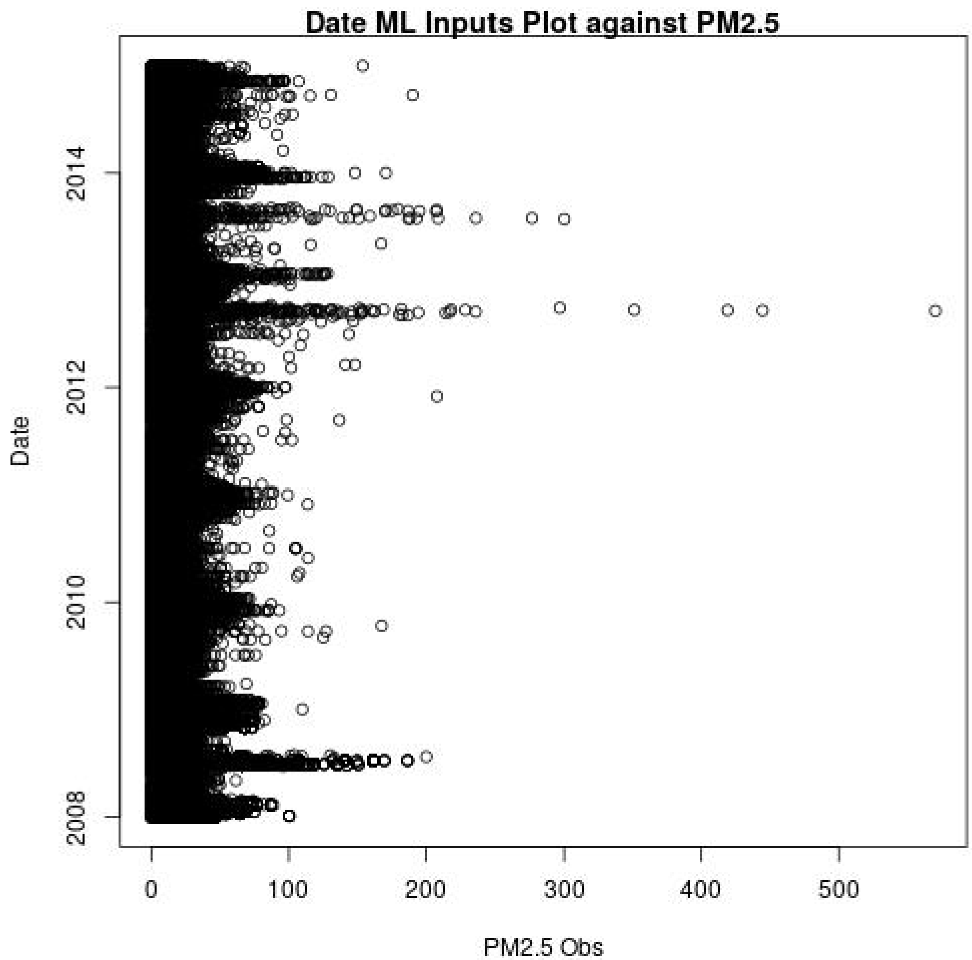

Figure 51: Date ML Inputs Plot against PM2.5

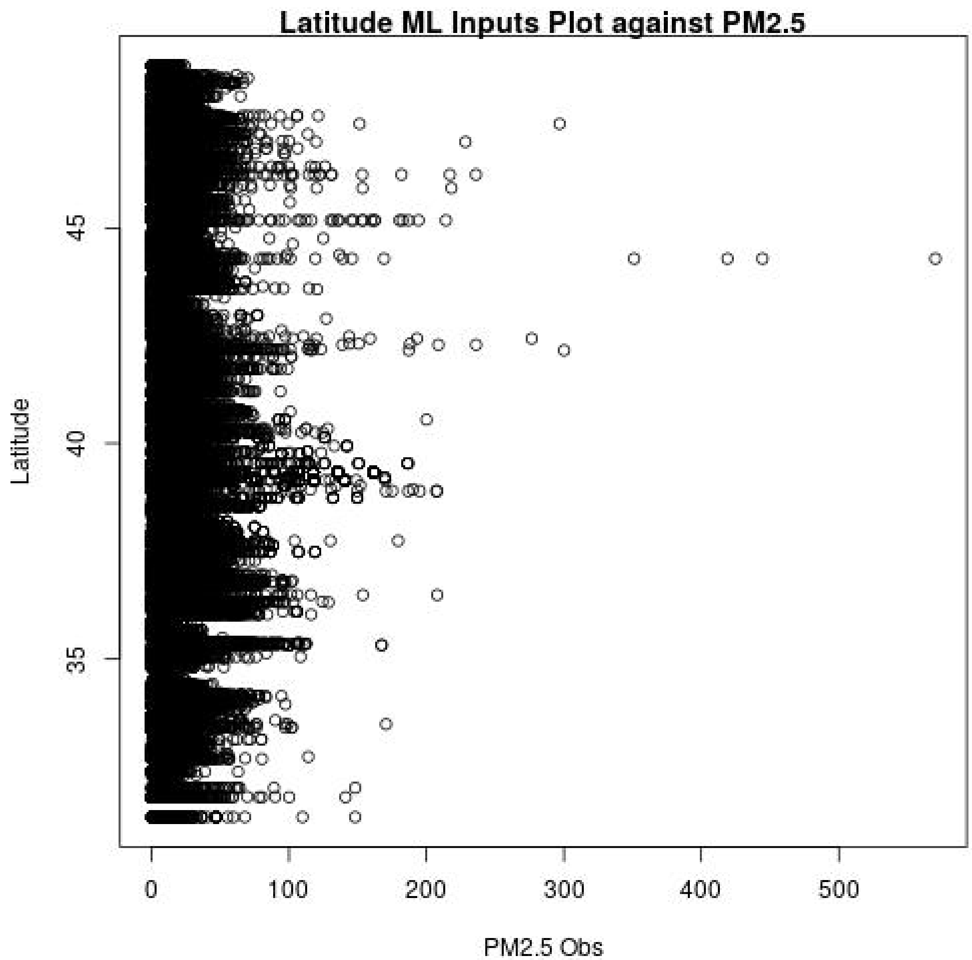

Figure 52: Latitude ML Inputs Plot against PM2.5

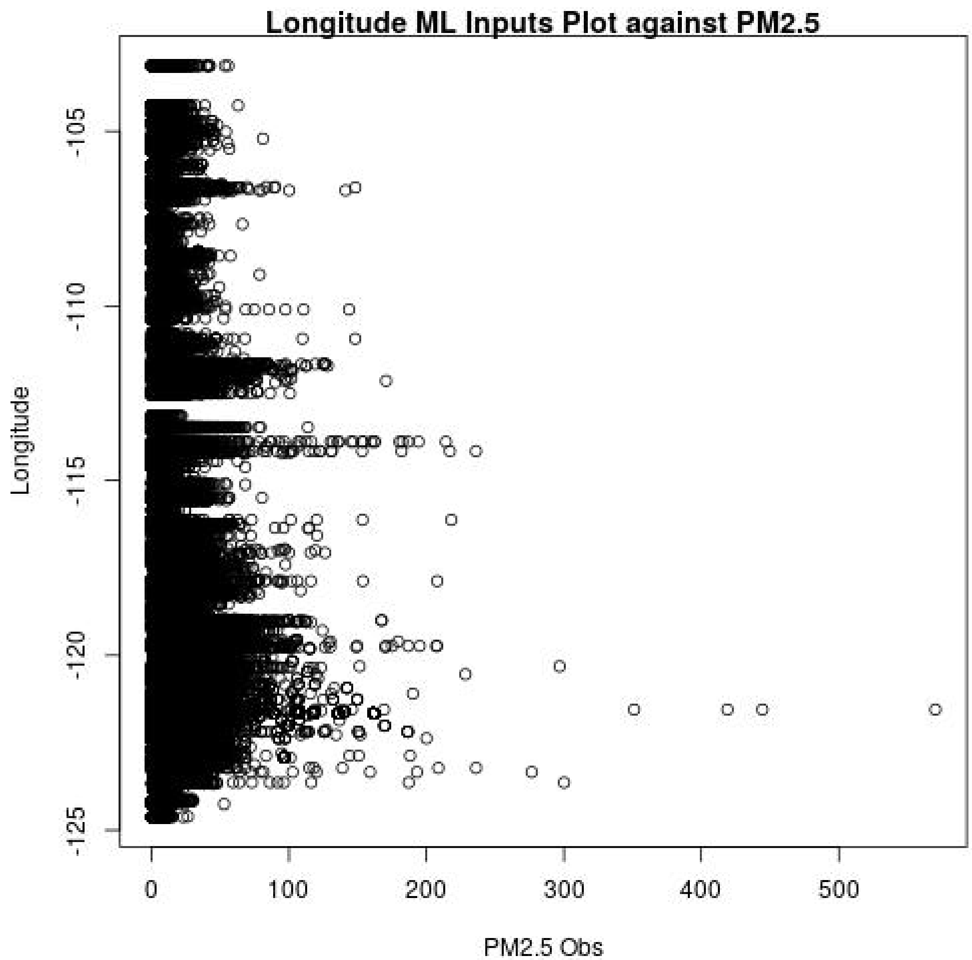

Figure 53: Longitude ML Inputs Plot against PM2.5

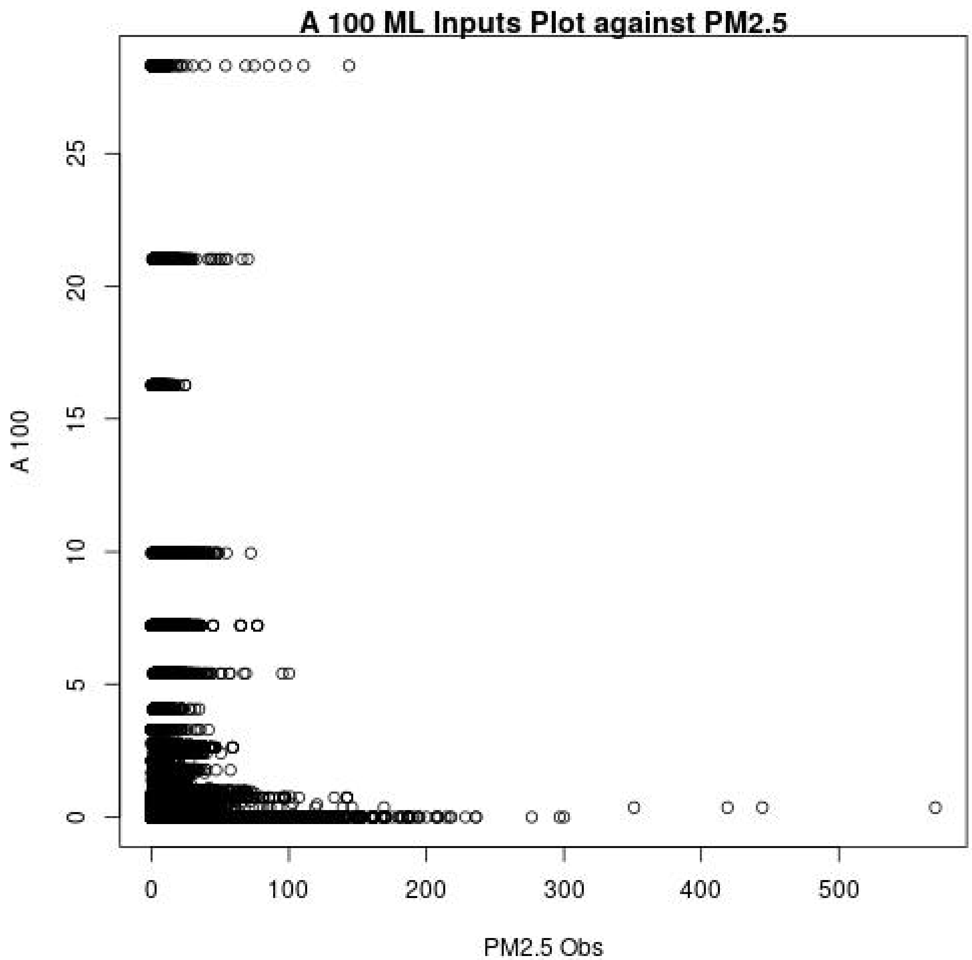

Figure 54: A 100 ML Inputs Plot against PM2.5

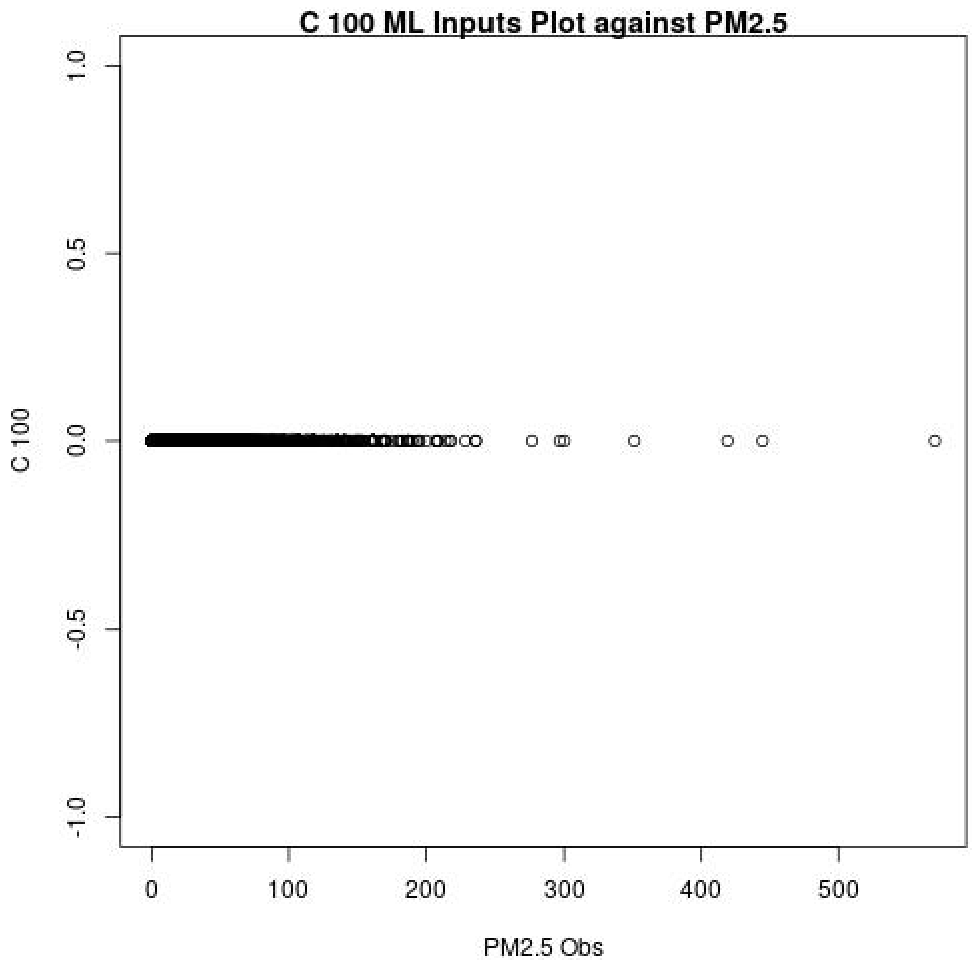

Figure 55: C 100 ML Inputs Plot against PM2.5

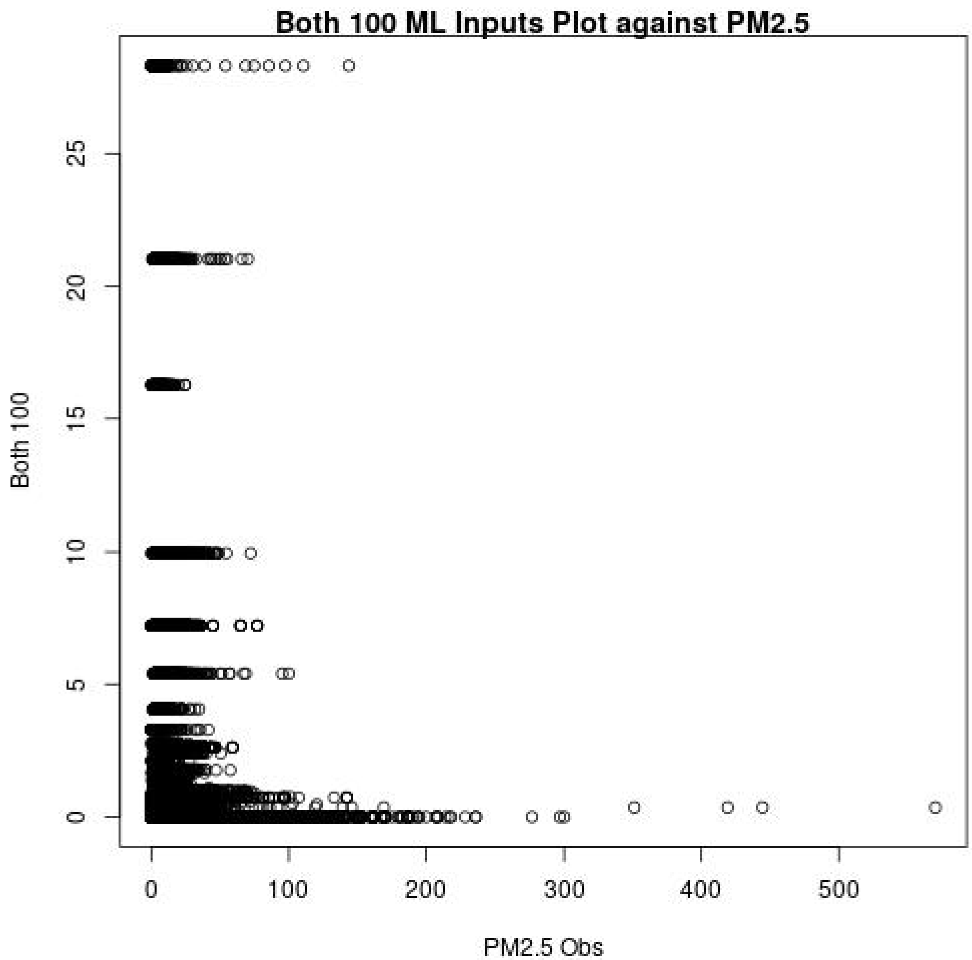

Figure 56: Both 100 ML Inputs Plot against PM2.5

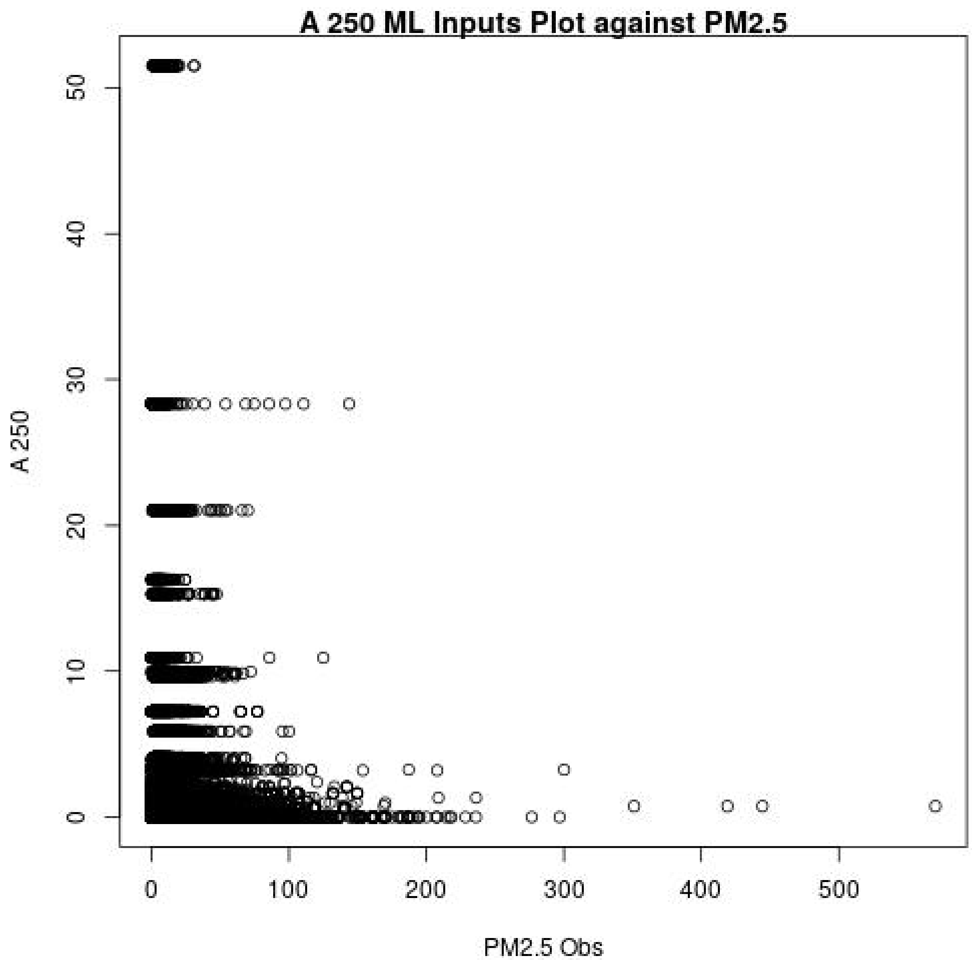

Figure 57: A 250 ML Inputs Plot against PM2.5

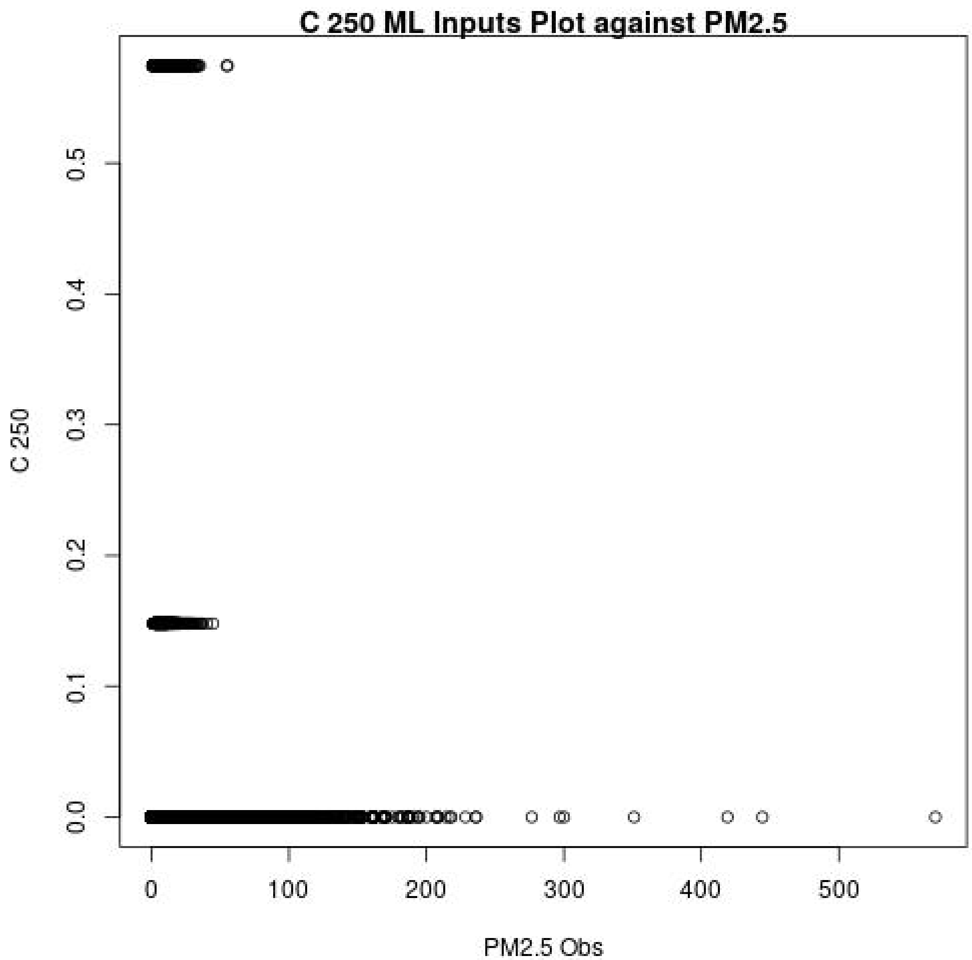

Figure 58: C 250 ML Inputs Plot against PM2.5

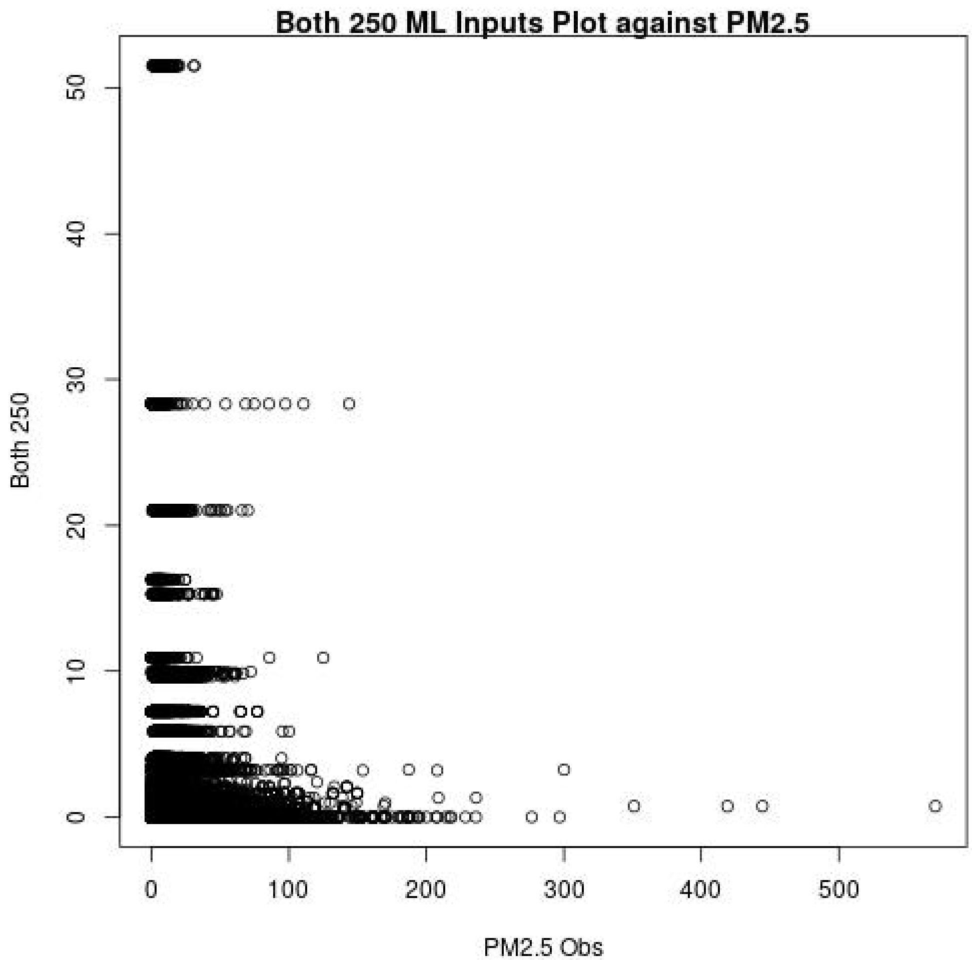

Figure 59: Both 250 ML Inputs Plot against PM2.5

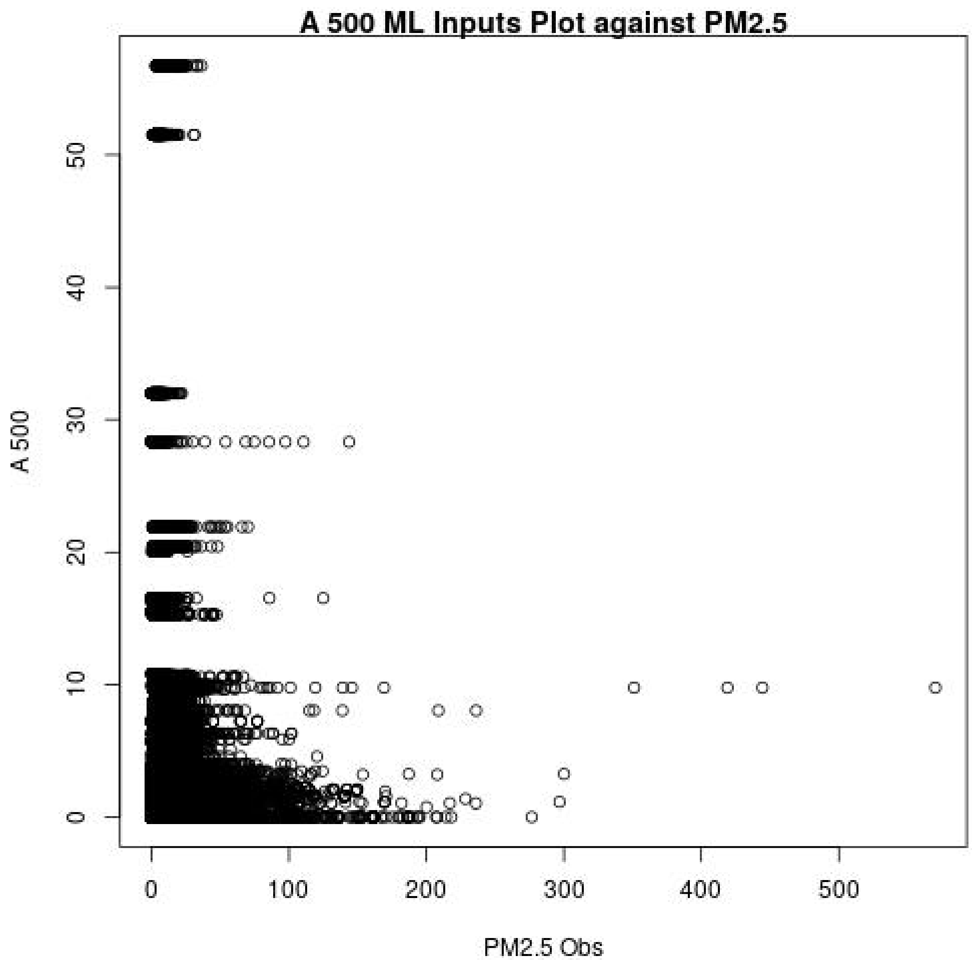

Figure 60: A 500 ML Inputs Plot against PM2.5

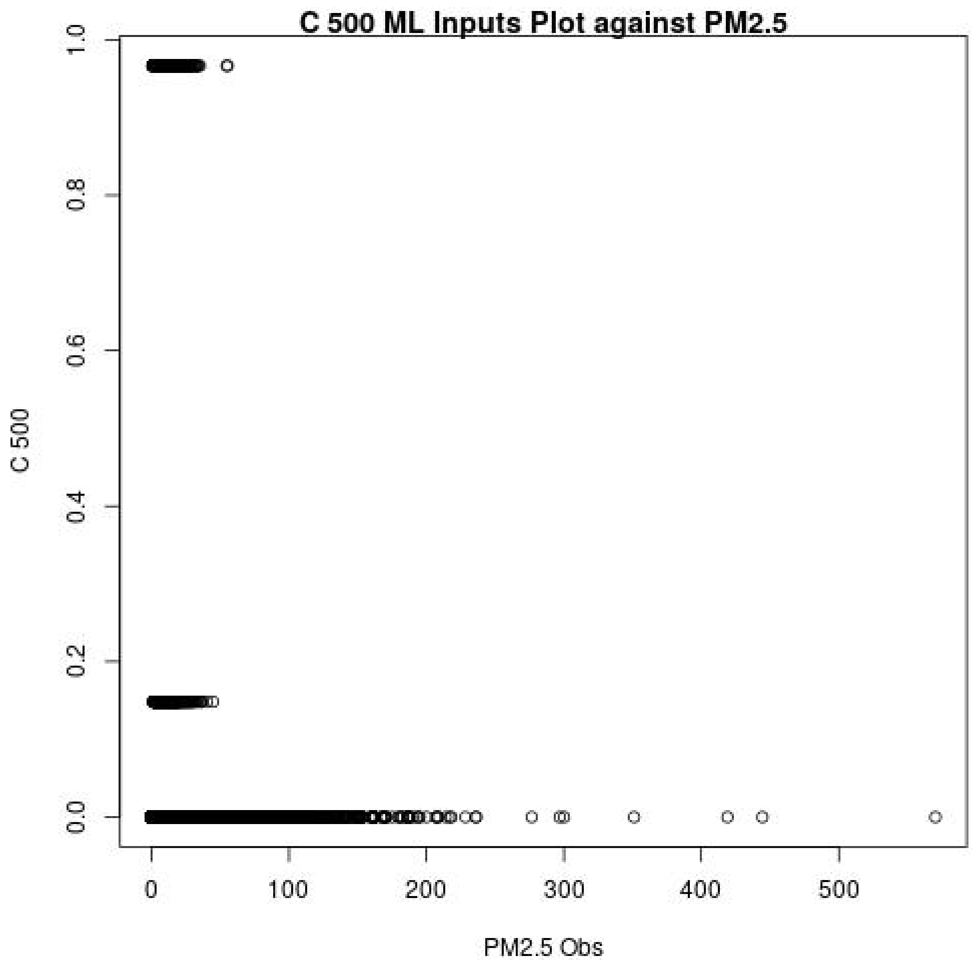

Figure 61: C 500 ML Inputs Plot against PM2.5

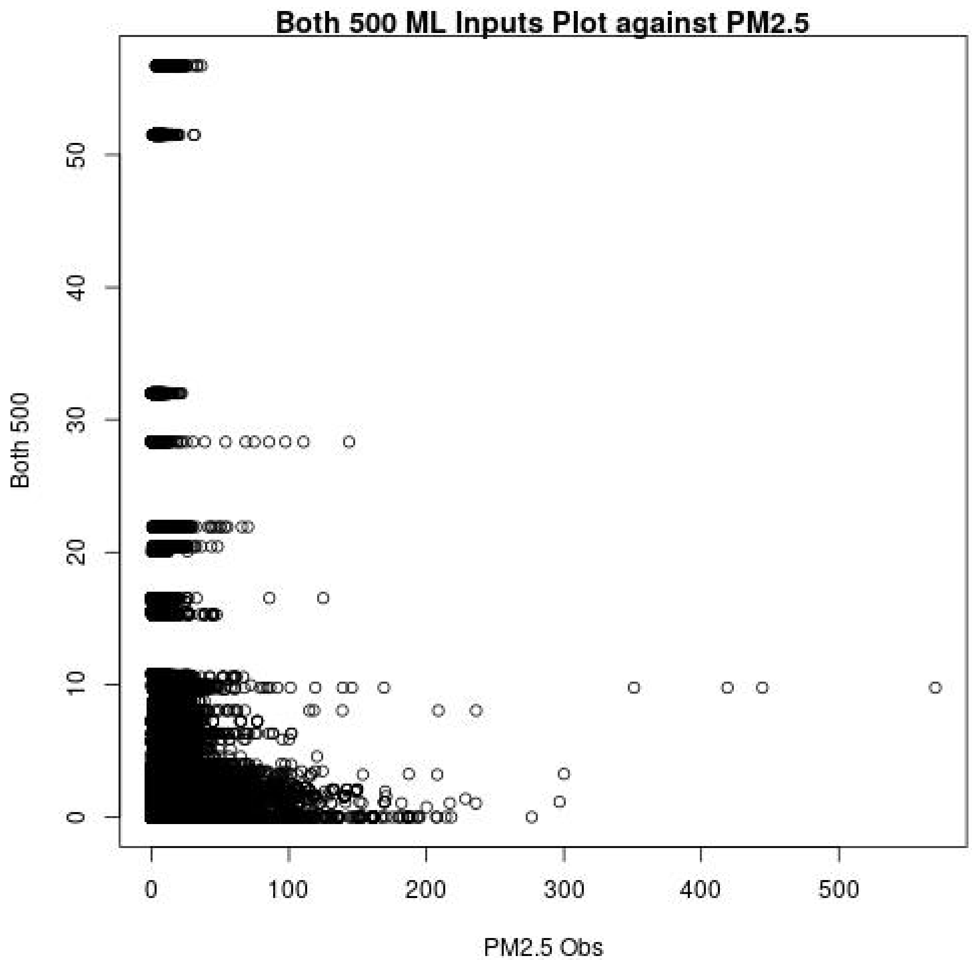

Figure 62: Both 500 ML Inputs Plot against PM2.5

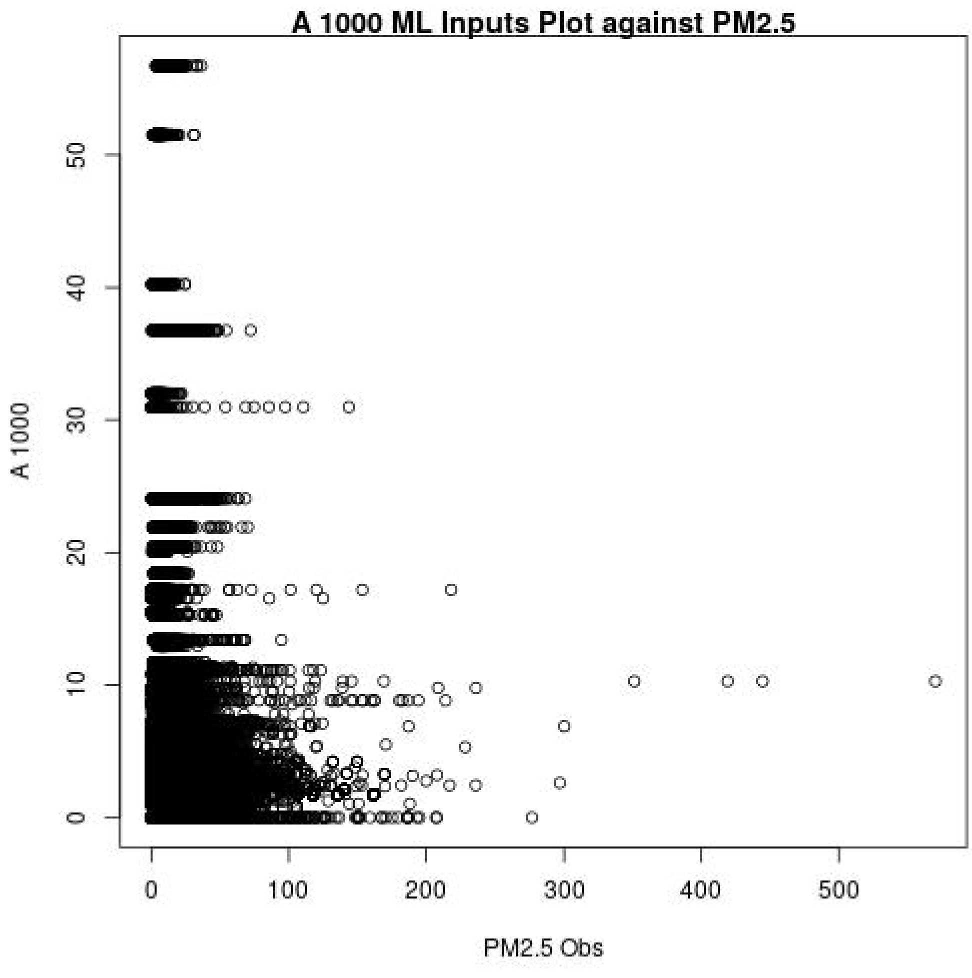

Figure 63: A 1000 ML Inputs Plot against PM2.5

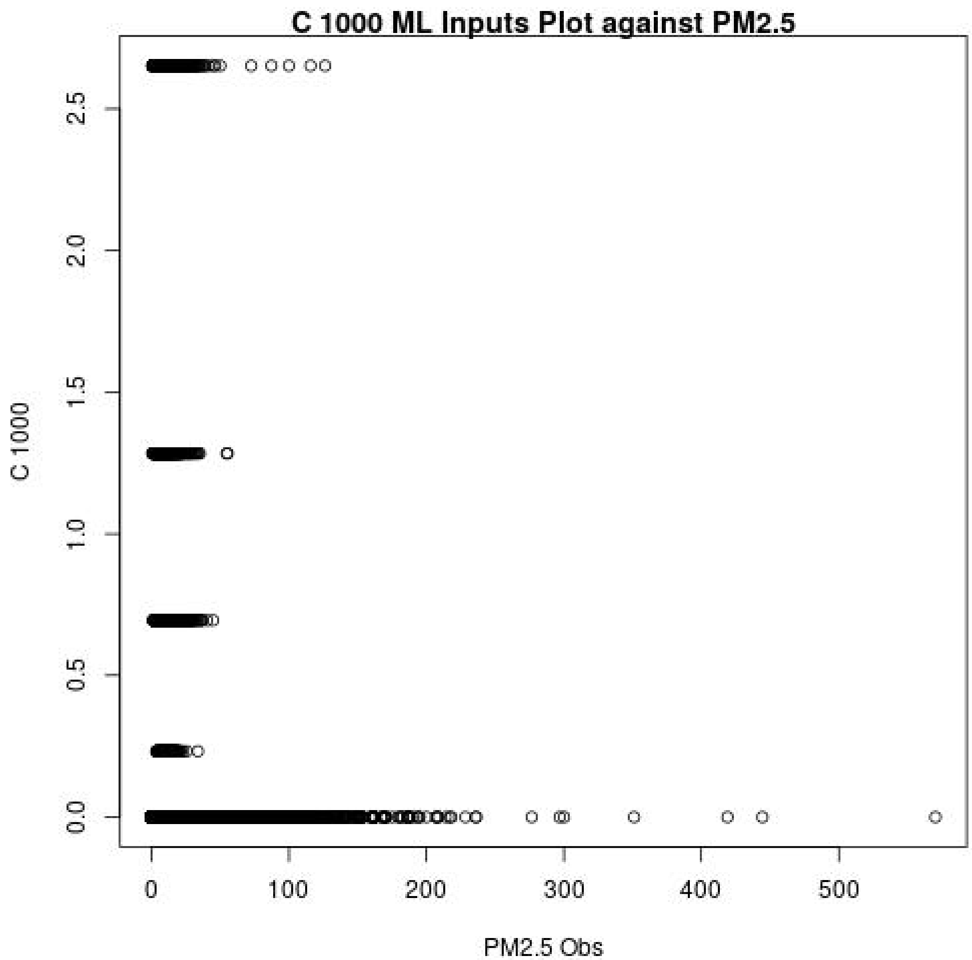

Figure 64: C 1000 ML Inputs Plot against PM2.5

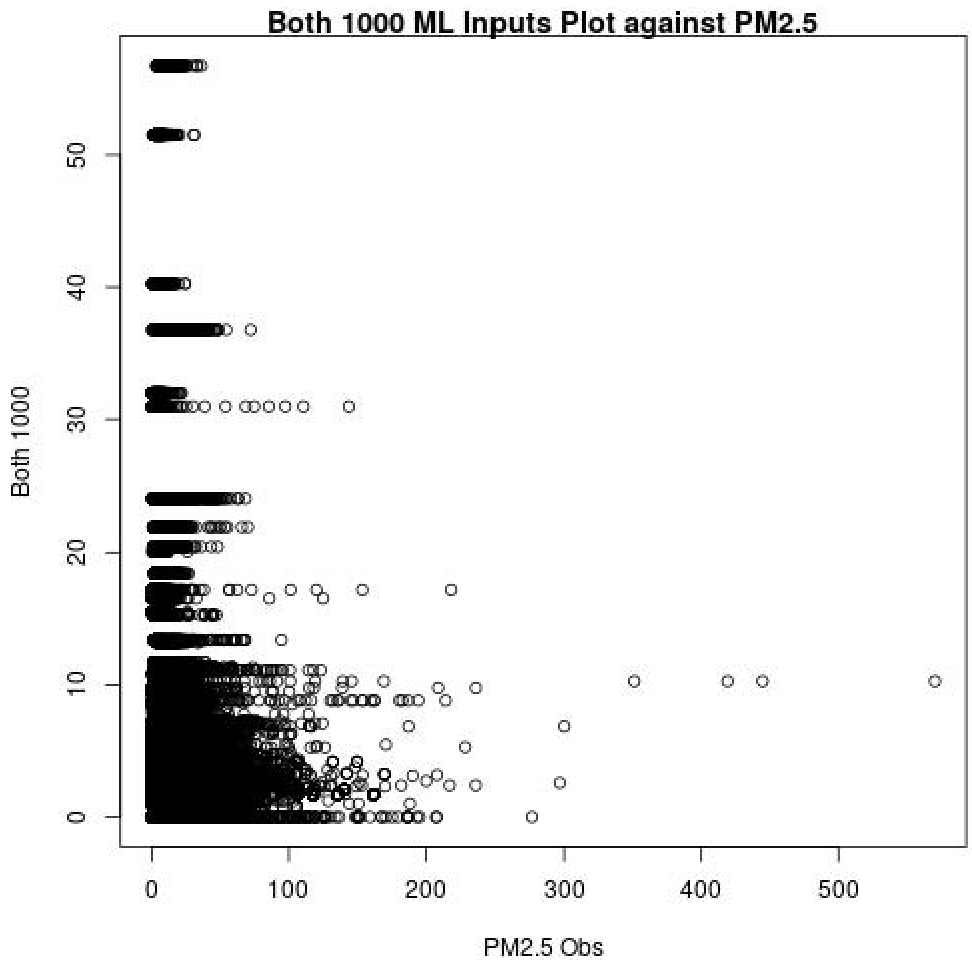

Figure 65: Both 1000 ML Inputs Plot against PM2.5

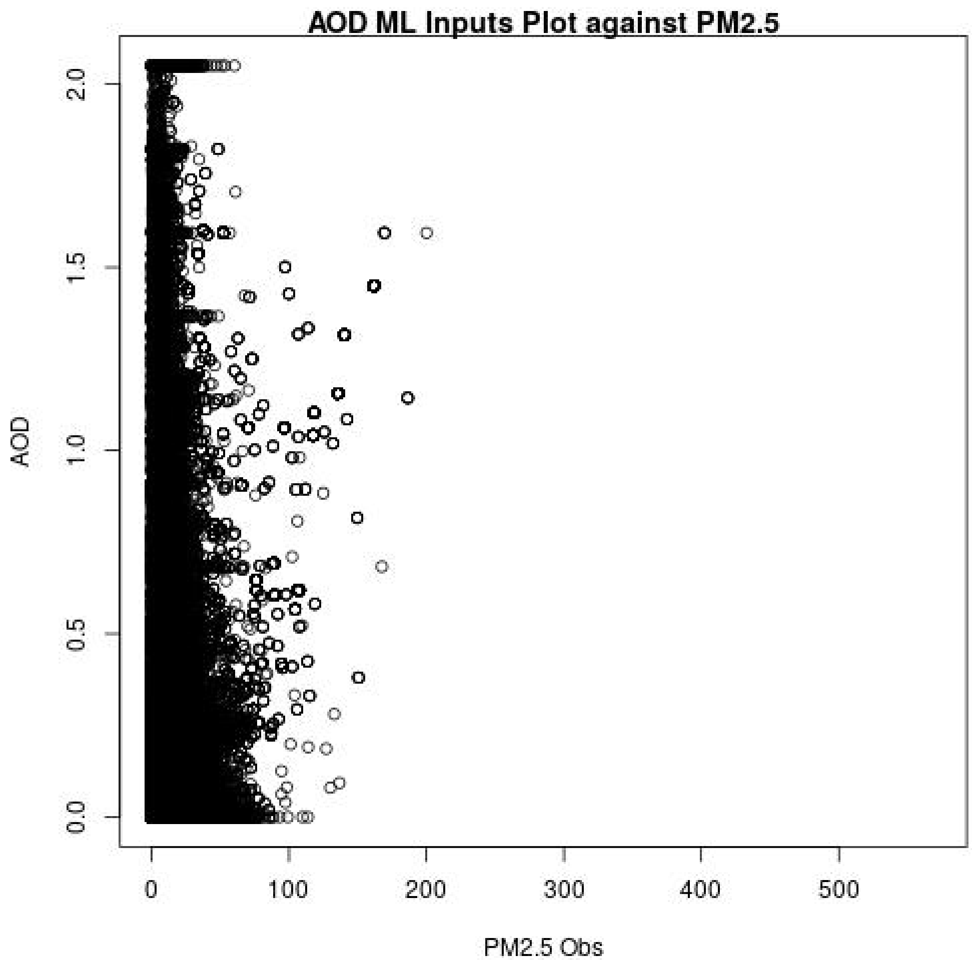

Figure 66: AOD ML Inputs Plot against PM2.5

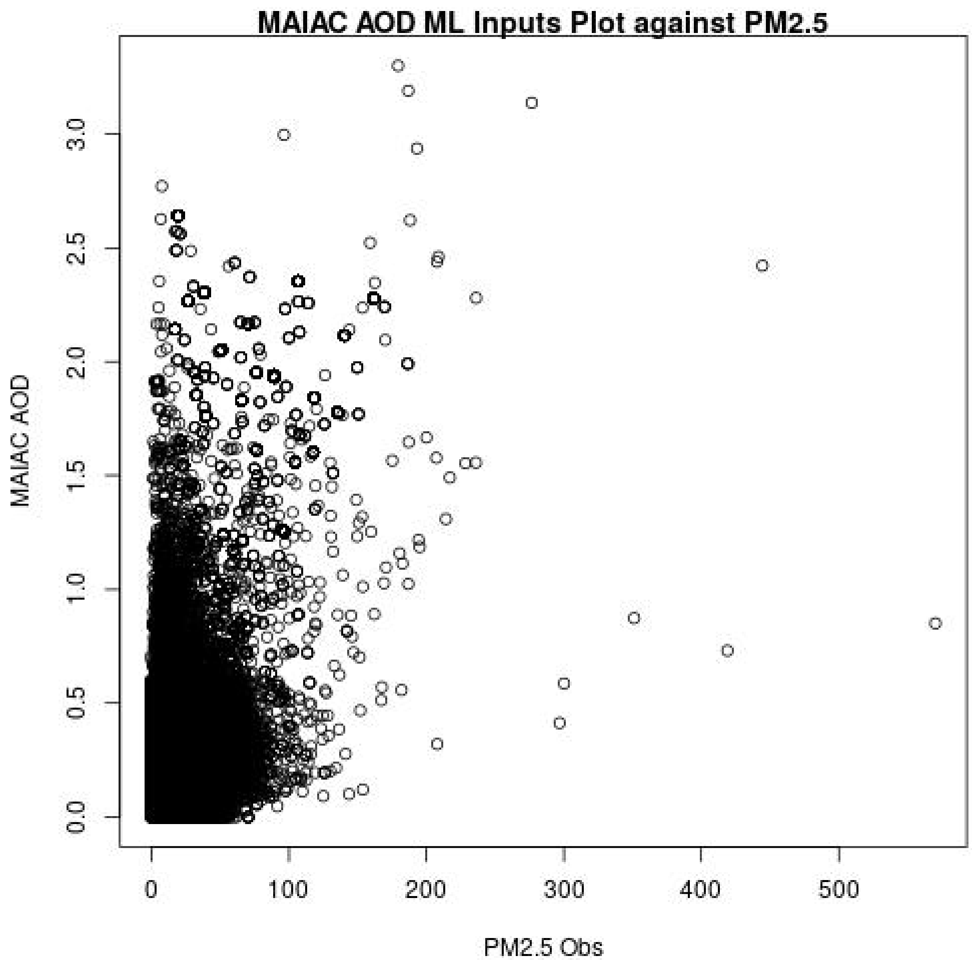

Figure 67: MAIAC AOD ML Inputs Plot against PM2.5

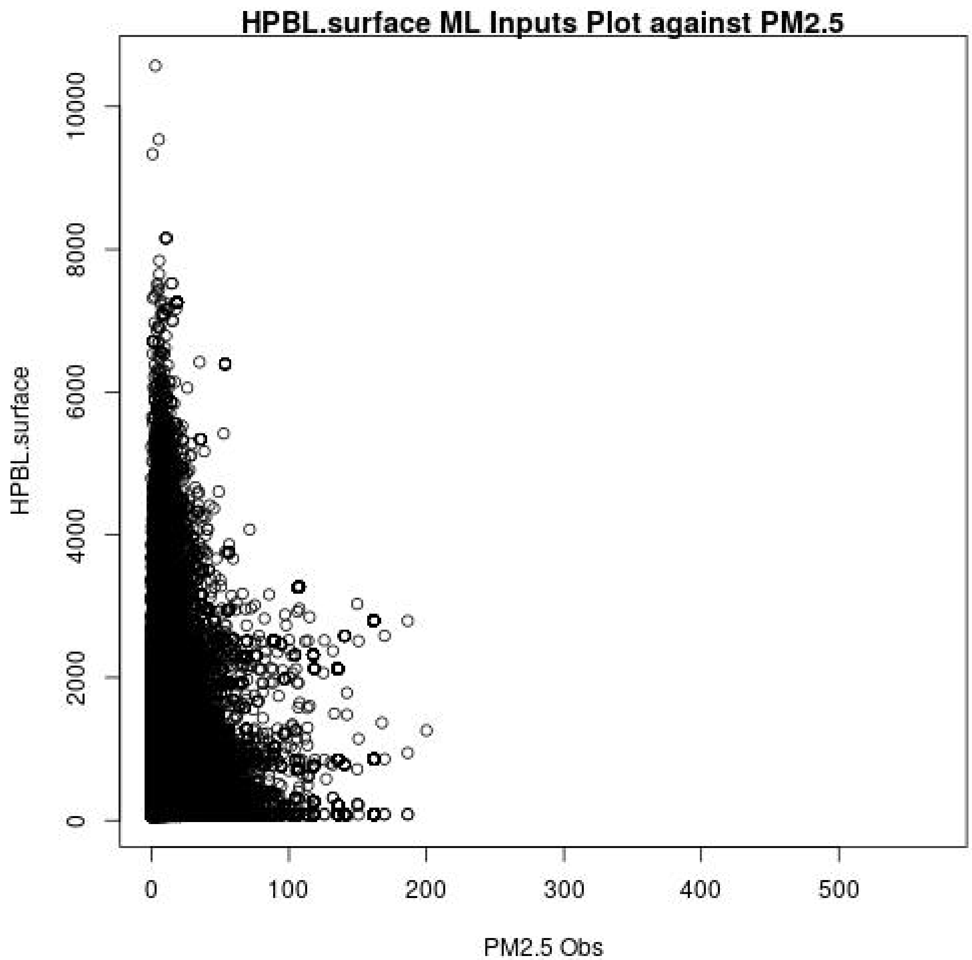

Figure 68: HPBL.surface ML Inputs Plot against PM2.5

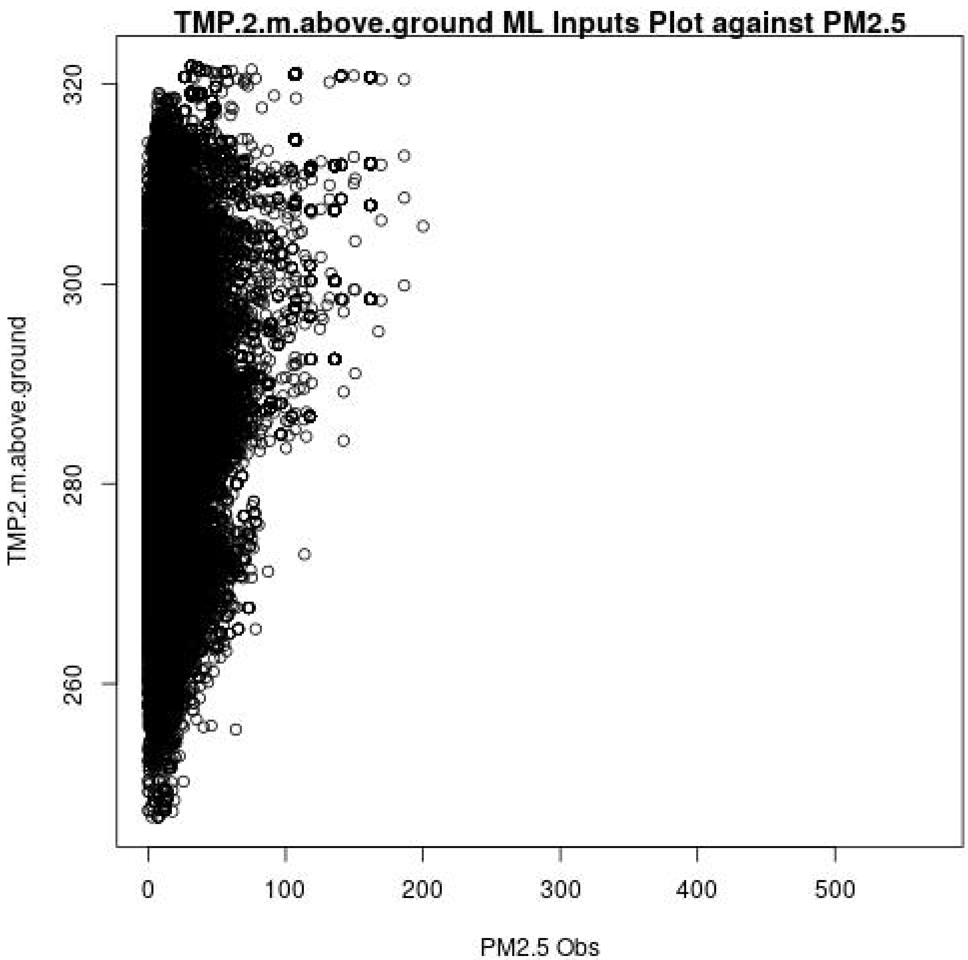

Figure 69: TMP.2.m.above.ground ML Inputs Plot against PM2.5

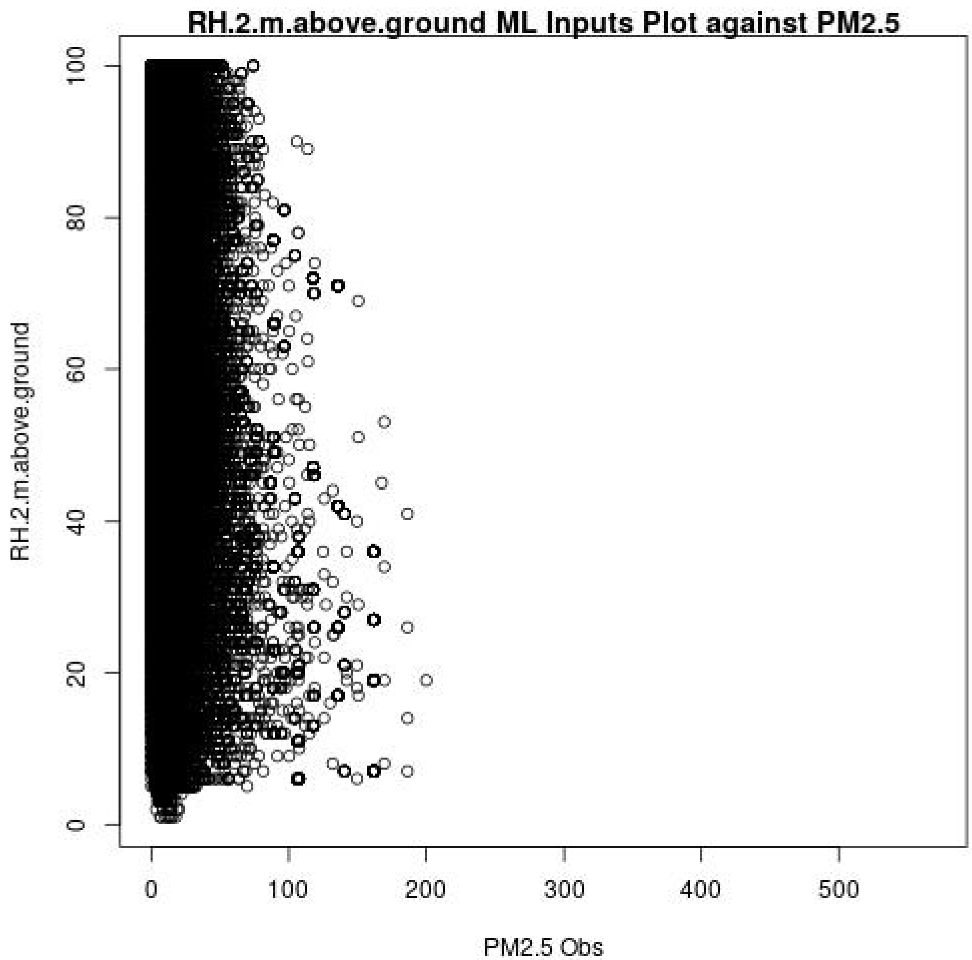

Figure 70: RH.2.m.above.ground ML Inputs Plot against PM2.5

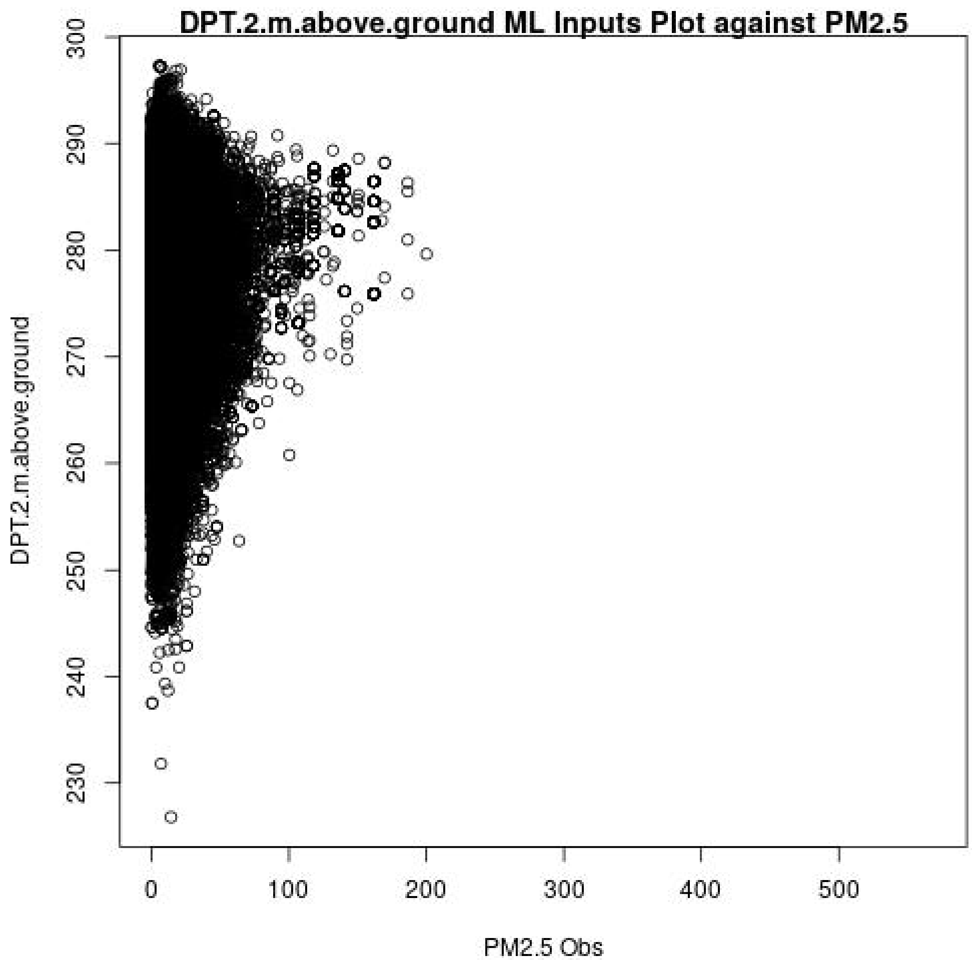

Figure 71: DPT.2.m.above.ground ML Inputs Plot against PM2.5

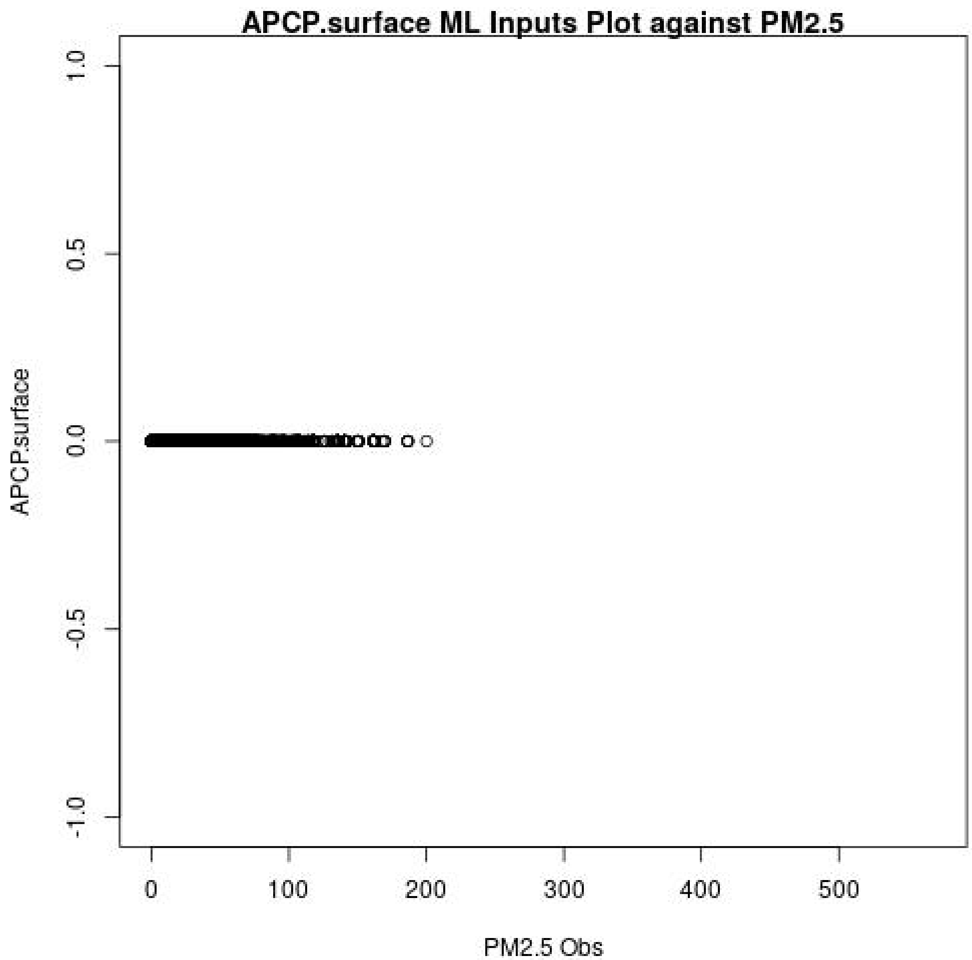

Figure 72: APCP.surface ML Inputs Plot against PM2.5

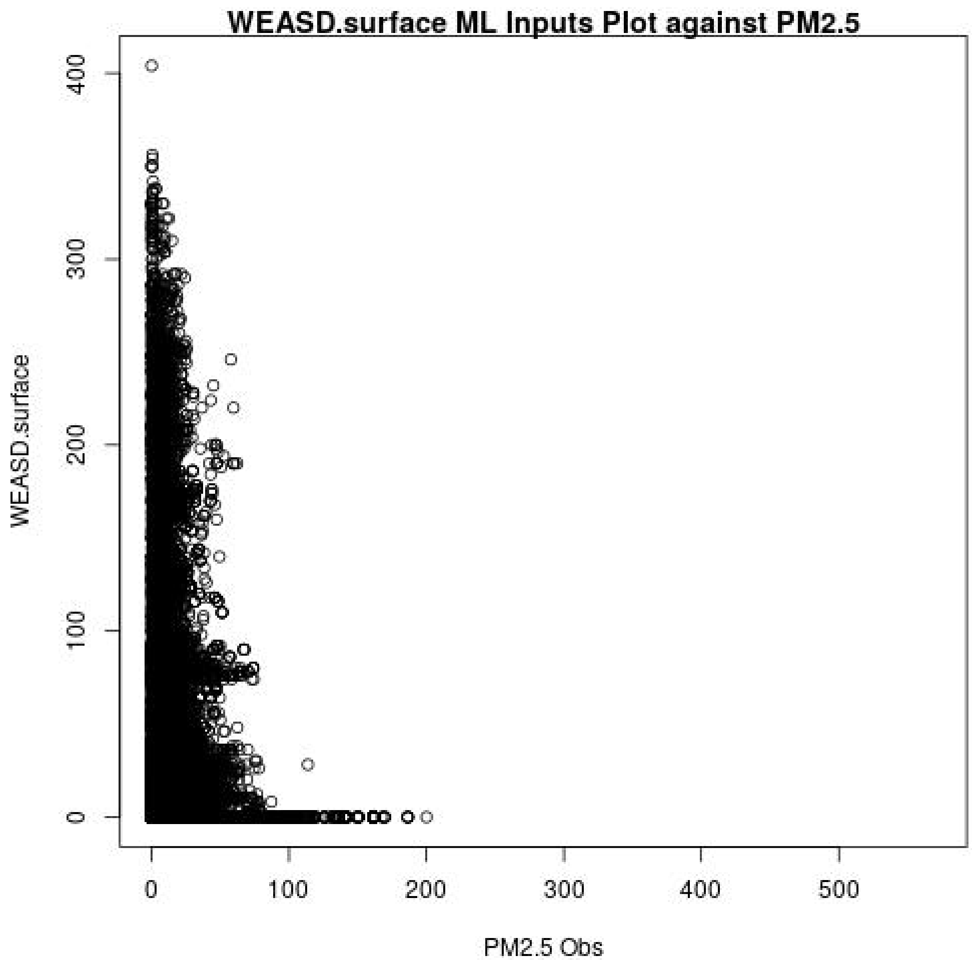

Figure 73: WEASD.surface ML Inputs Plot against PM2.5

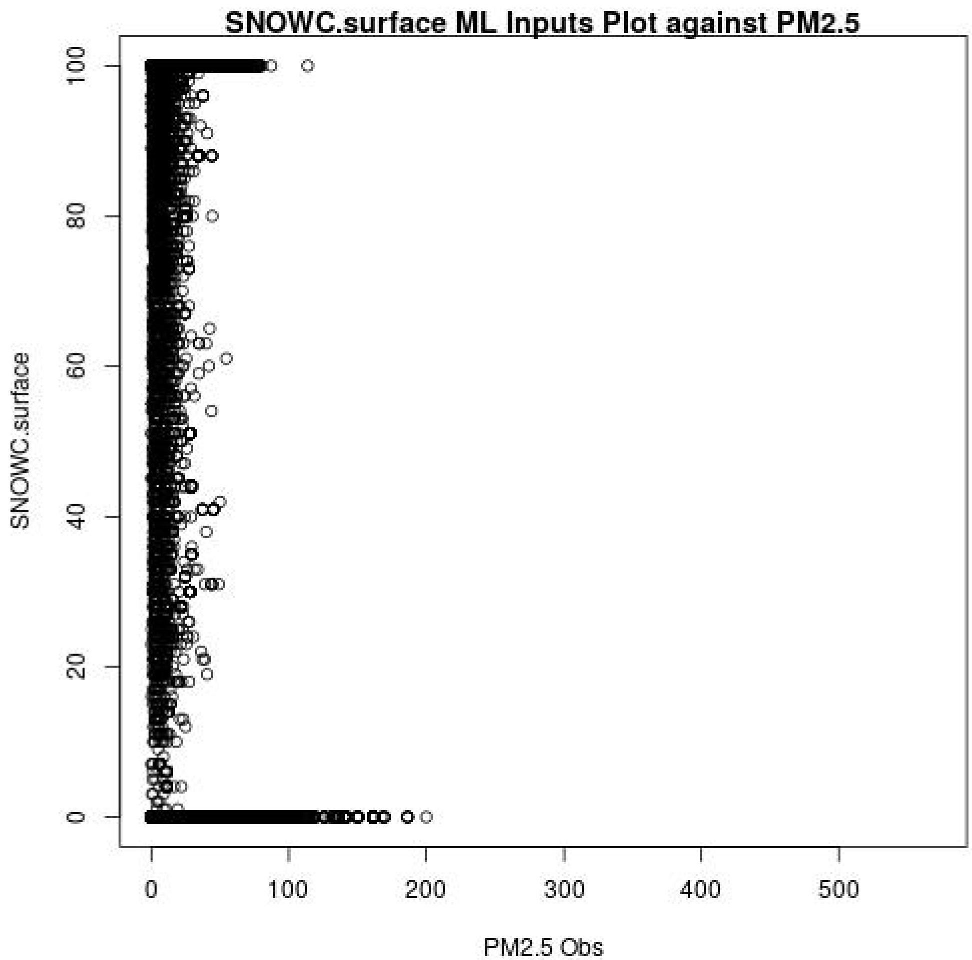

Figure 74: SNOWC.surface ML Inputs Plot against PM2.5

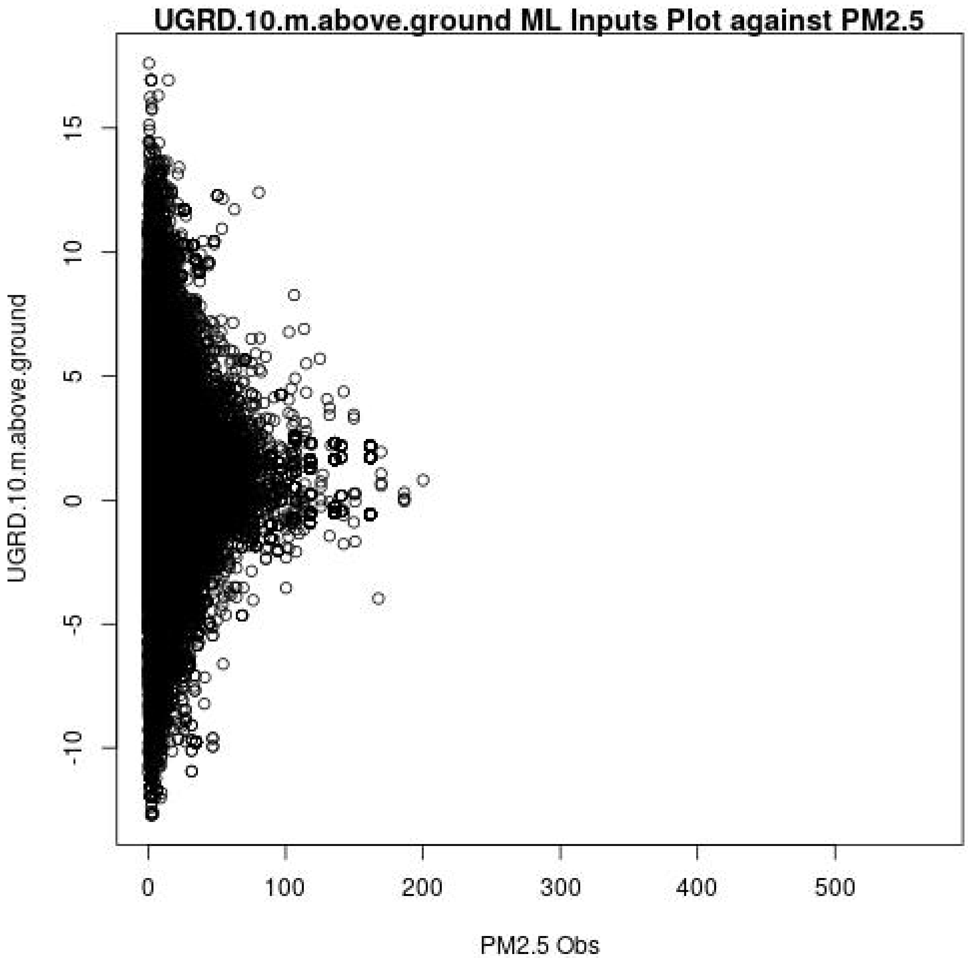

Figure 75: UGRD.10.m.above.ground ML Inputs Plot against PM2.5

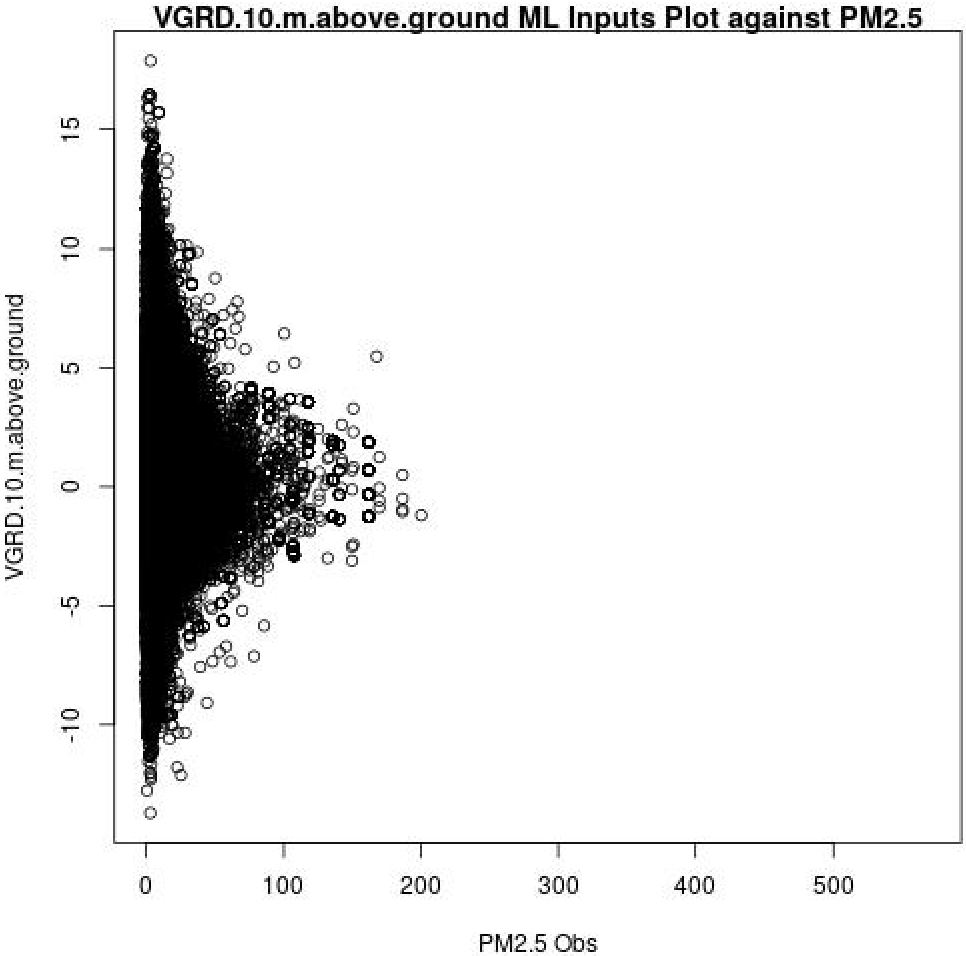

Figure 76: VGRD.10.m.above.ground ML Inputs Plot against PM2.5

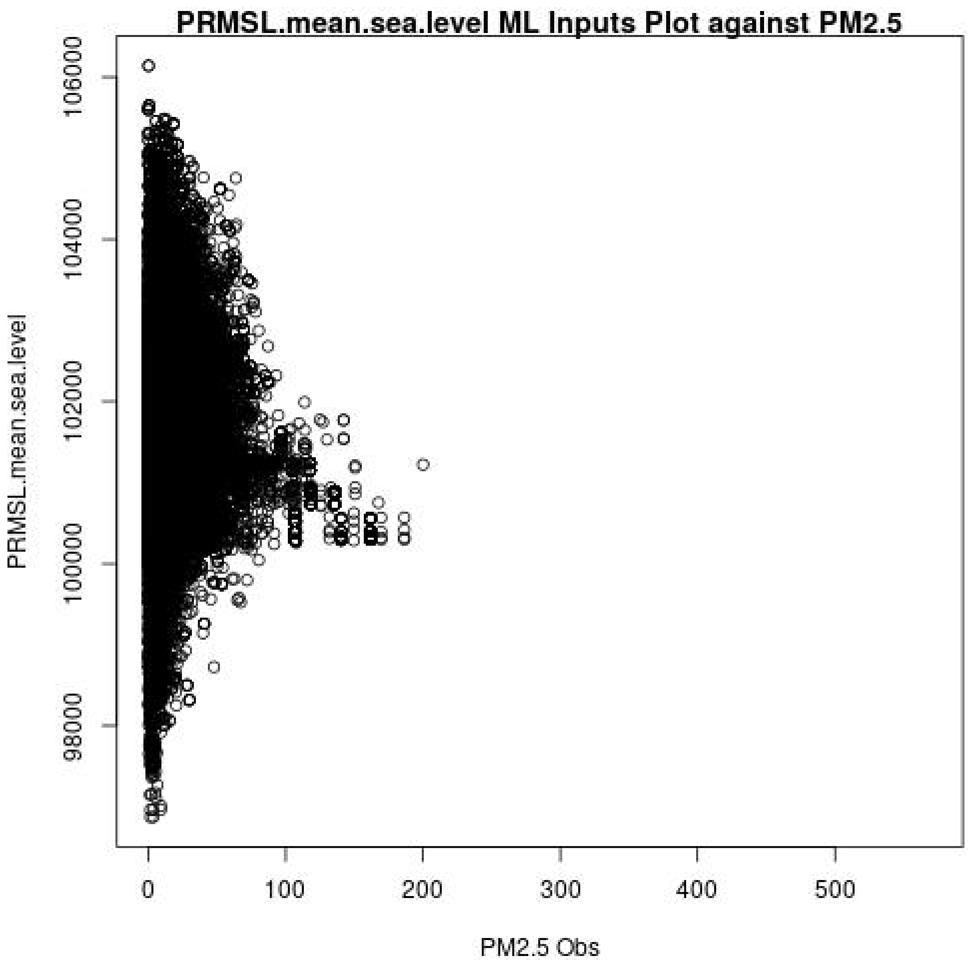

Figure 77: PRMSL.mean.sea.level ML Inputs Plot against PM2.5

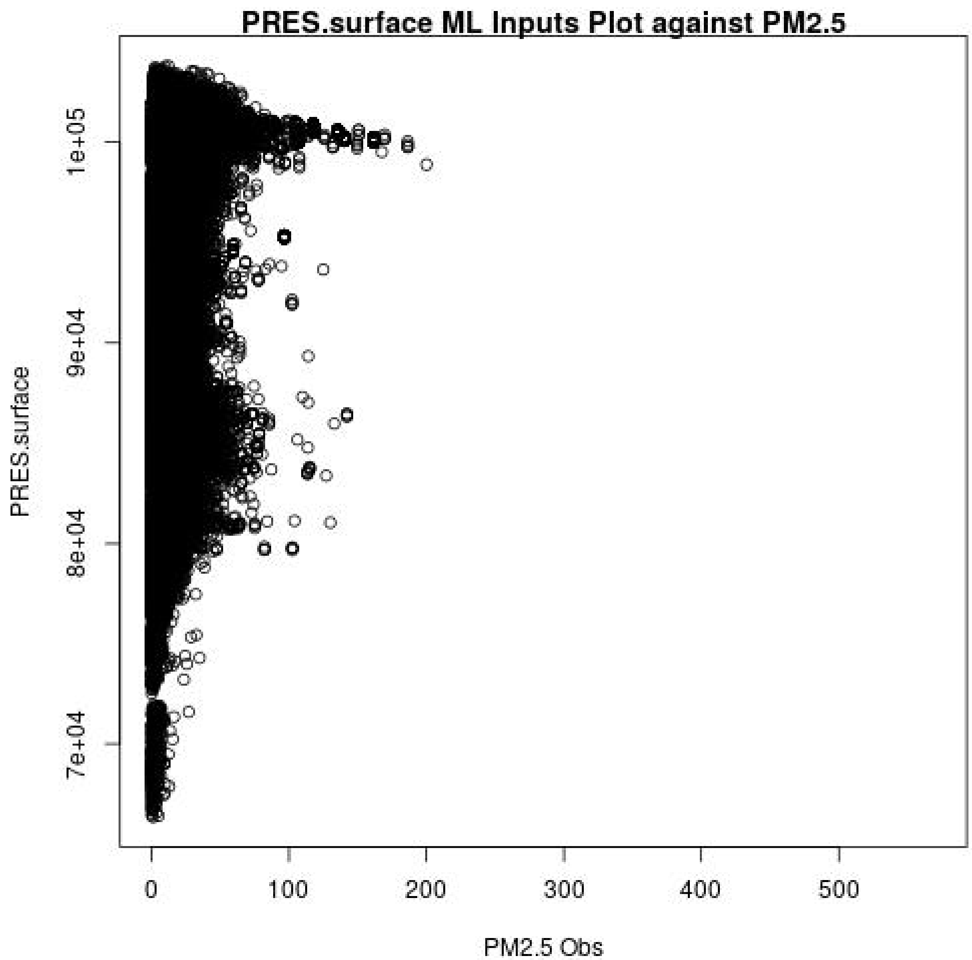

Figure 78: PRES.surface ML Inputs Plot against PM2.5

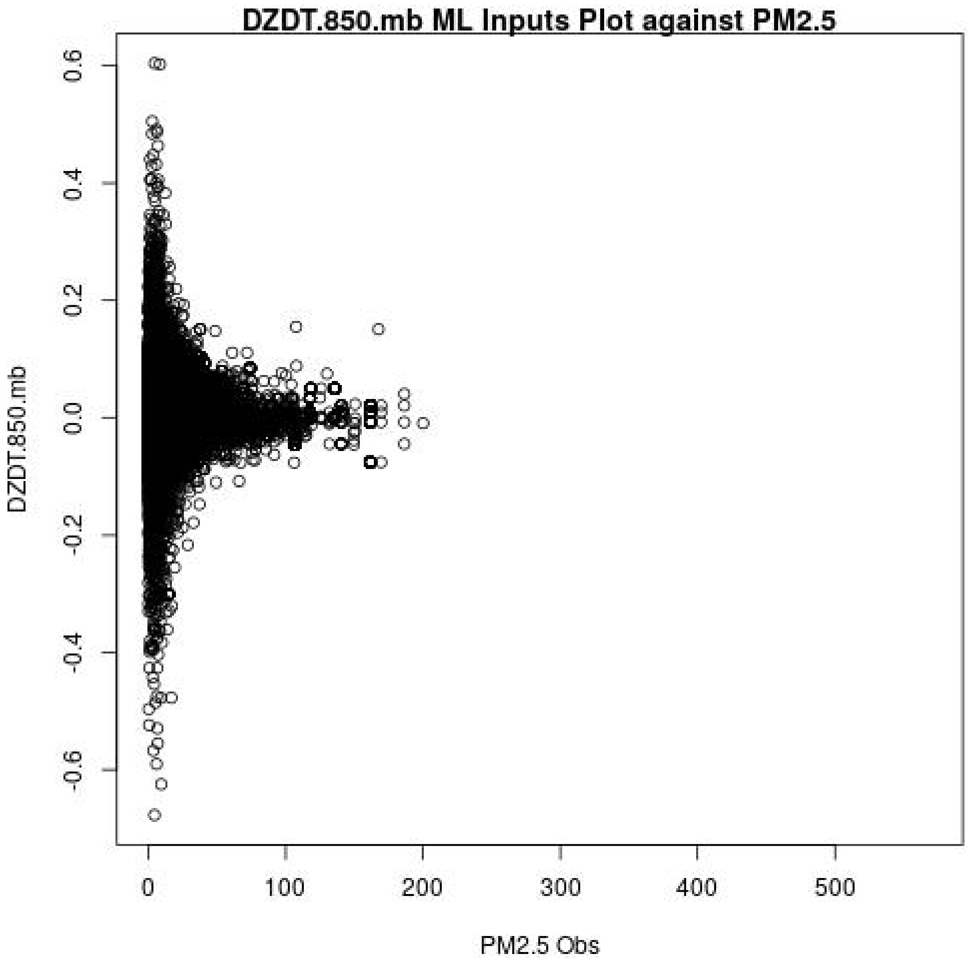

Figure 79: DZDT.850.mb ML Inputs Plot against PM2.5

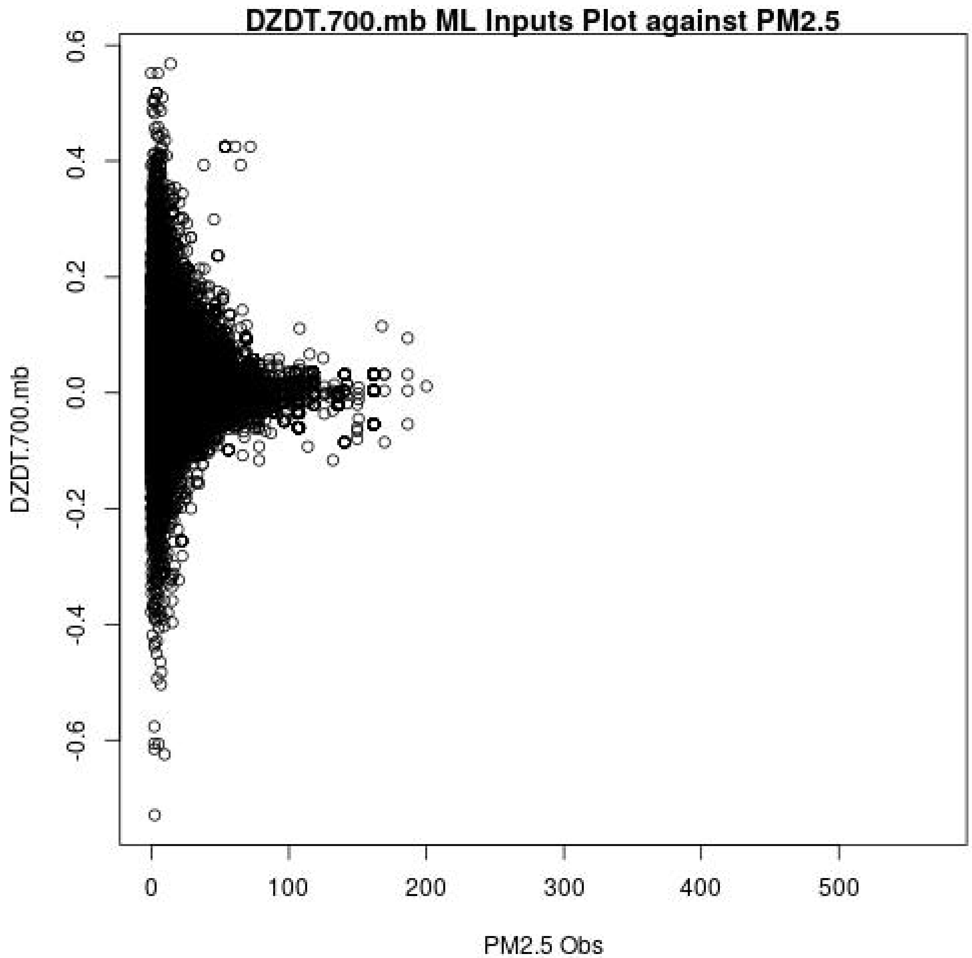

Figure 80: DZDT.700.mb ML Inputs Plot against PM2.5

10.4 ML Inputs Map subset of days Images

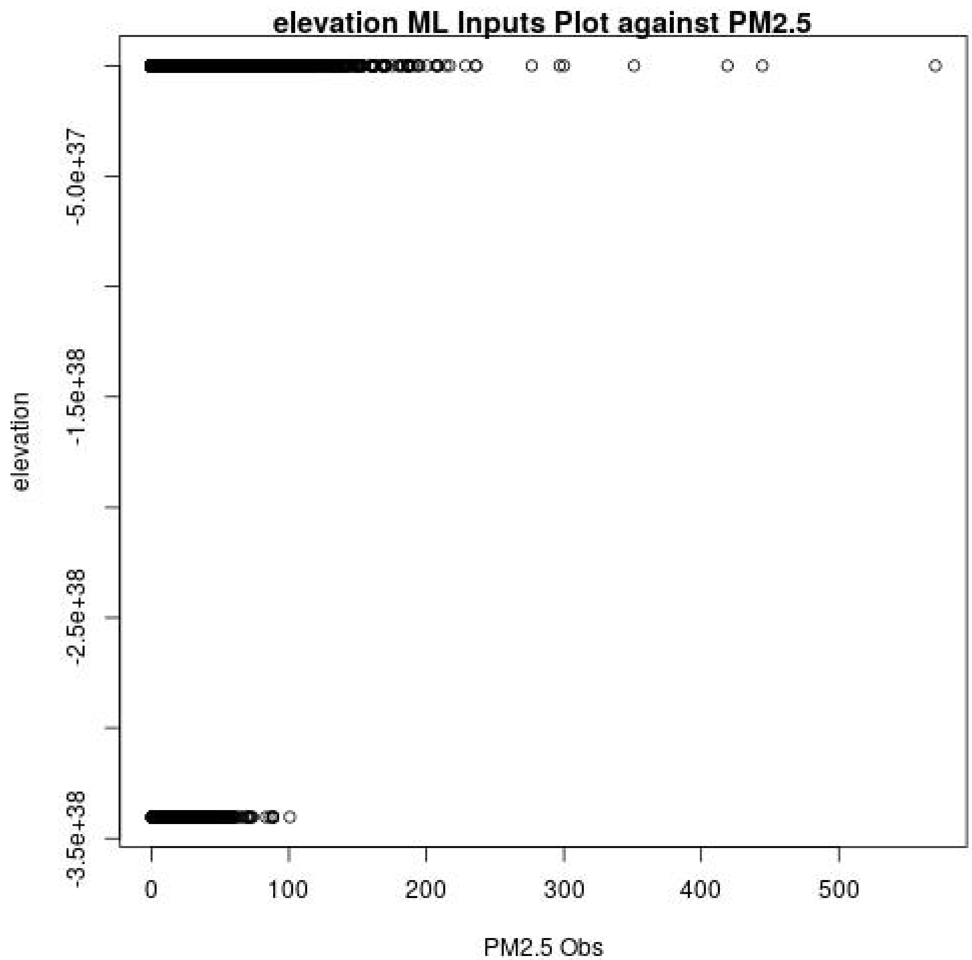

Figure 81: elevation ML Inputs Plot against PM2.5

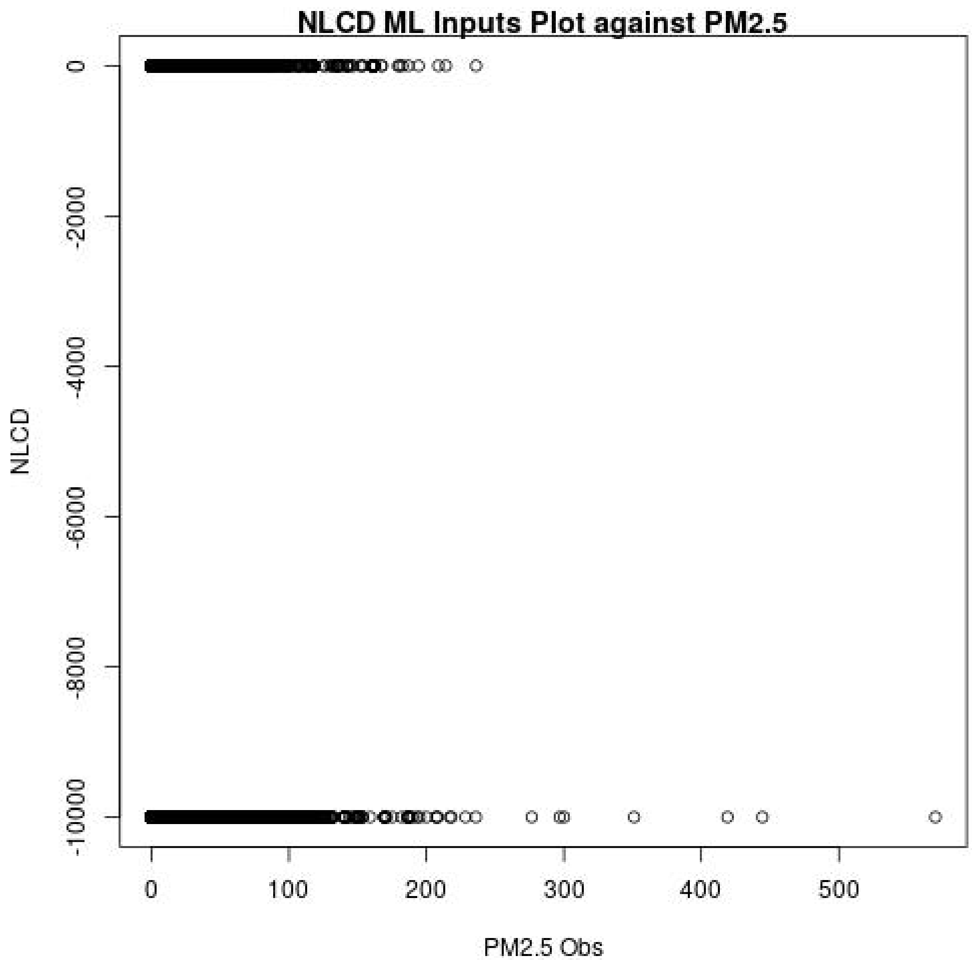

Figure 82: NLCD ML Inputs Plot against PM2.5

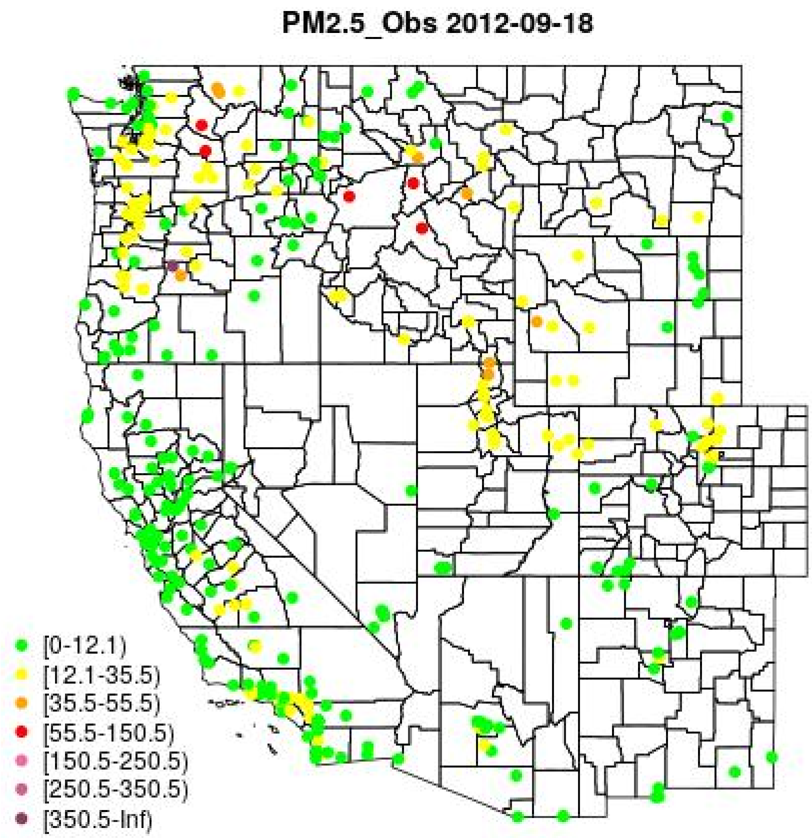

Figure 83: PM2.5-Obs 2012-09-18

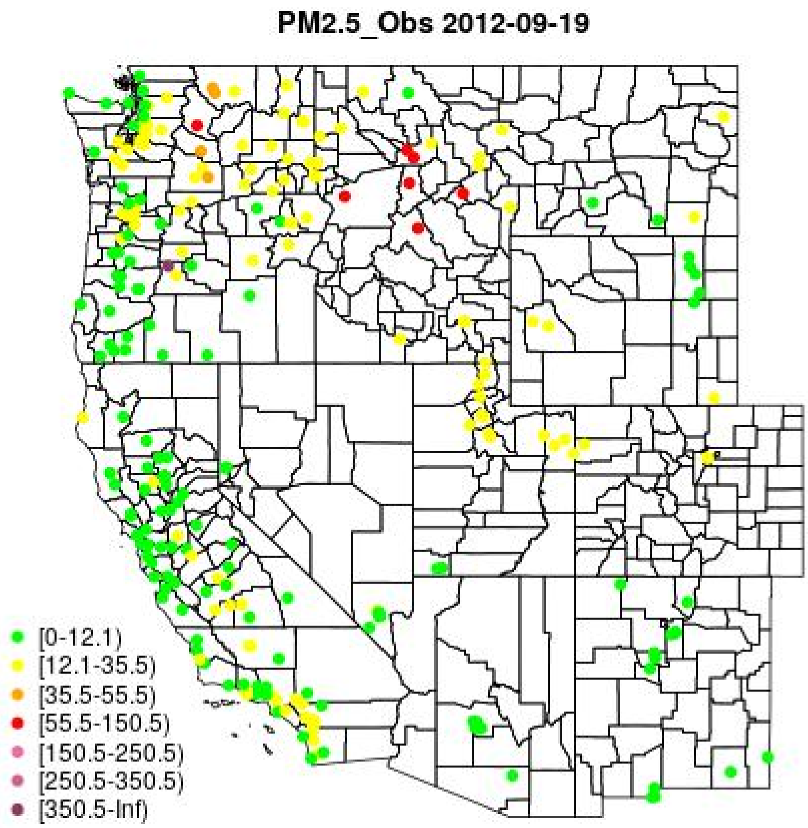

Figure 84: PM2.5-Obs 2012-09-19

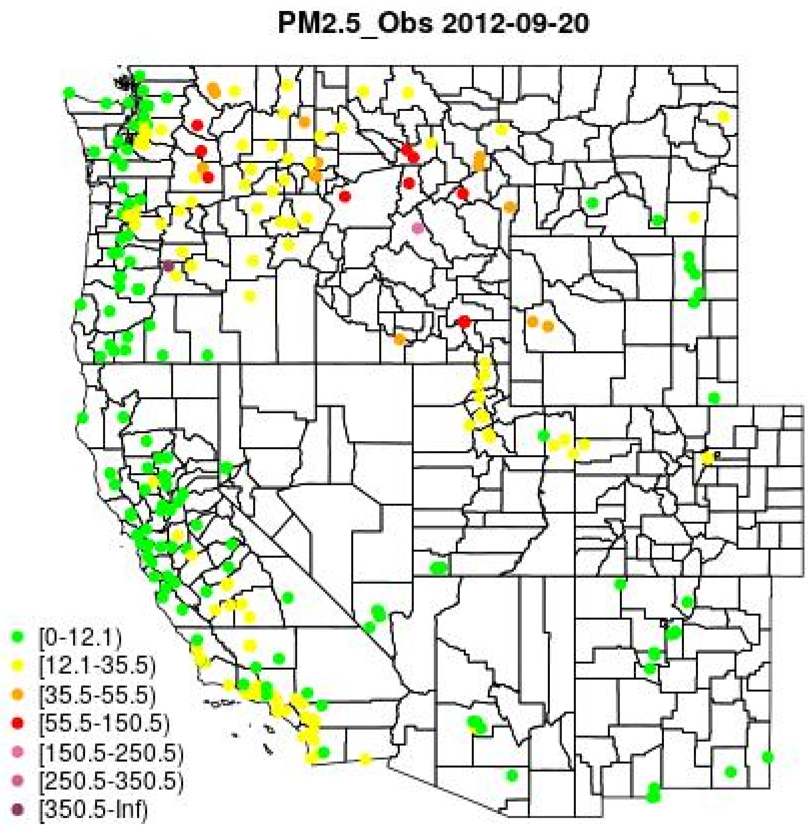

Figure 85: PM2.5-Obs 2012-09-20

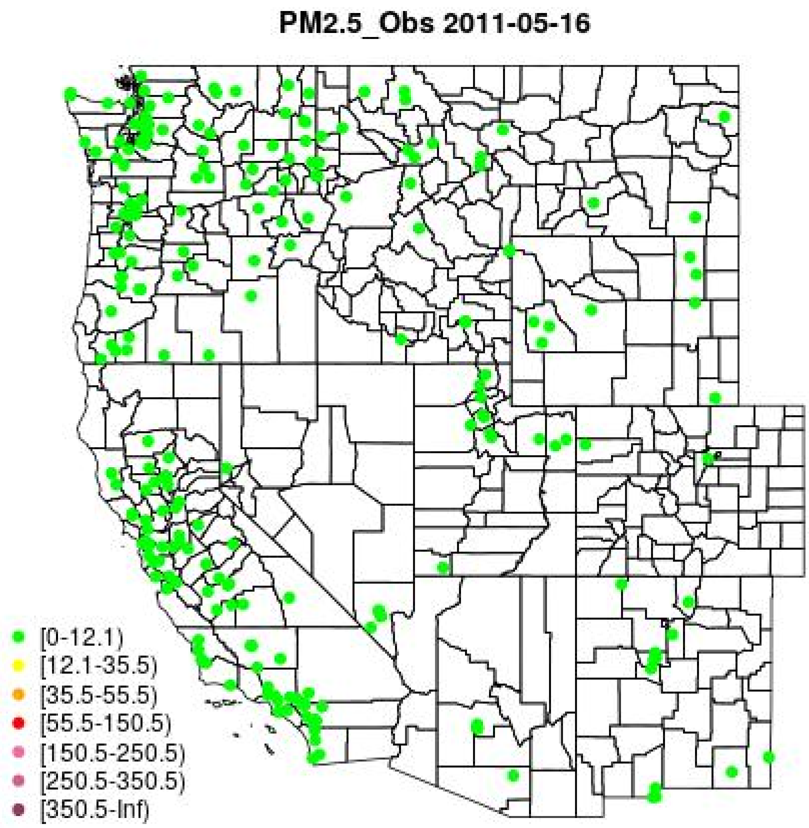

Figure 86: PM2.5-Obs 2011-05-16

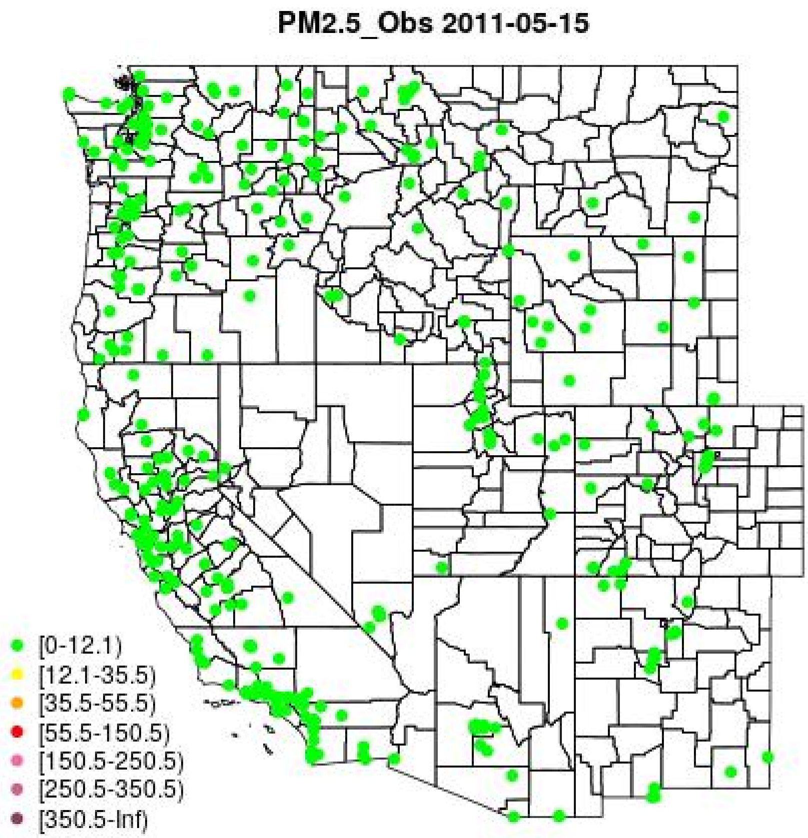

Figure 87: PM2.5-Obs 2011-05-15

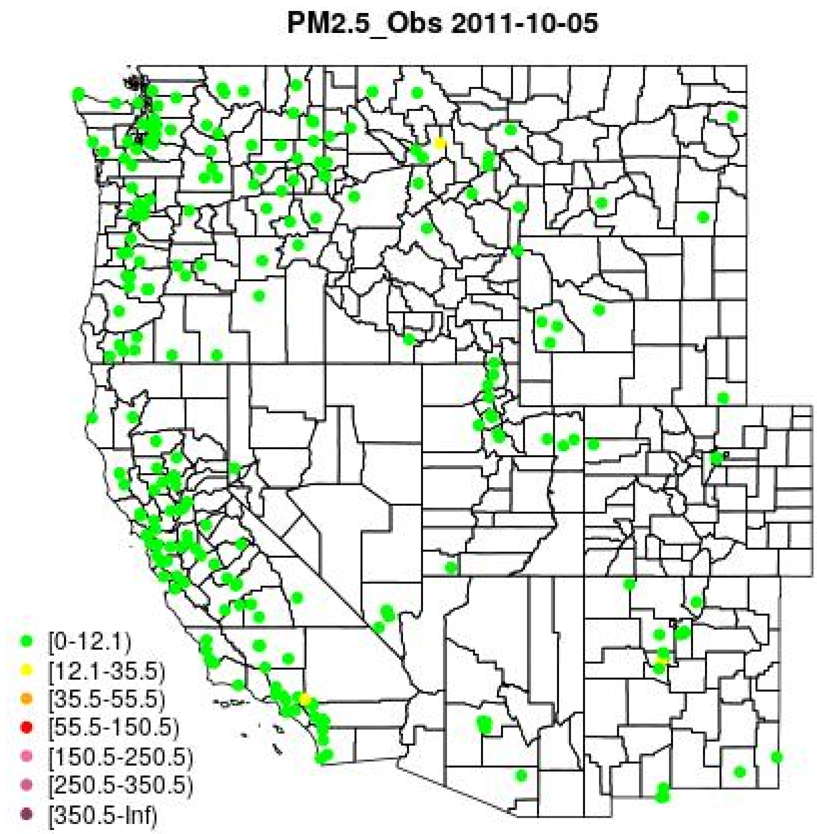

Figure 88: PM2.5-Obs 2011-10-05

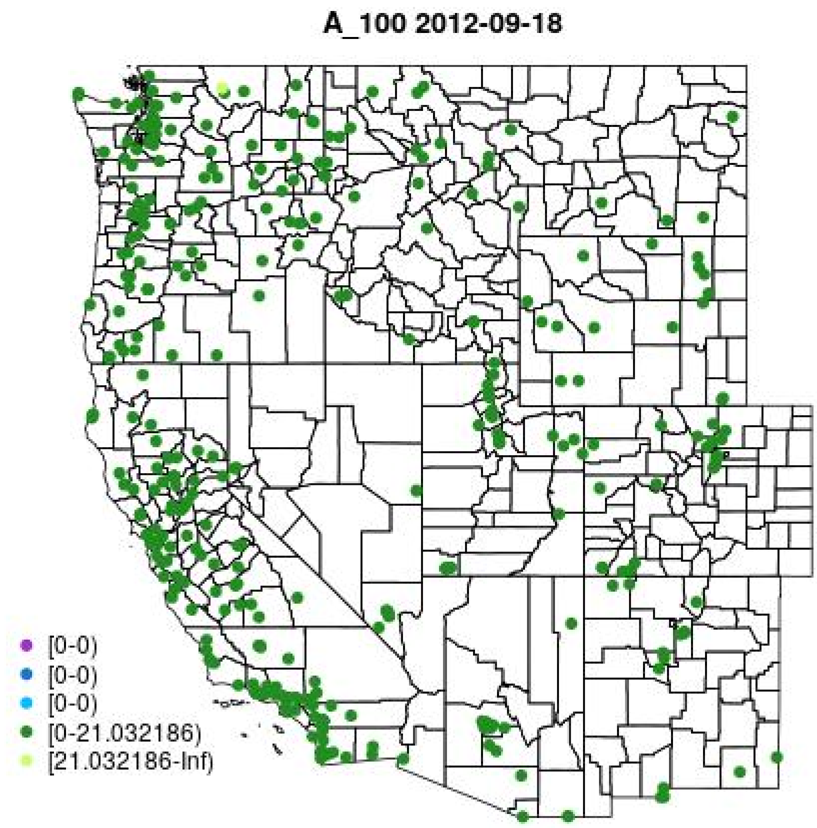

Figure 89: A-100 2012-09-18

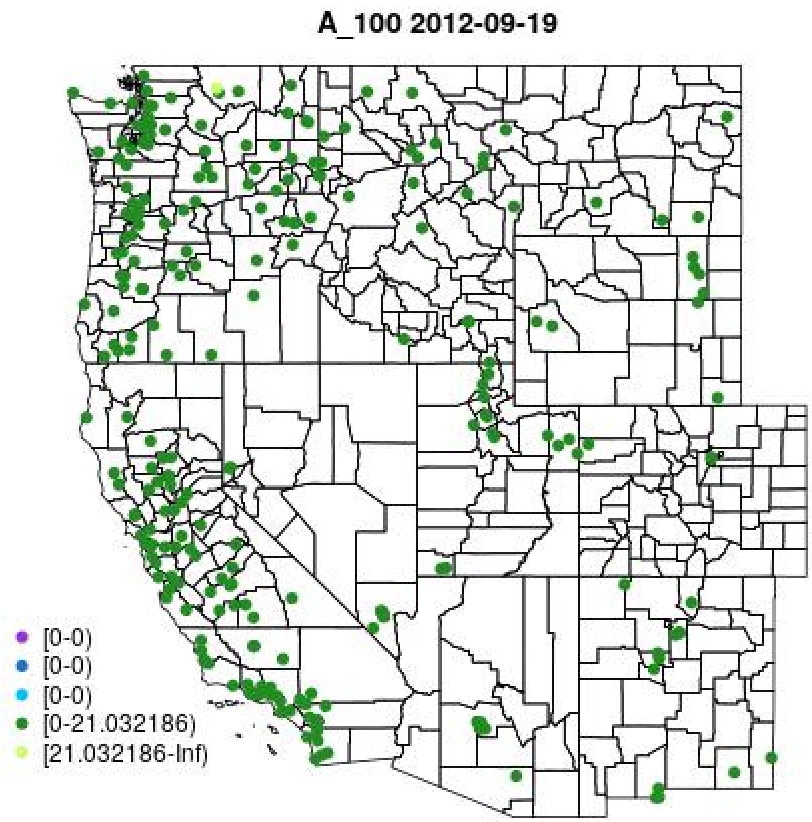

Figure 90: A-100 2012-09-19

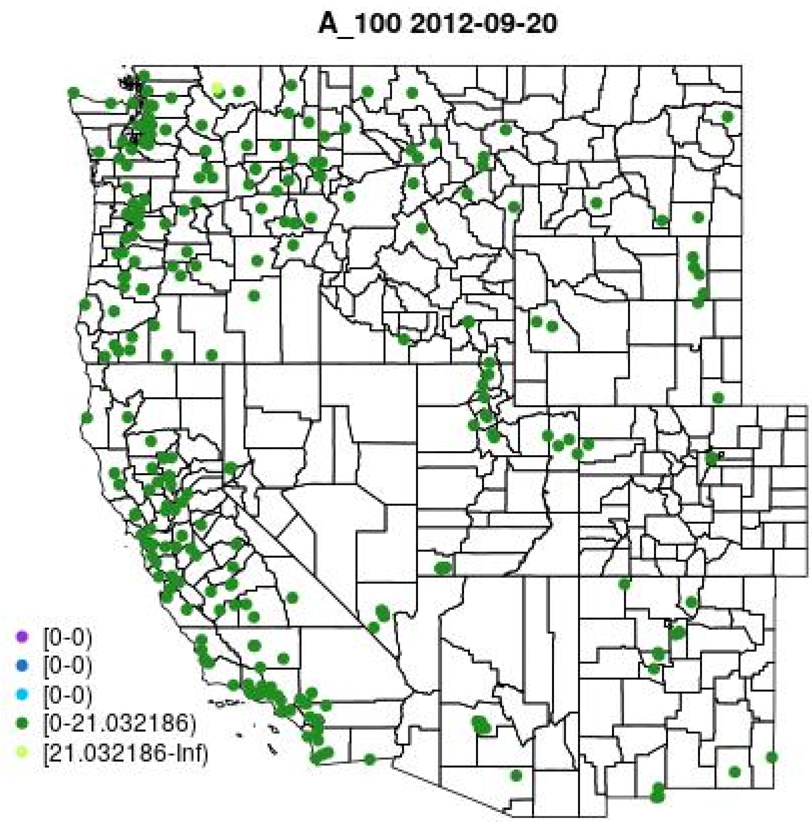

Figure 91: A-100 2012-09-20

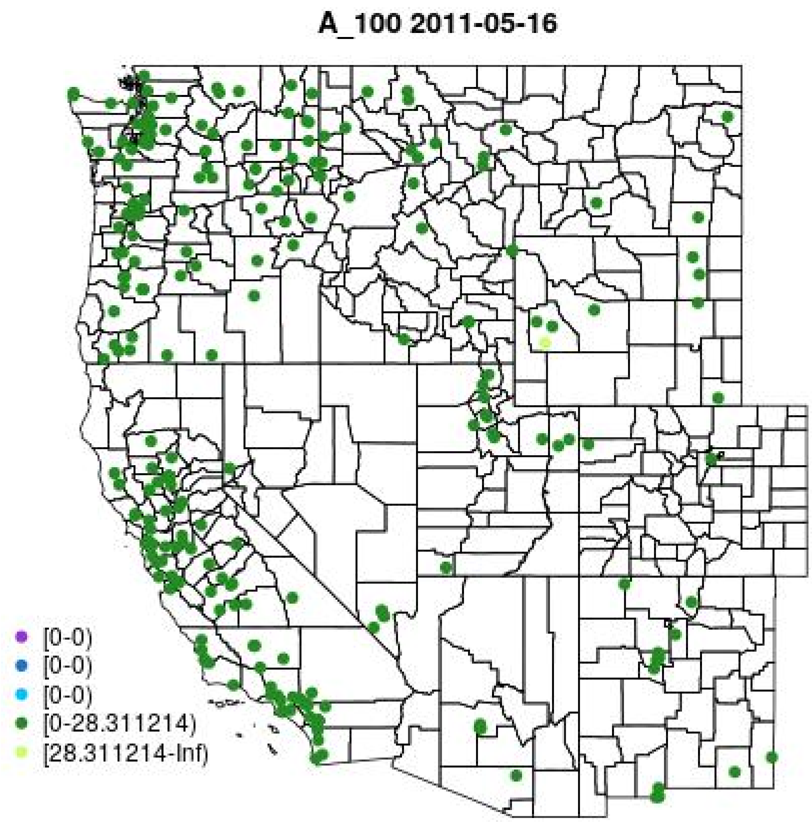

Figure 92: A-100 2011-05-16

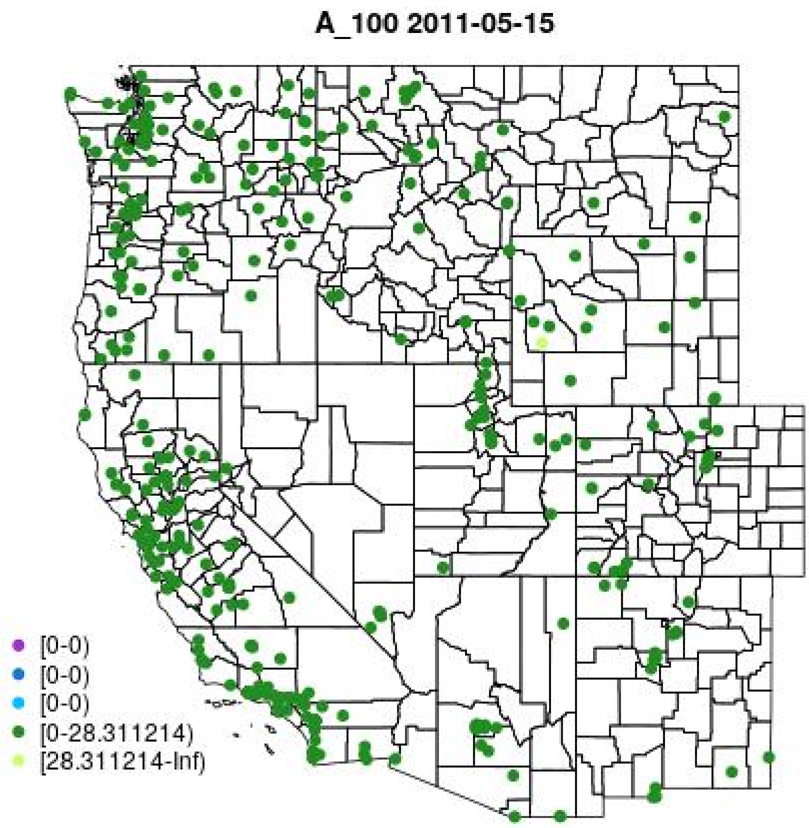

Figure 93: A-100 2011-05-15

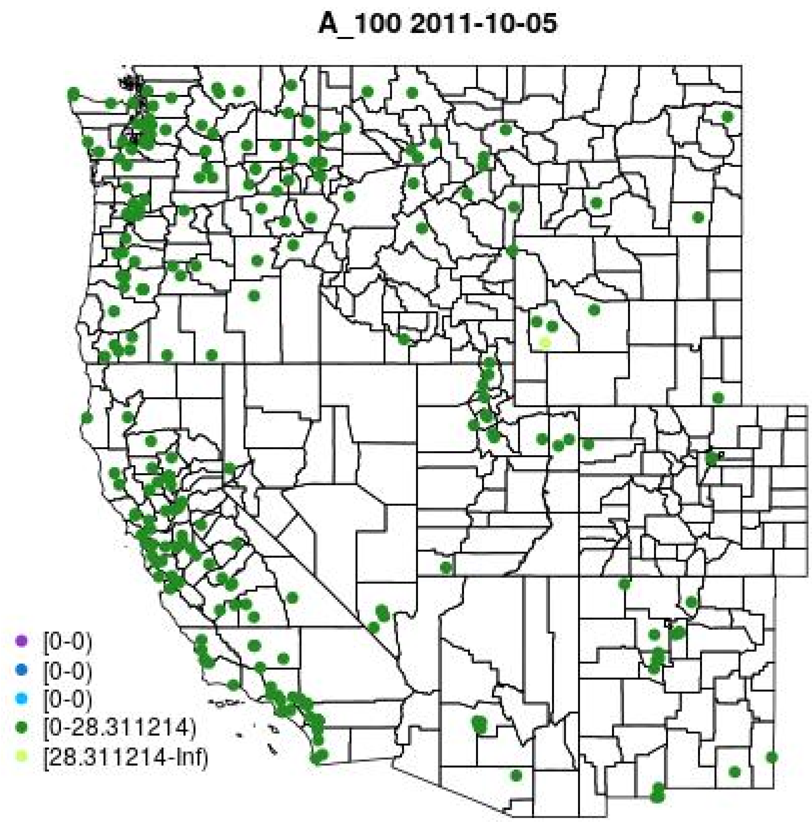

Figure 94: A-100 2011-10-05

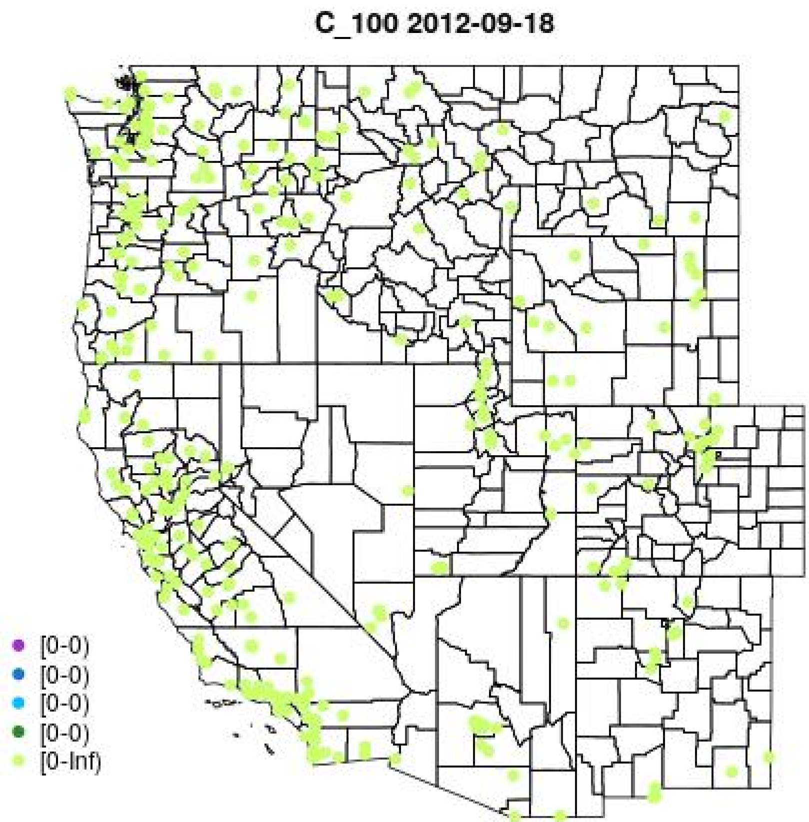

Figure 95: C-100 2012-09-18

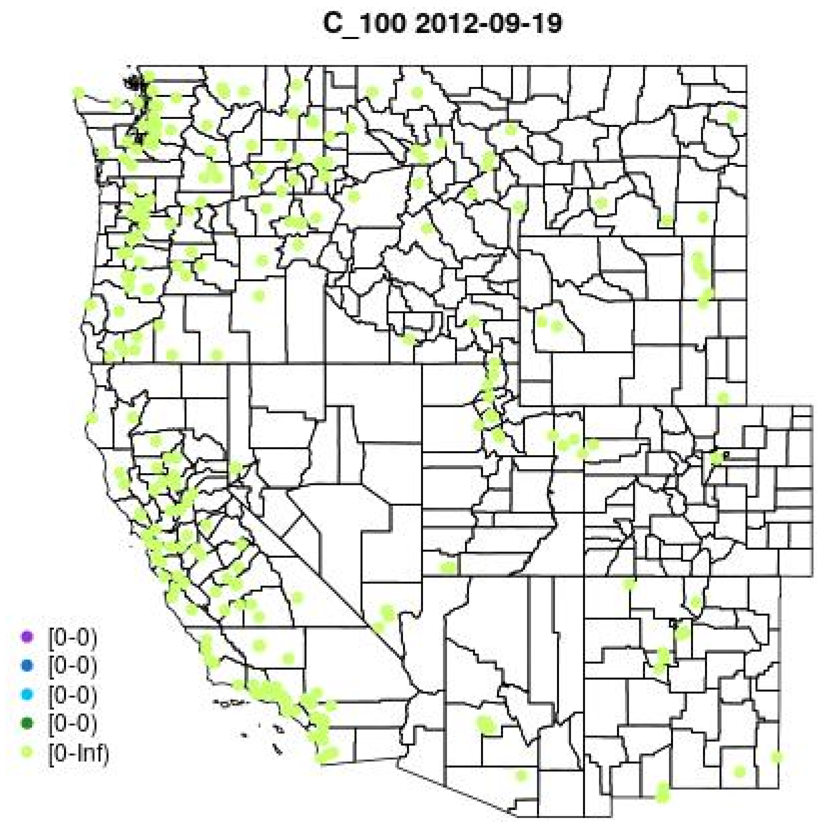

Figure 96: C-100 2012-09-19

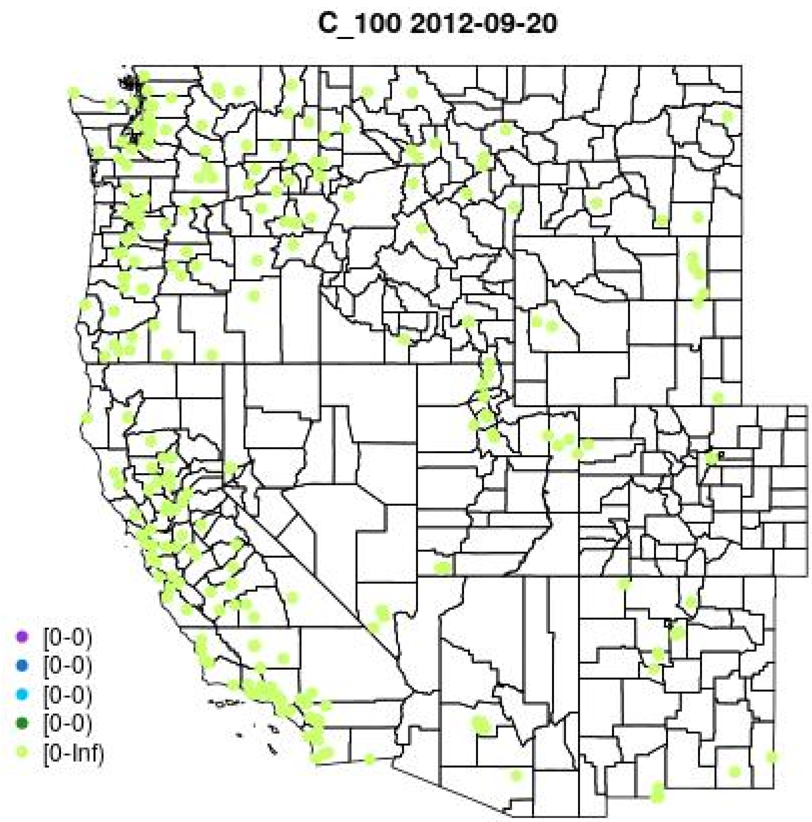

Figure 97: C-100 2012-09-20

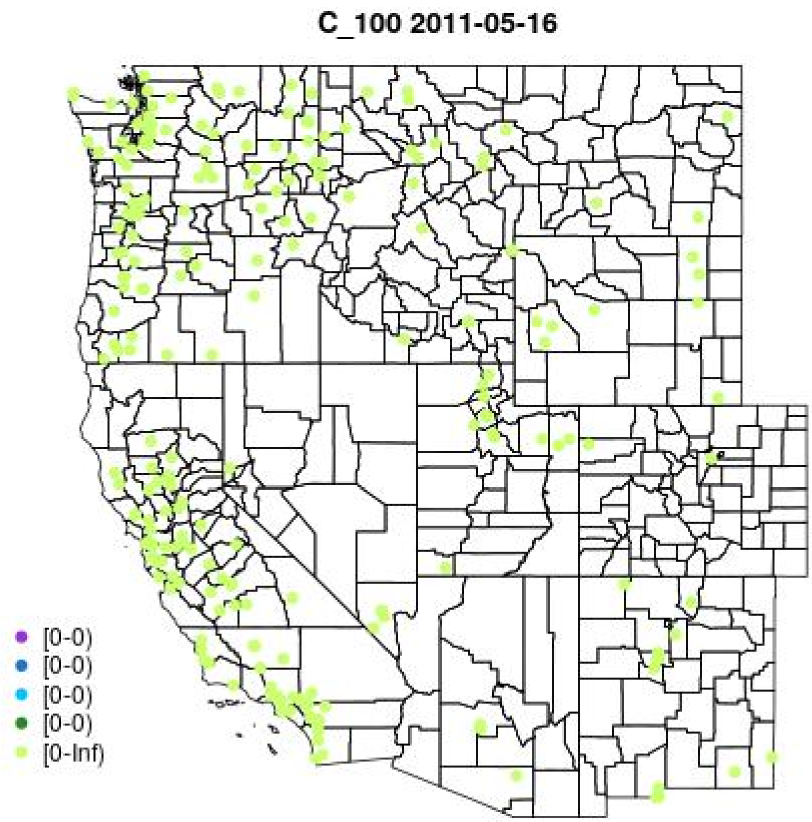

Figure 98: C-100 2011-05-16

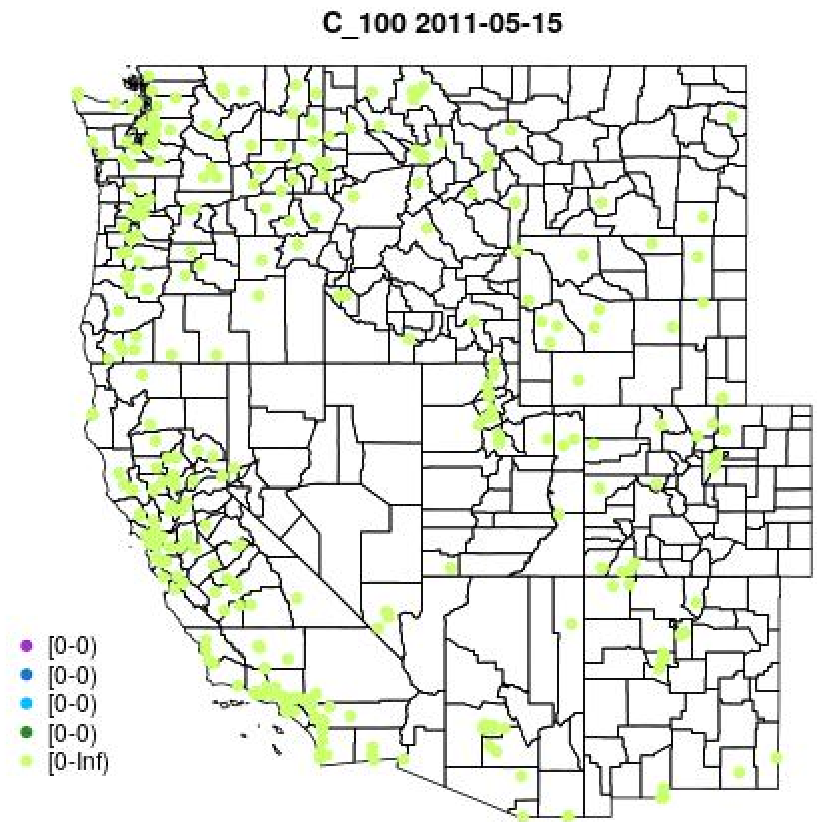

Figure 99: C-100 2011-05-15

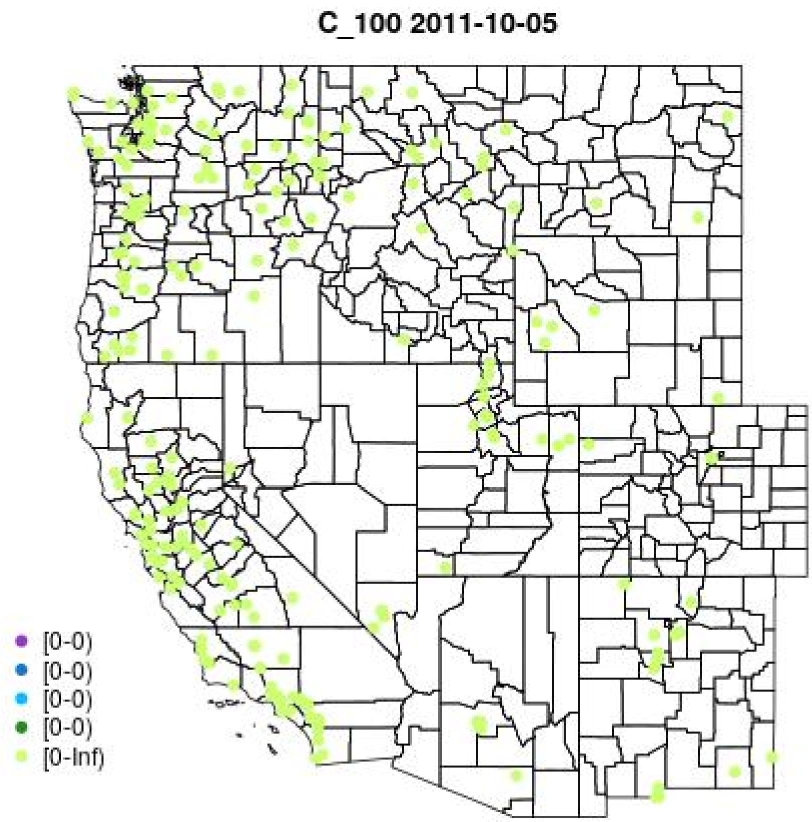

Figure 100: C-100 2011-10-05

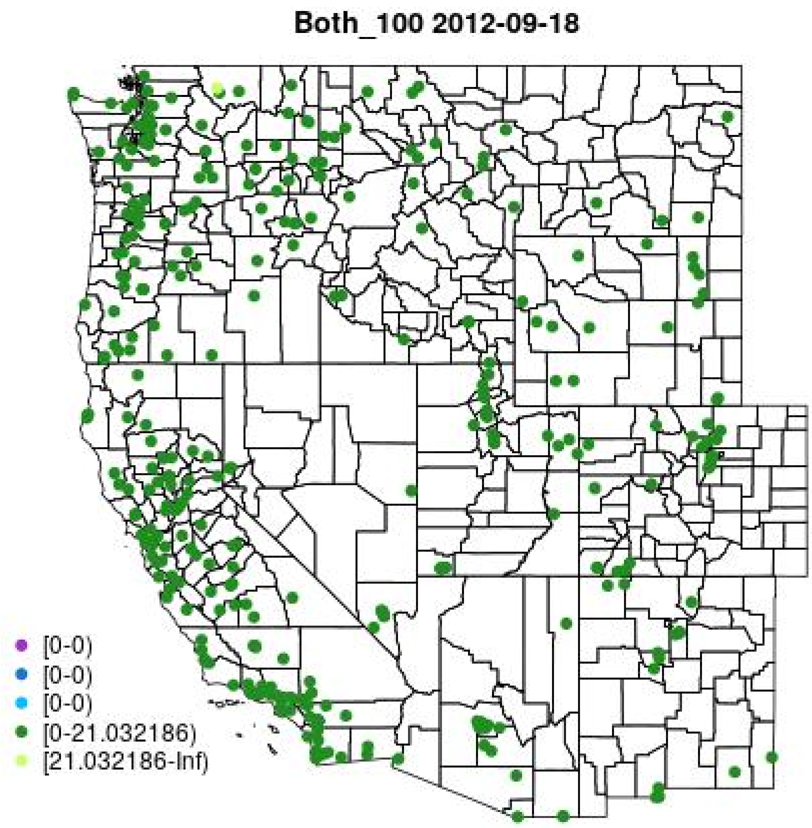

Figure 101: Both-100 2012-09-18

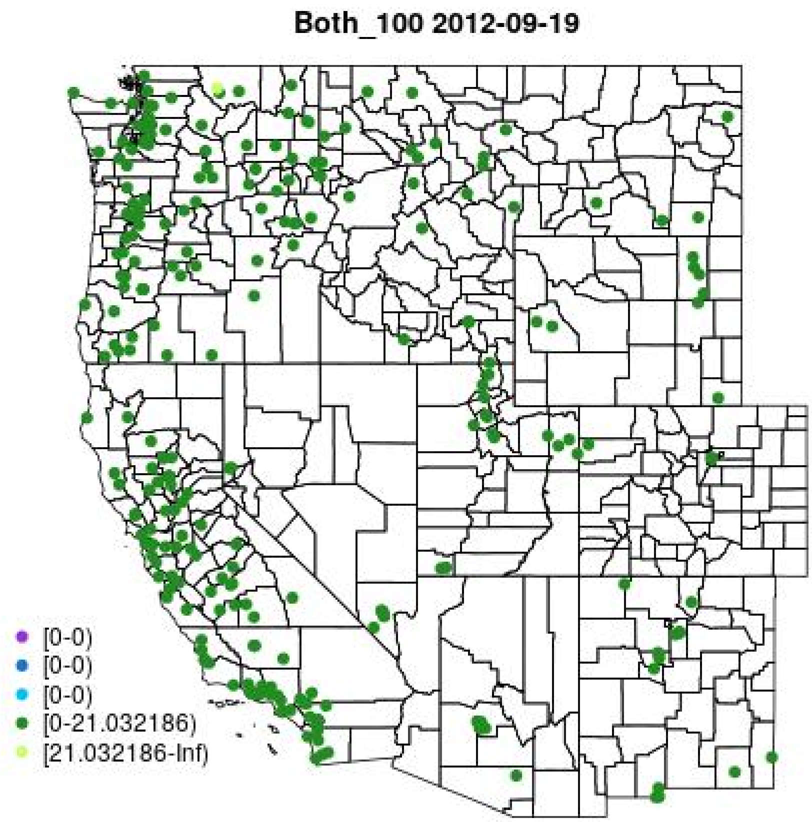

Figure 102: Both-100 2012-09-19

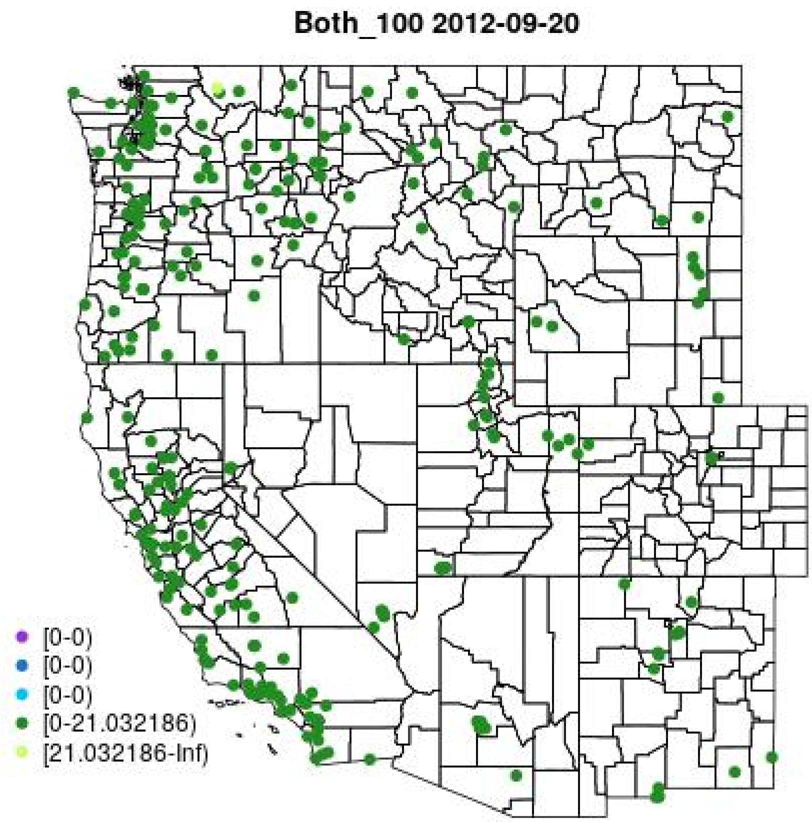

Figure 103: Both-100 2012-09-20

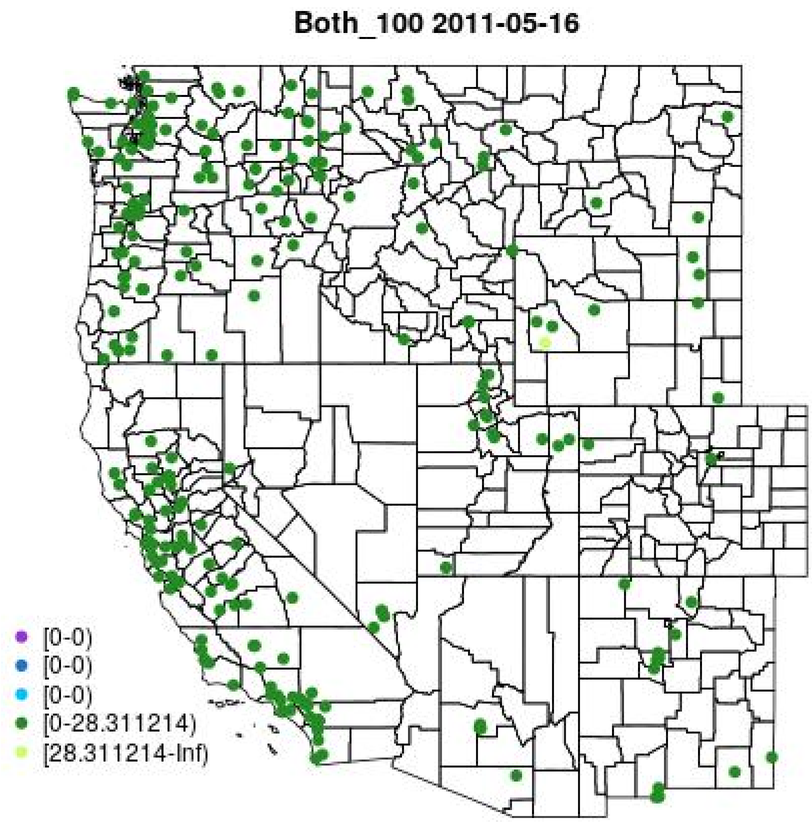

Figure 104: Both-100 2011-05-16

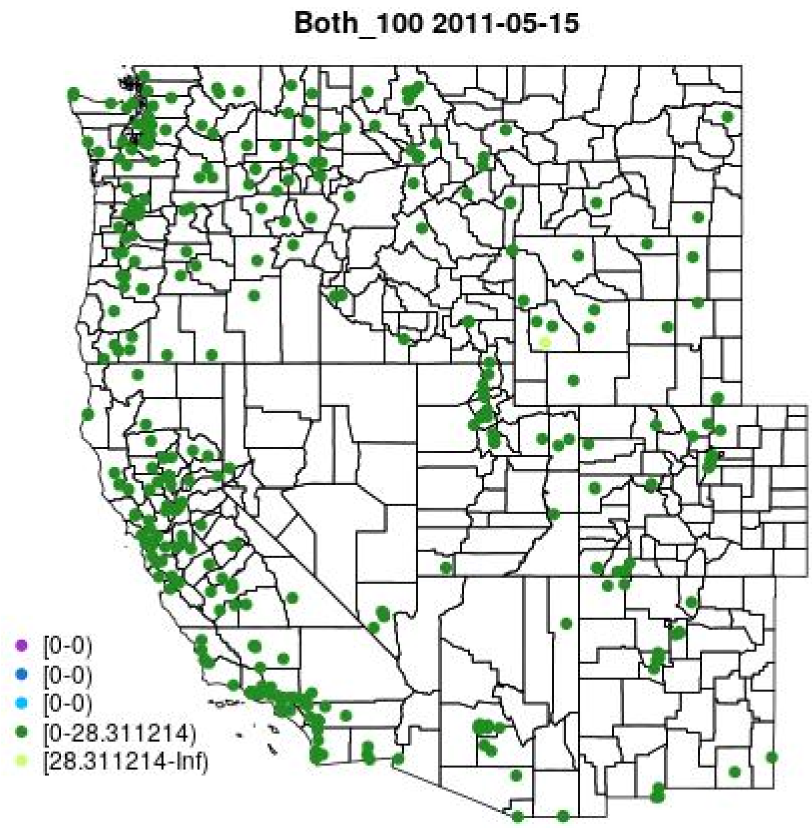

Figure 105: Both-100 2011-05-15

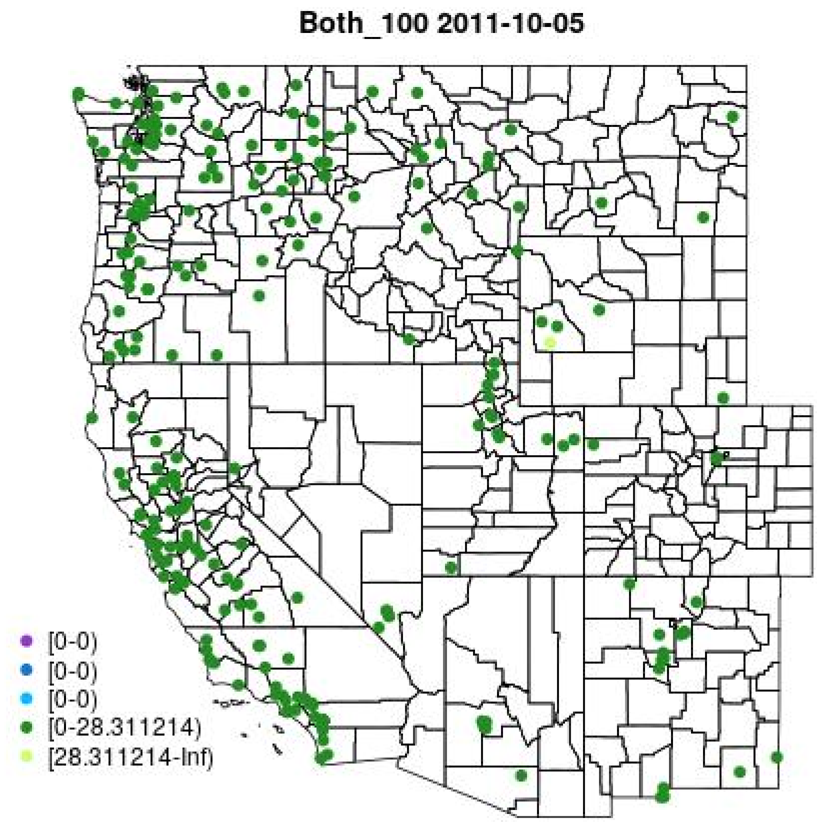

Figure 106: Both-100 2011-10-05

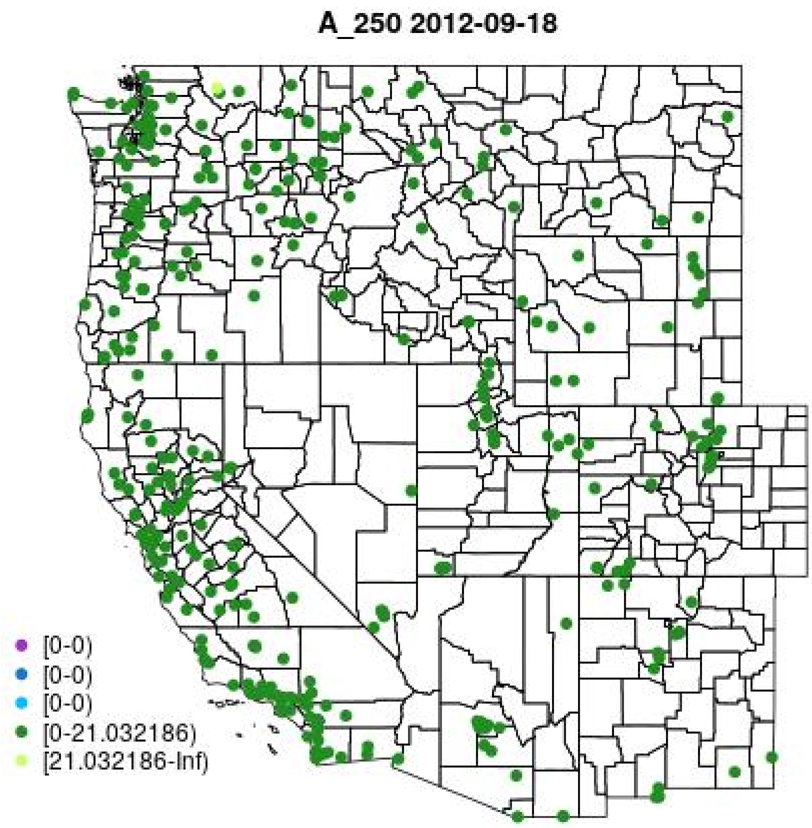

Figure 107: A-250 2012-09-18

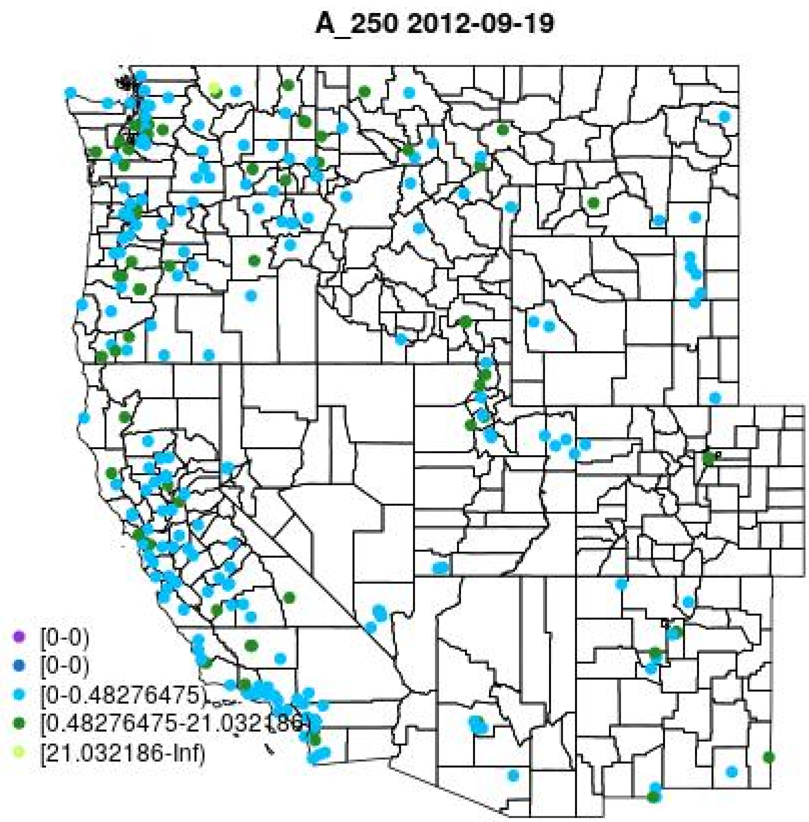

Figure 108: A-250 2012-09-19

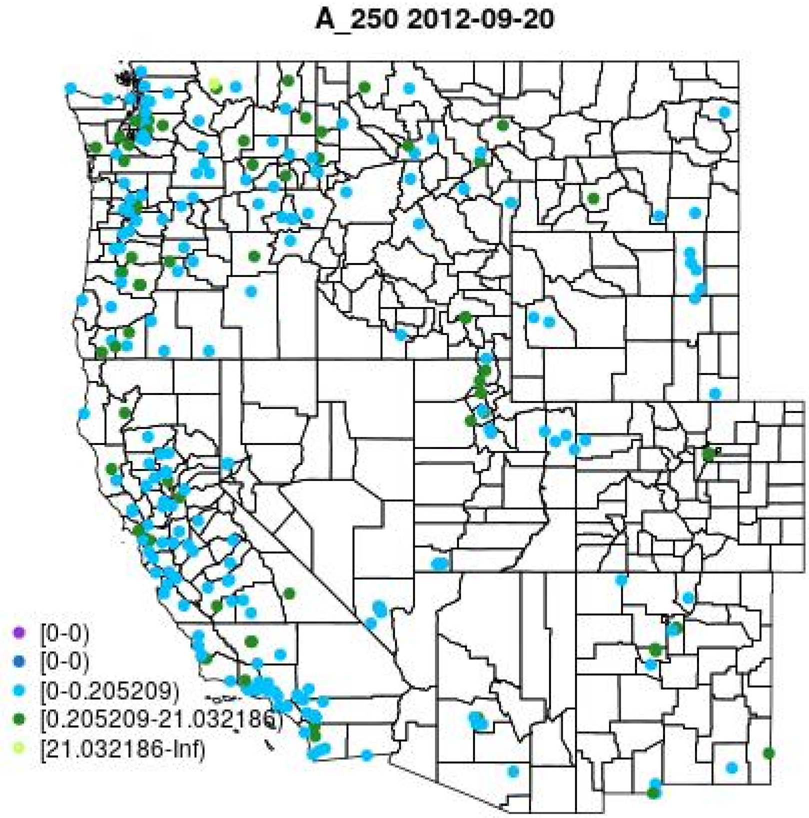

Figure 109: A-250 2012-09-20

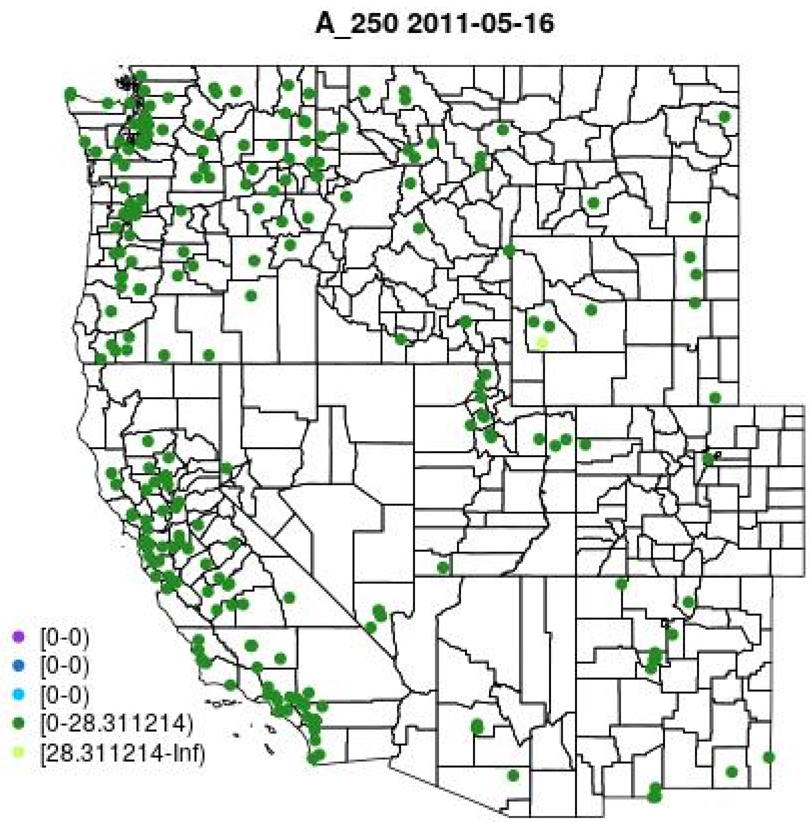

Figure 110: A-250 2011-05-16

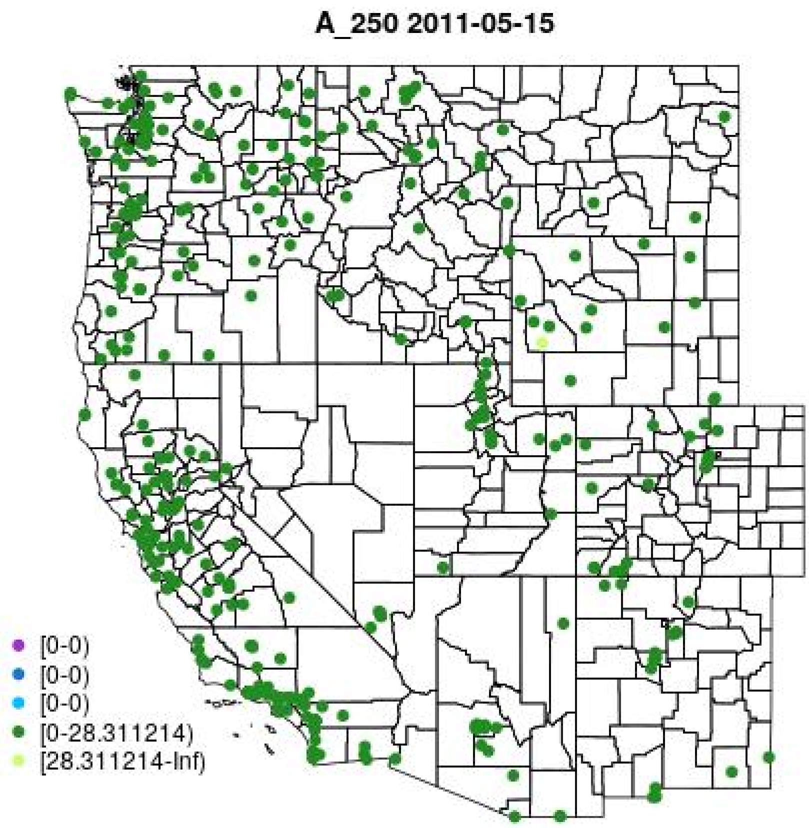

Figure 111: A-250 2011-05-15

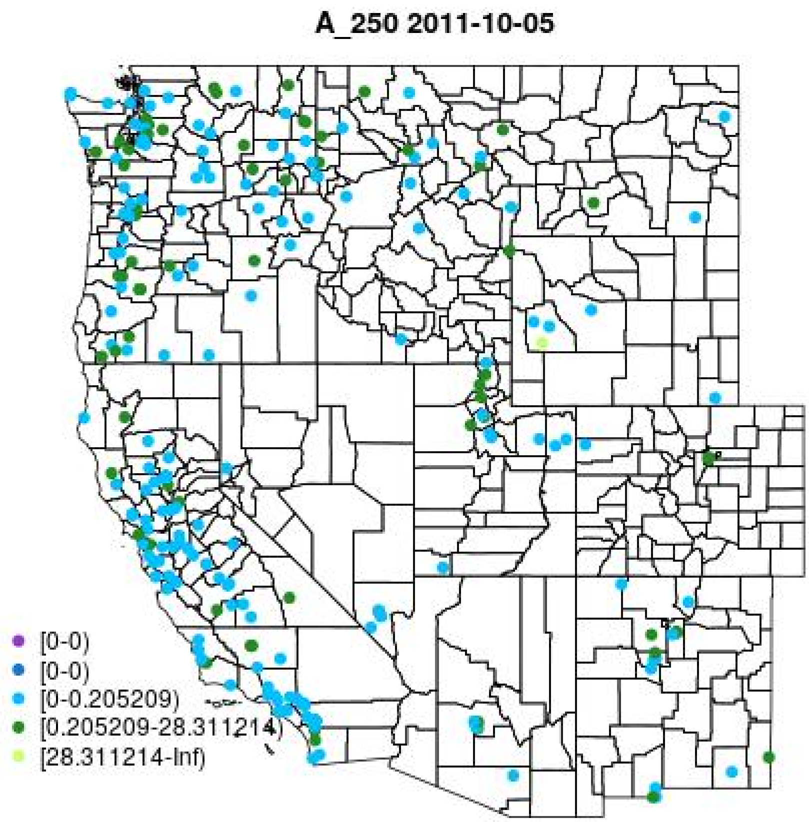

Figure 112: A-250 2011-10-05

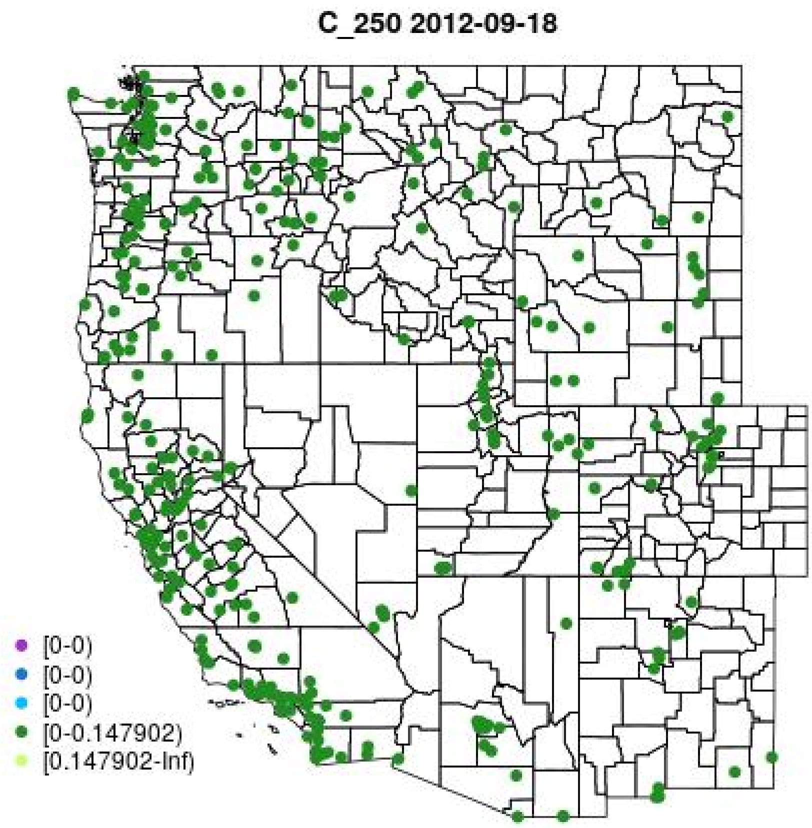

Figure 113: C-250 2012-09-18

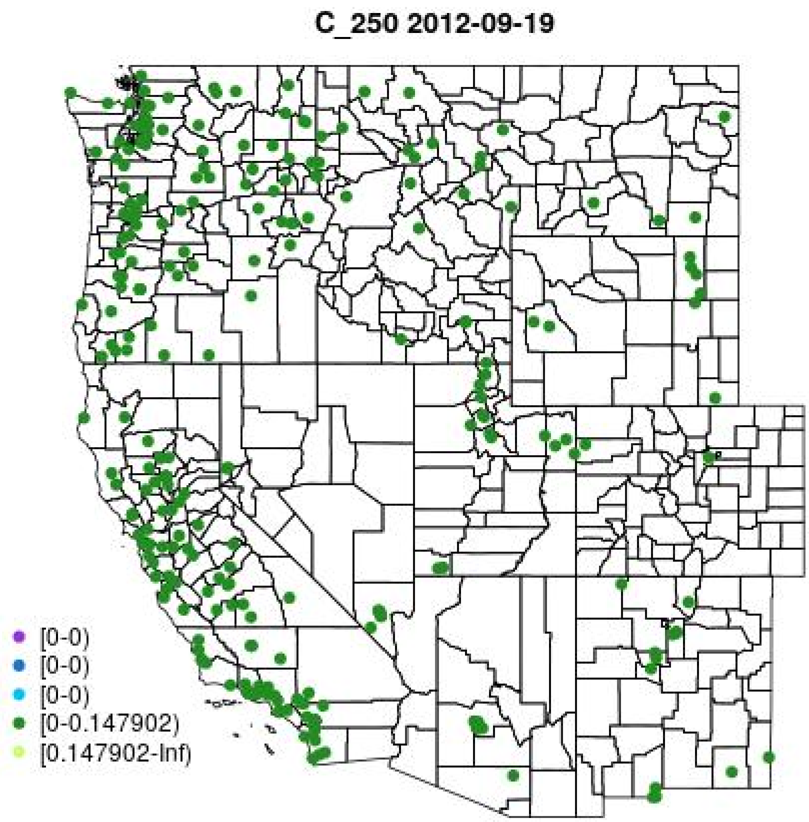

Figure 114: C-250 2012-09-19

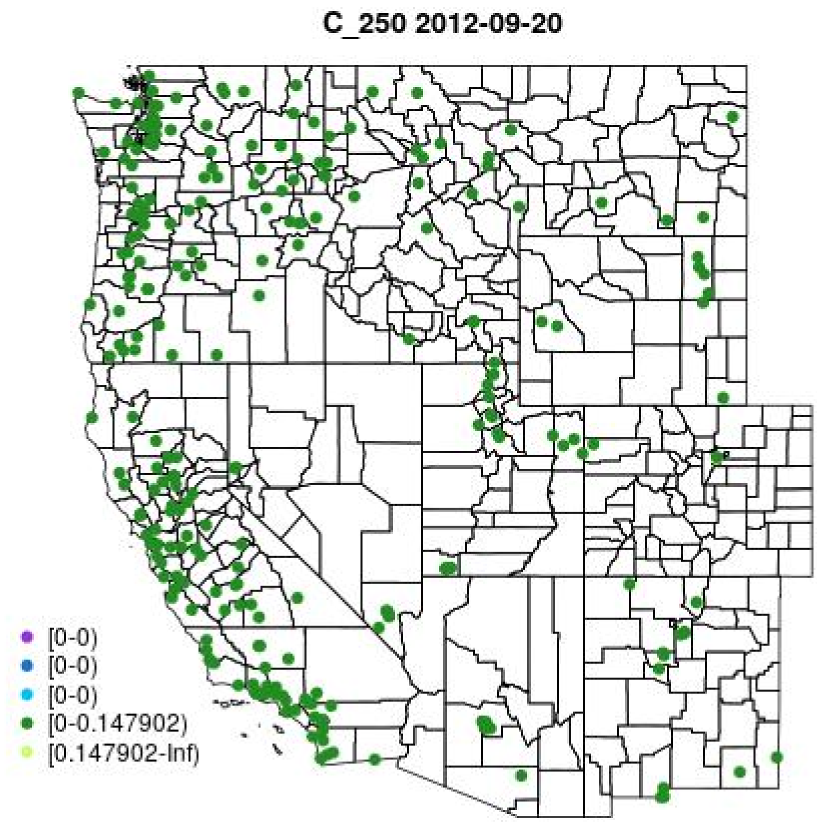

Figure 115: C-250 2012-09-20

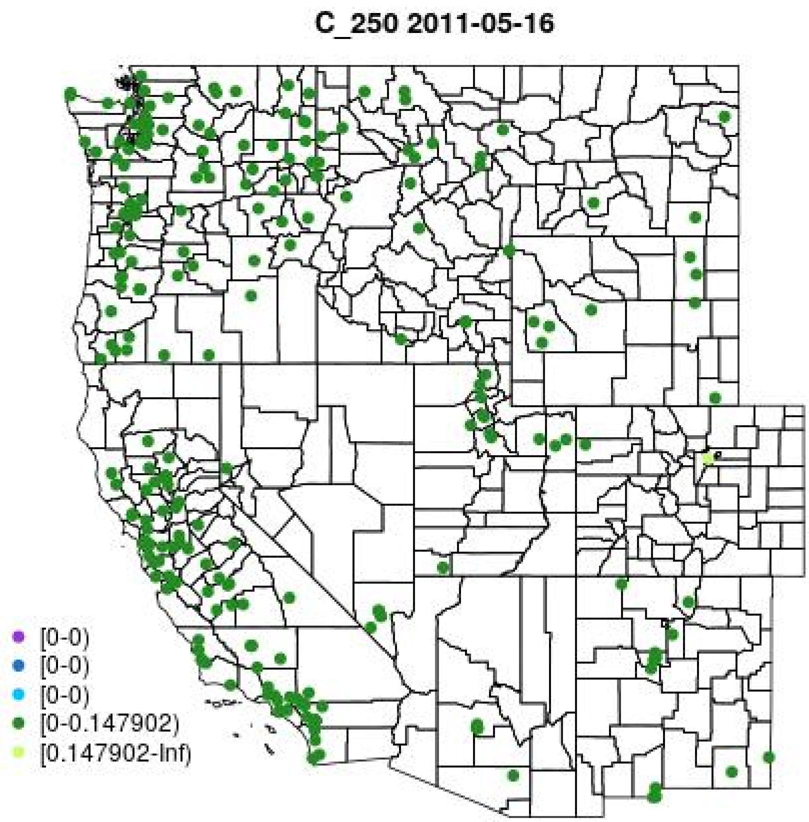

Figure 116: C-250 2011-05-16

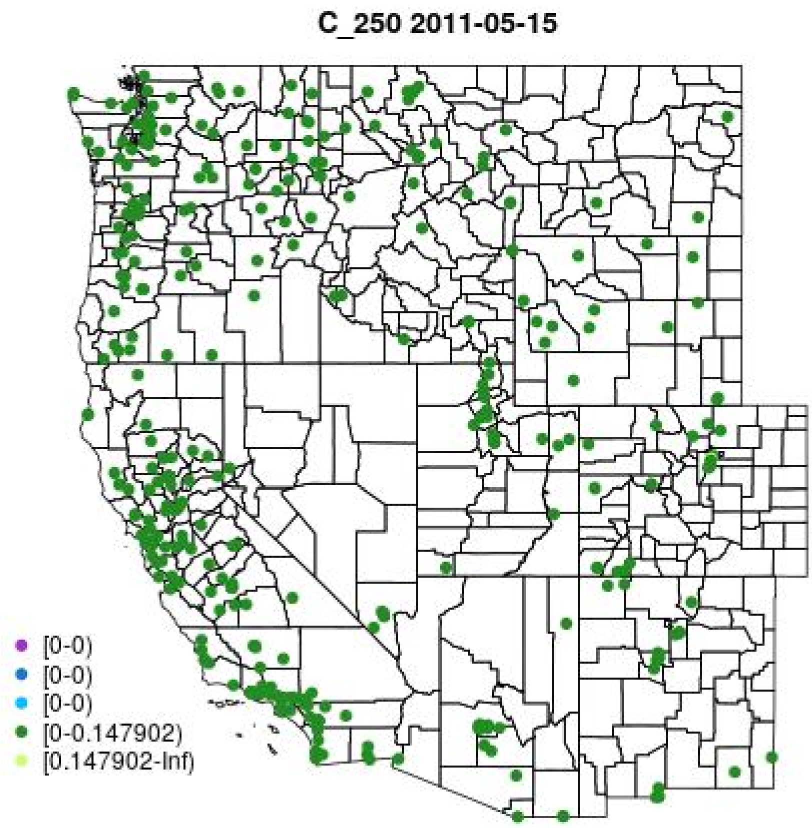

Figure 117: C-250 2011-05-15

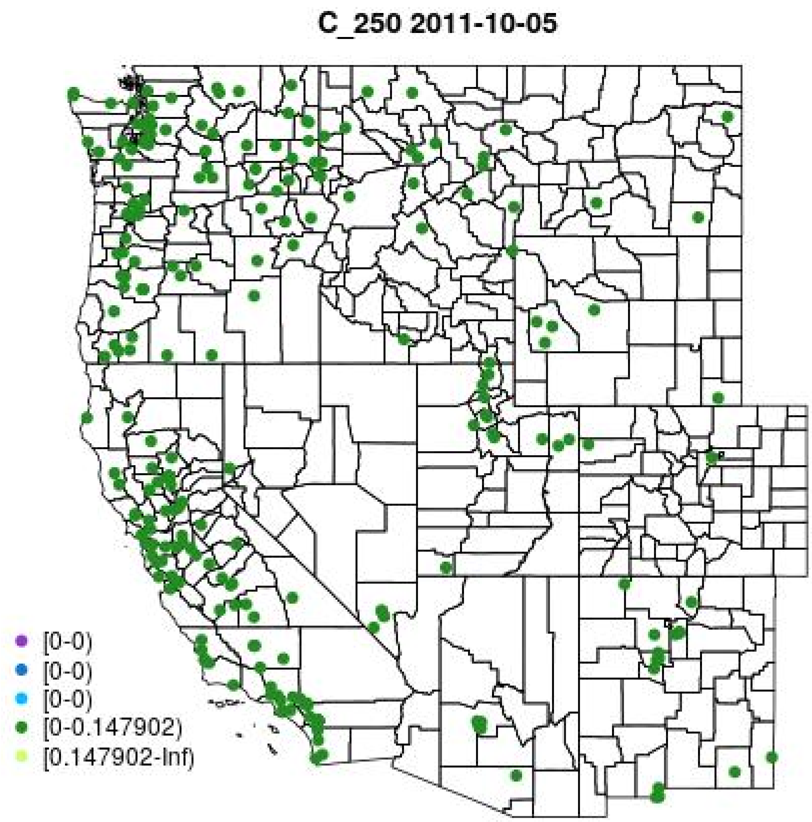

Figure 118: C-250 2011-10-05

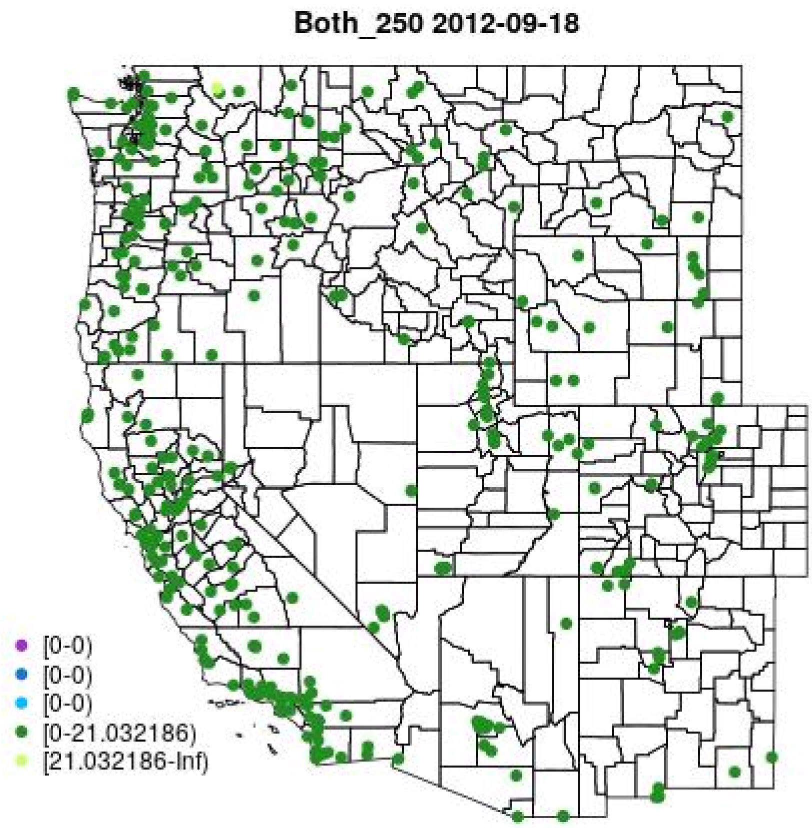

Figure 119: Both-250 2012-09-18

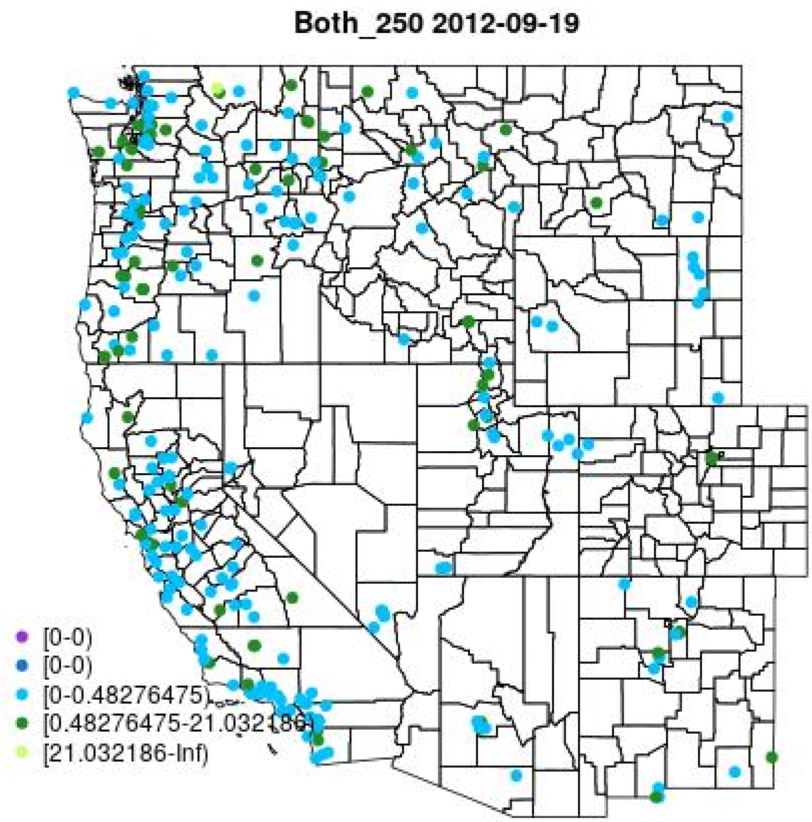

Figure 120: Both-250 2012-09-19

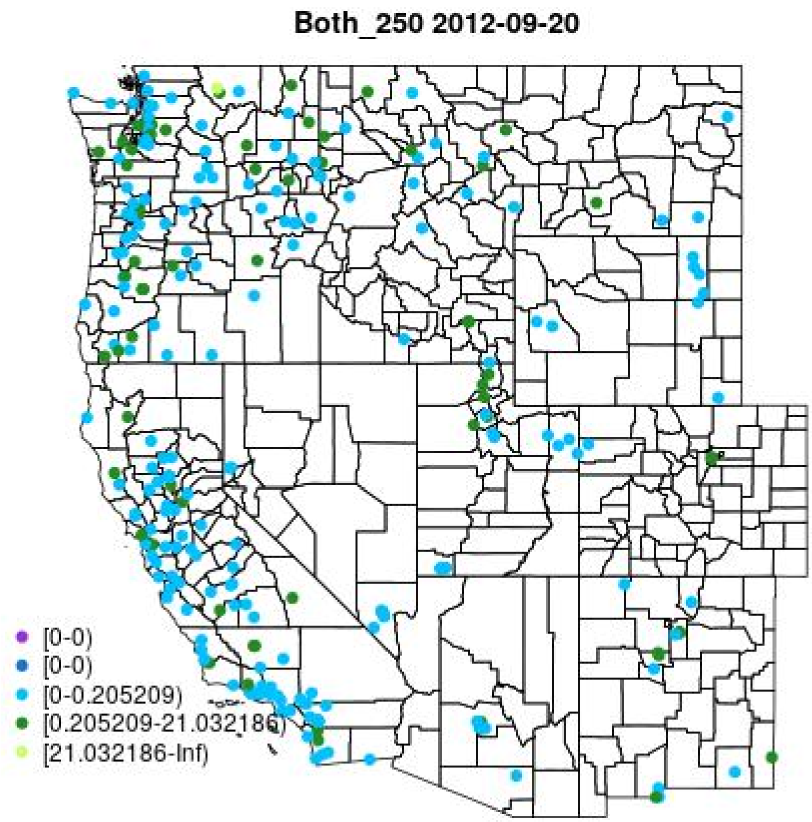

Figure 121: Both-250 2012-09-20

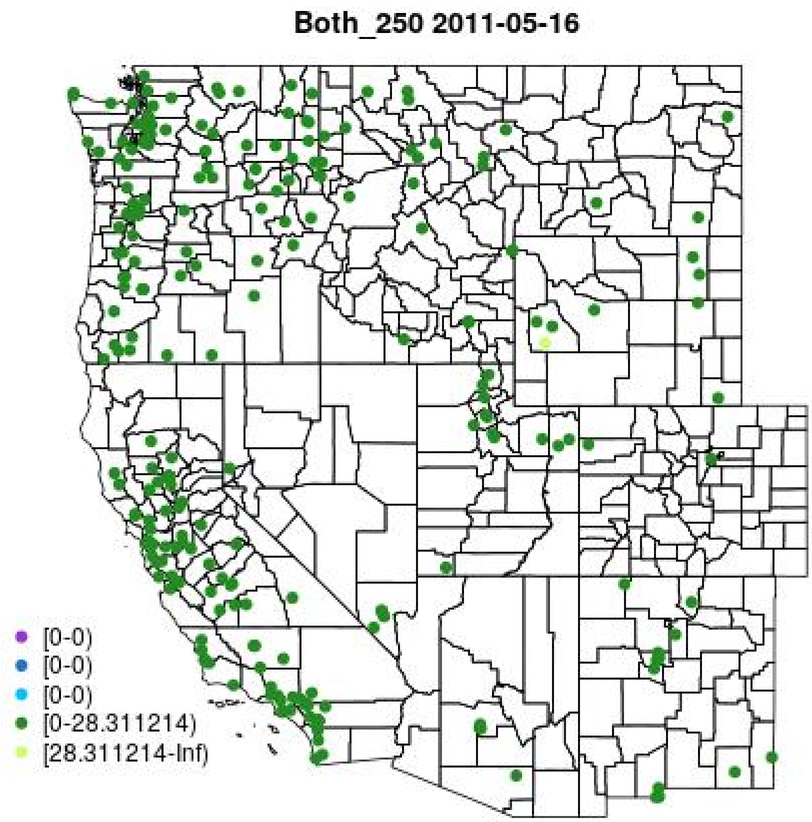

Figure 122: Both-250 2011-05-16

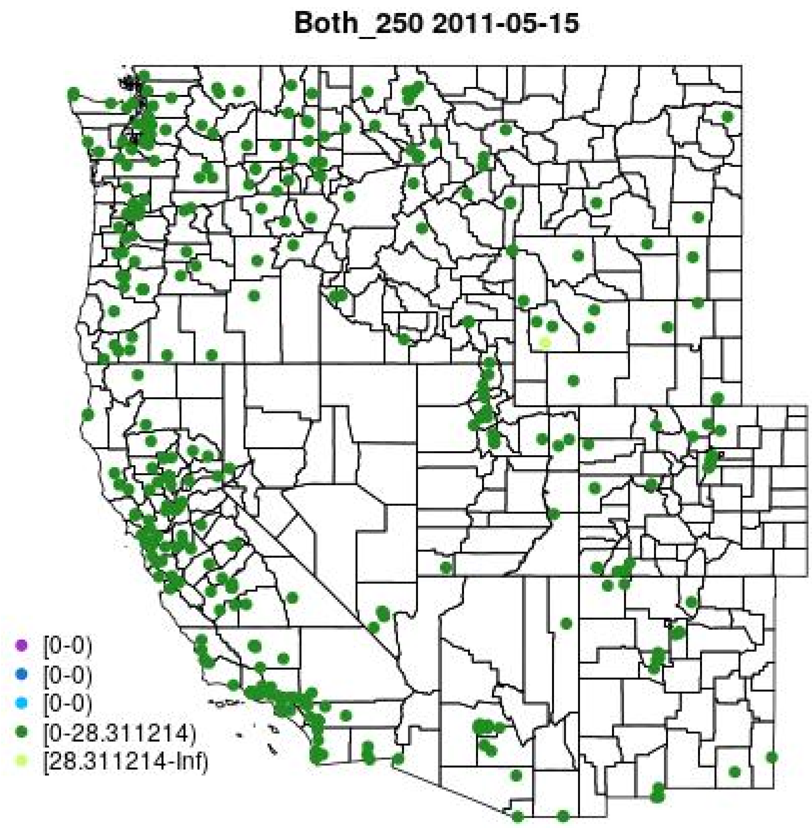

Figure 123: Both-250 2011-05-15

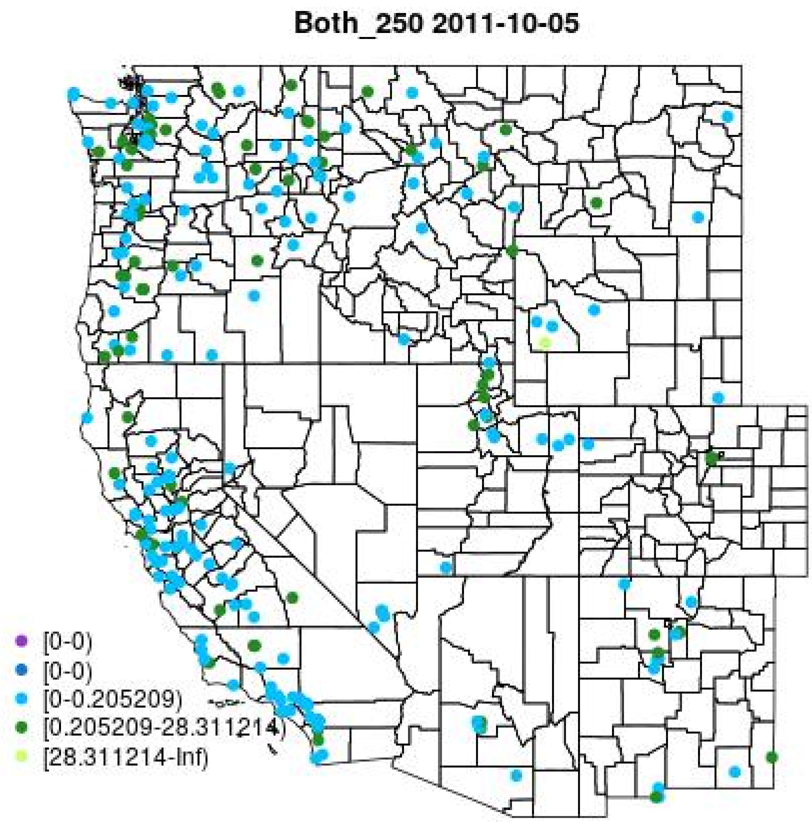

Figure 124: Both-250 2011-10-05

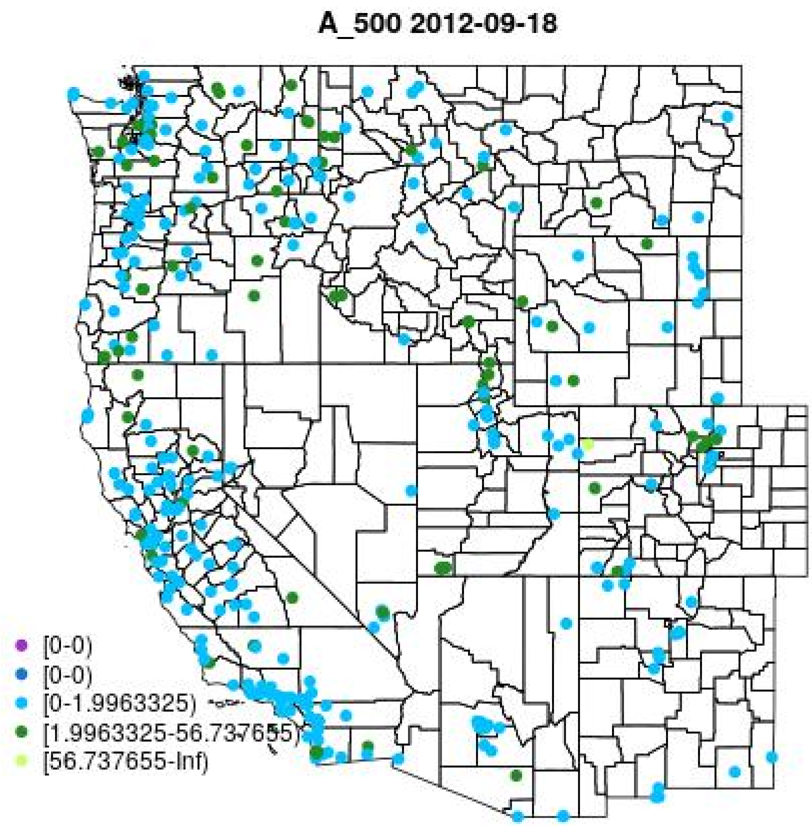

Figure 125: A-500 2012-09-18

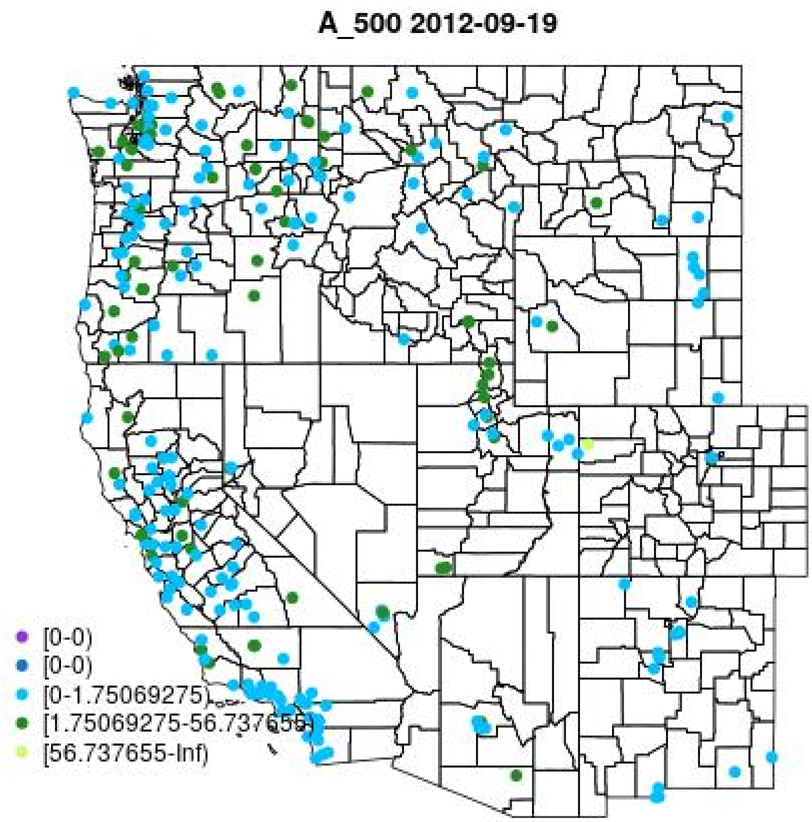

Figure 126: A-500 2012-09-19

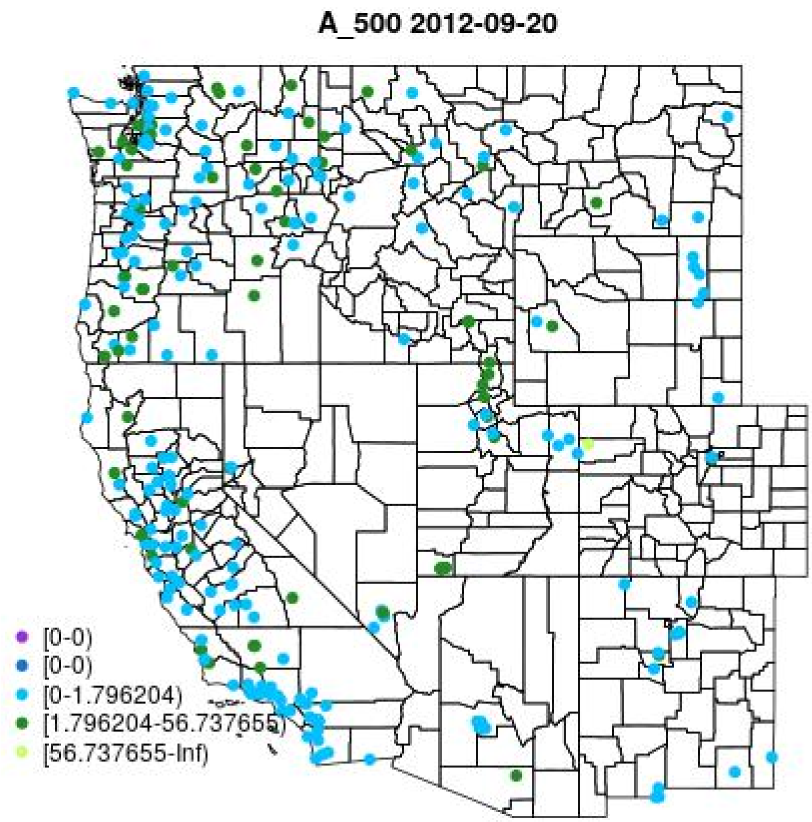

Figure 127: A-500 2012-09-20

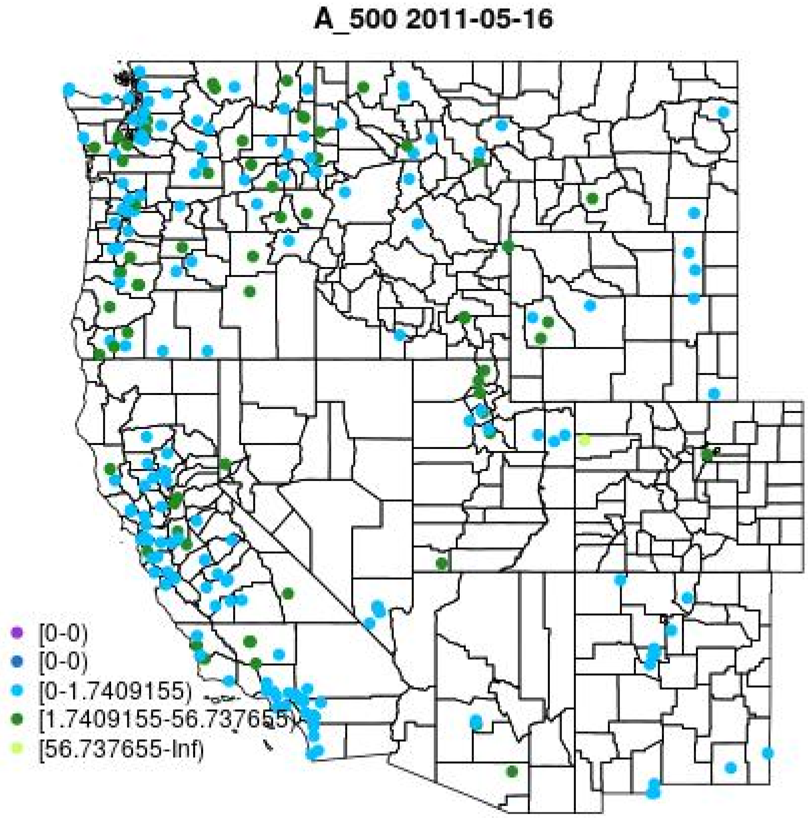

Figure 128: A-500 2011-05-16

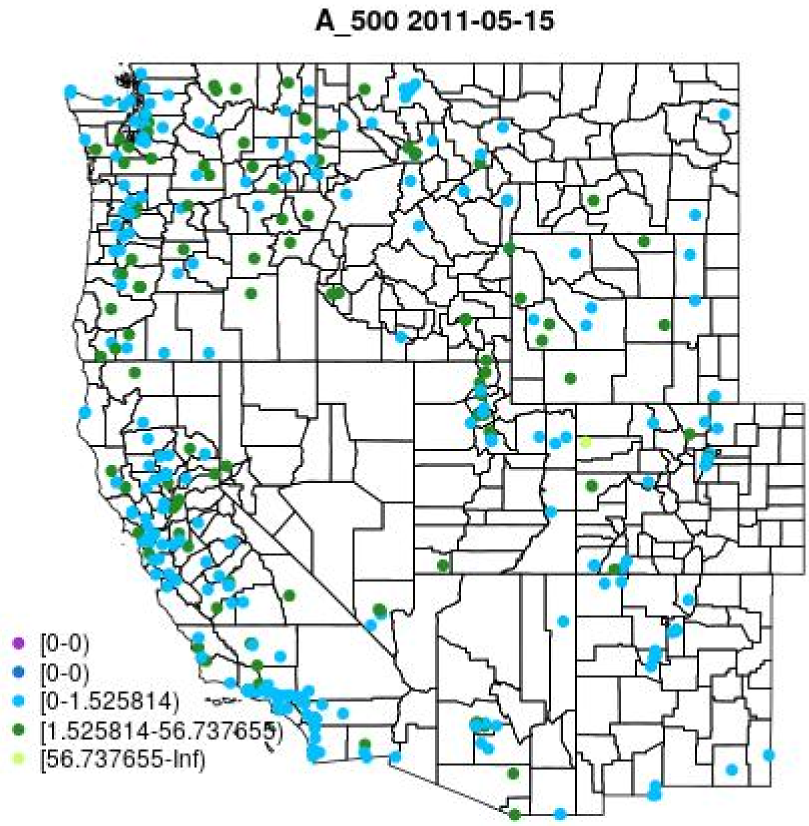

Figure 129: A-500 2011-05-15

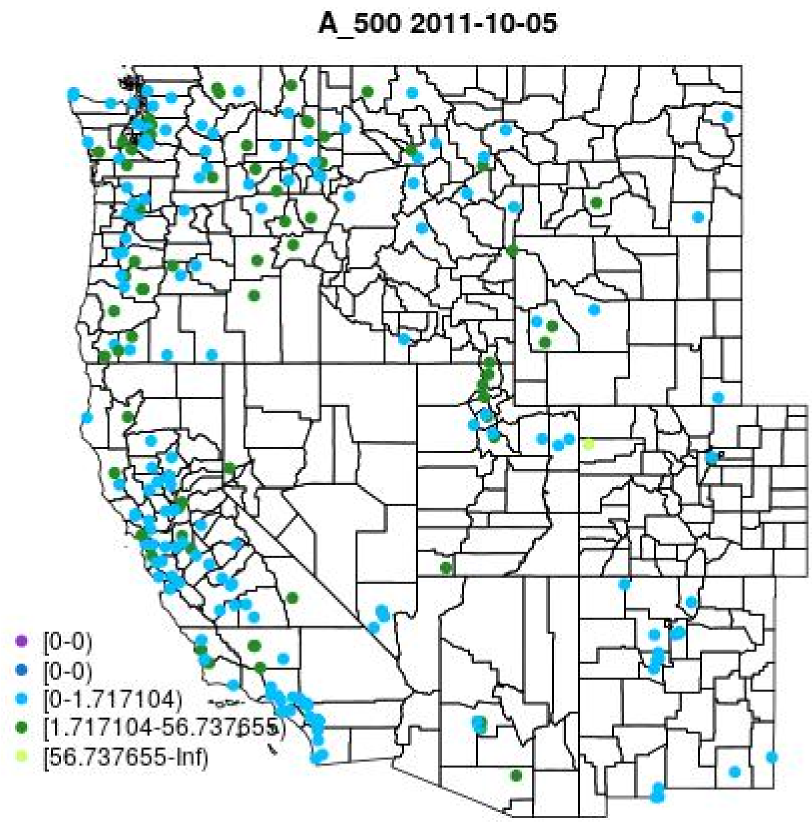

Figure 130: A-500 2011-10-05

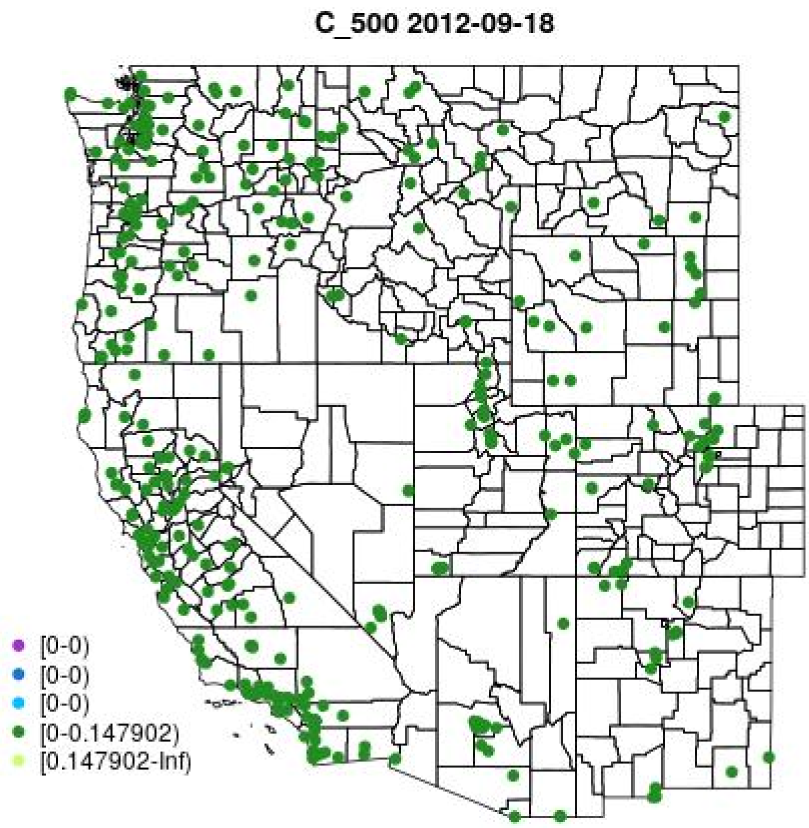

Figure 131: C-500 2012-09-18

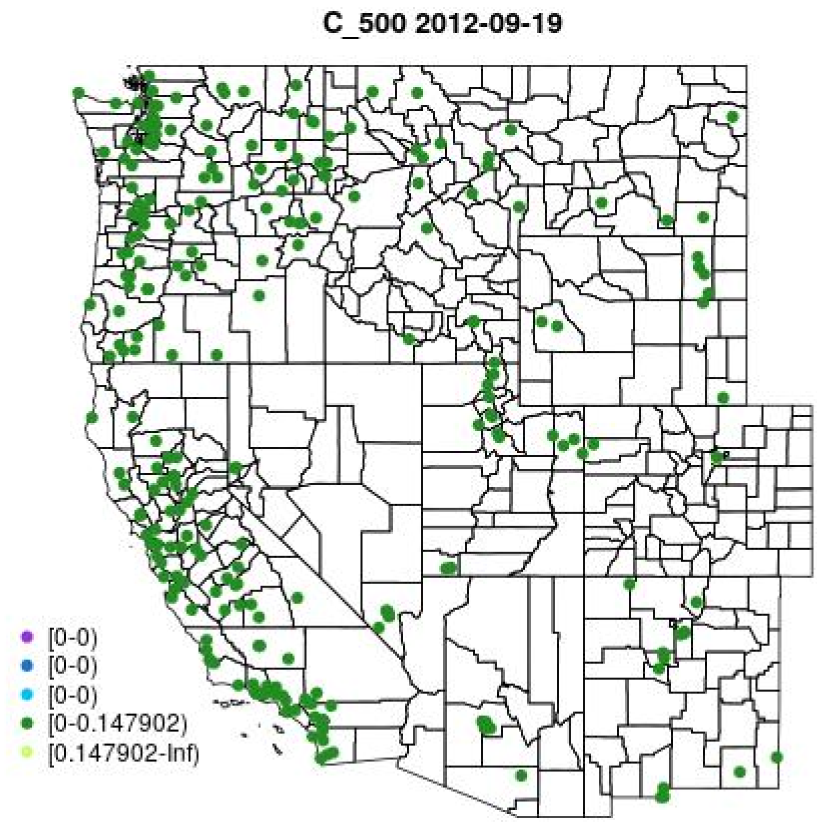

Figure 132: C-500 2012-09-19

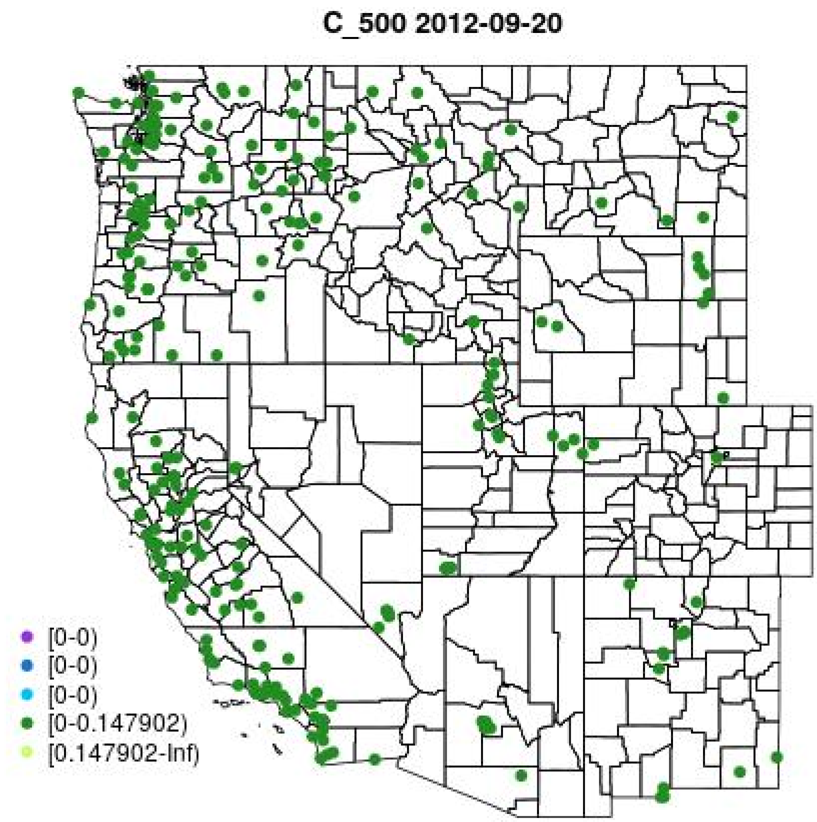

Figure 133: C-500 2012-09-20

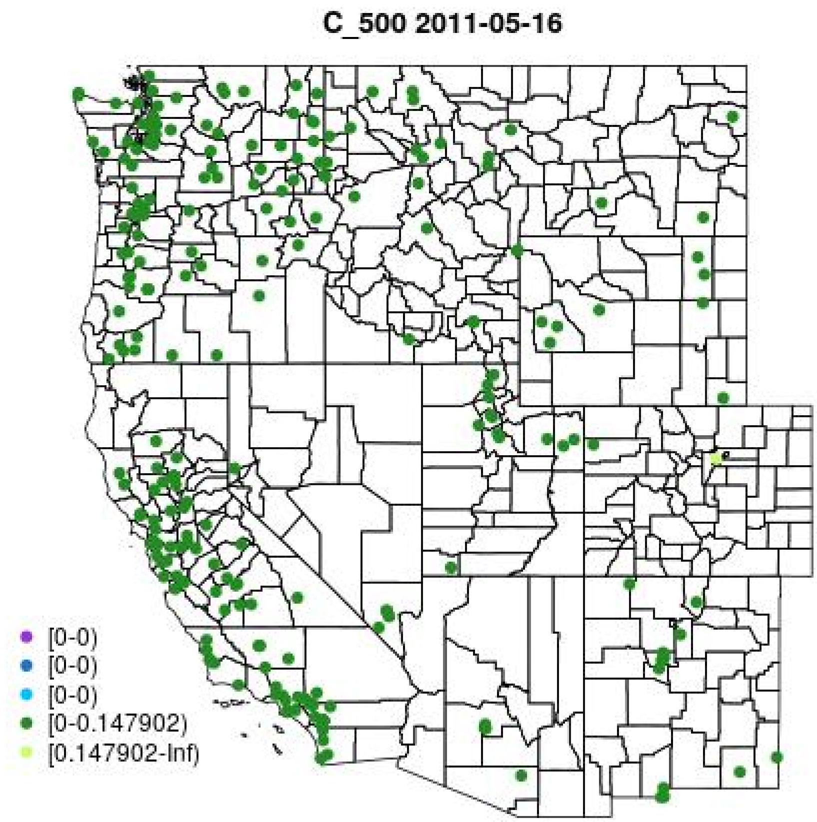

Figure 134: C-500 2011-05-16

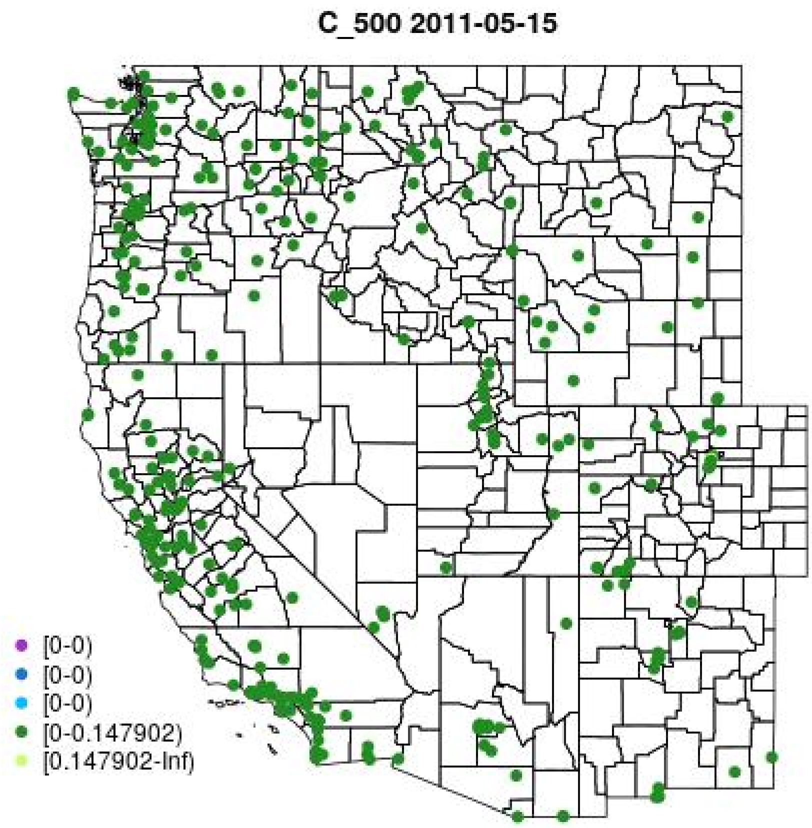

Figure 135: C-500 2011-05-15

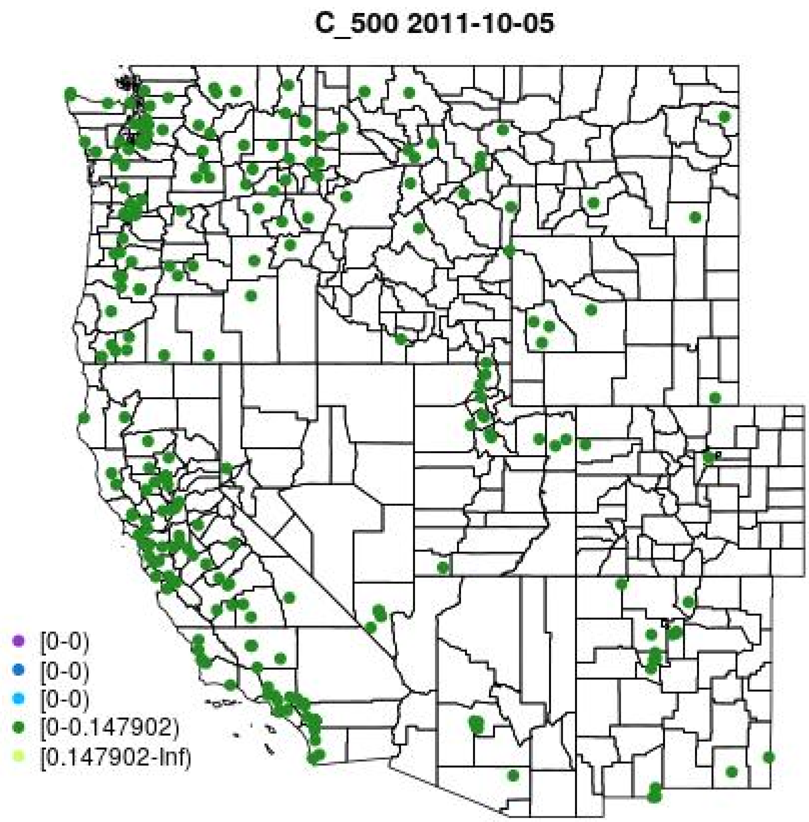

Figure 136: C-500 2011-10-05

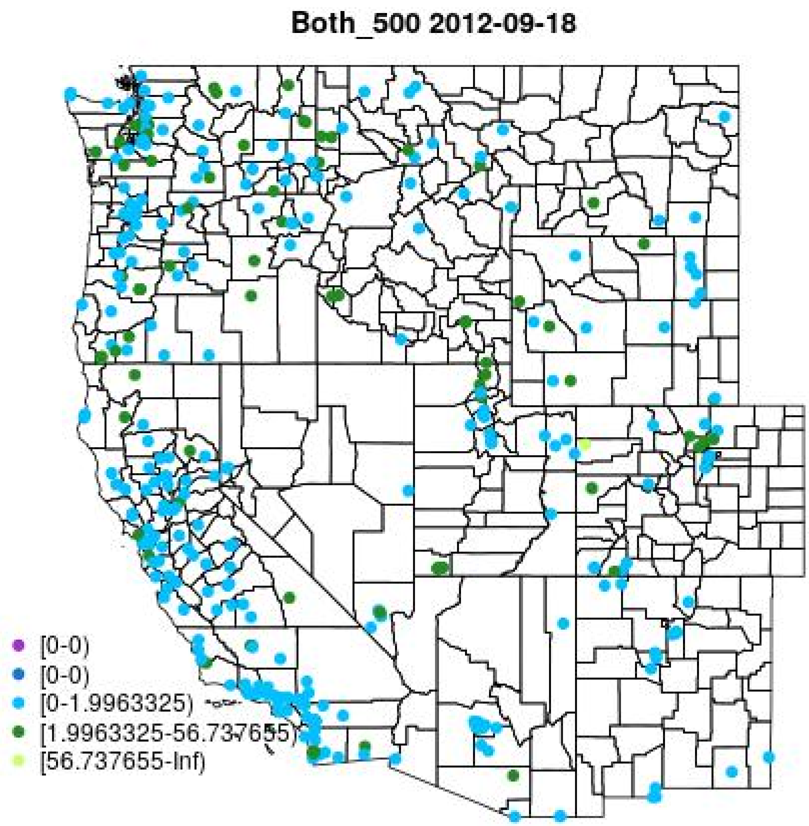

Figure 137: Both-500 2012-09-18

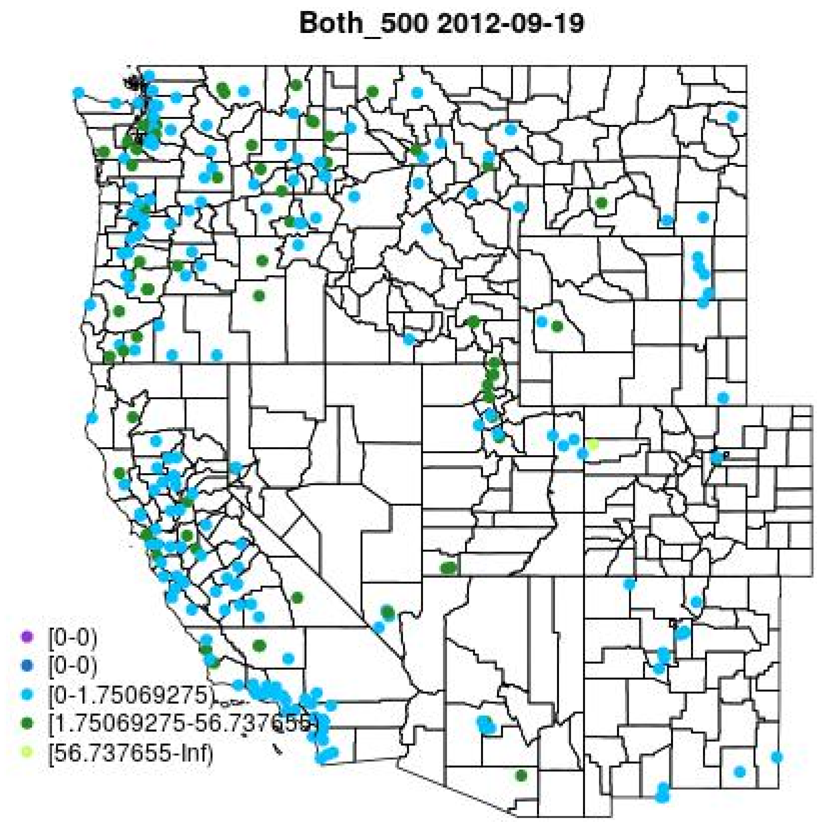

Figure 138: Both-500 2012-09-19

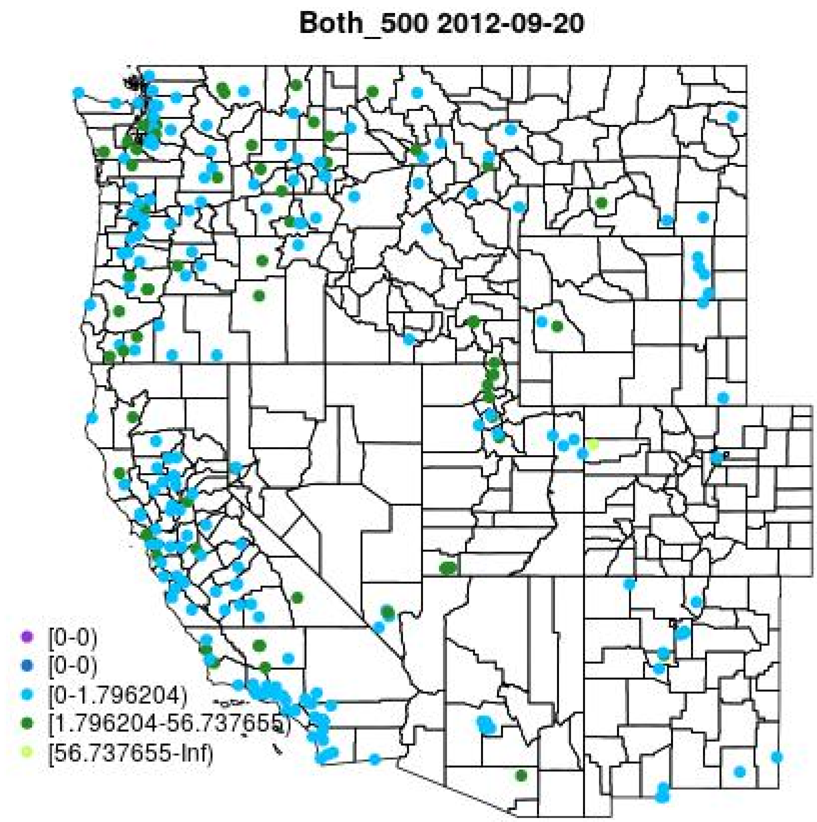

Figure 139: Both-500 2012-09-20

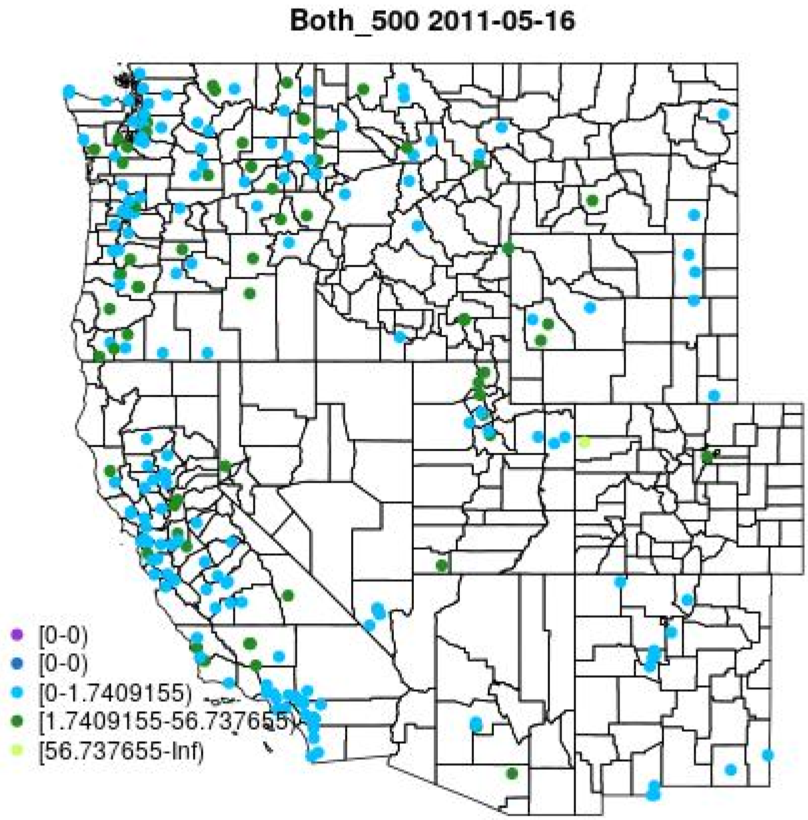

Figure 140: Both-500 2011-05-16

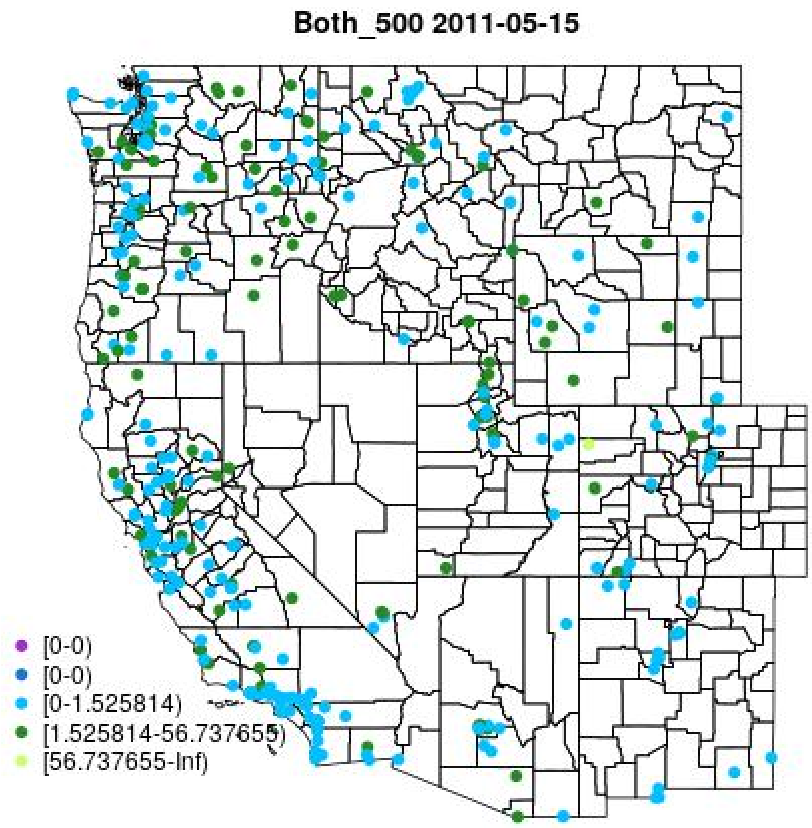

Figure 141: Both-500 2011-05-15

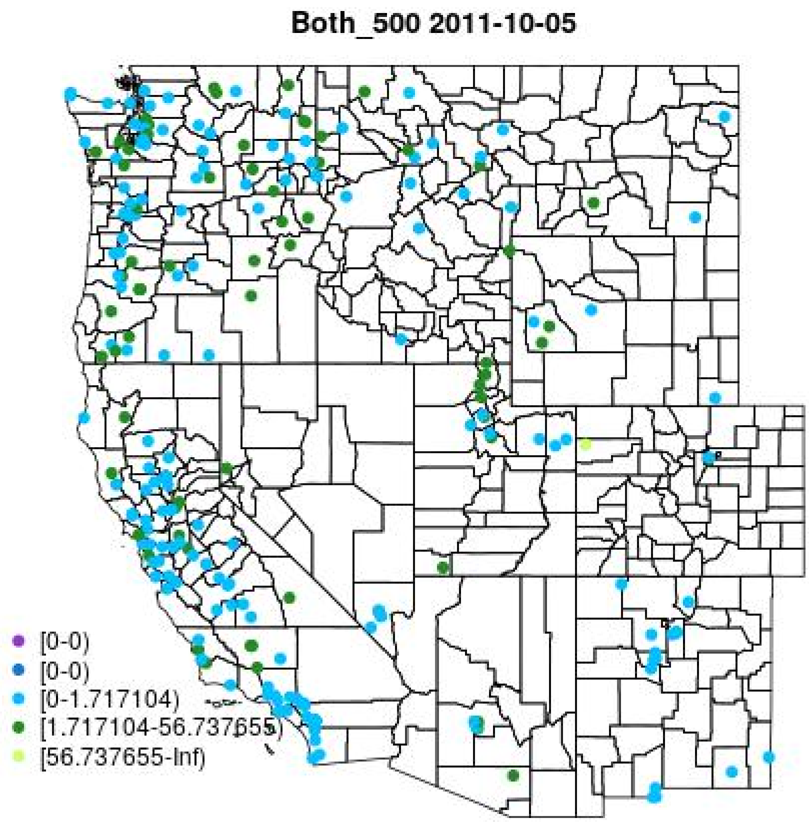

Figure 142: Both-500 2011-10-05

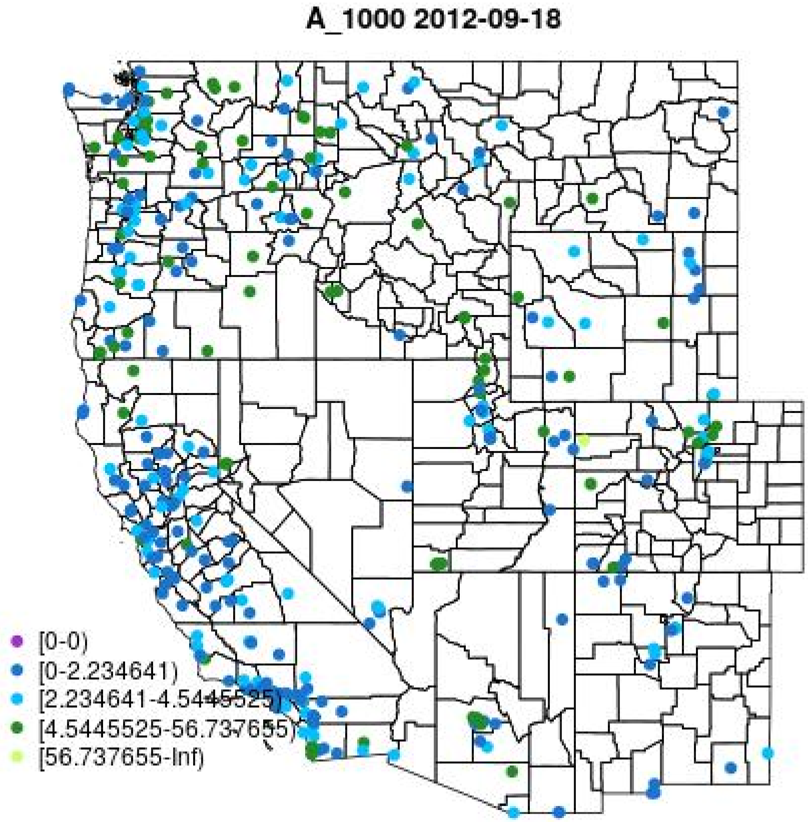

Figure 143: A-1000 2012-09-18

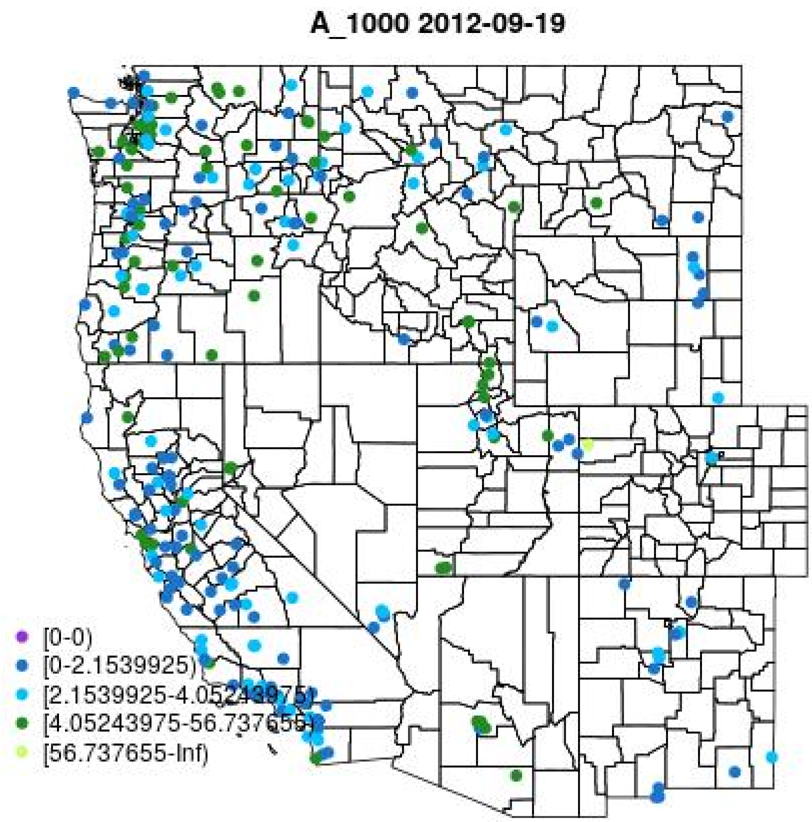

Figure 144: A-1000 2012-09-19

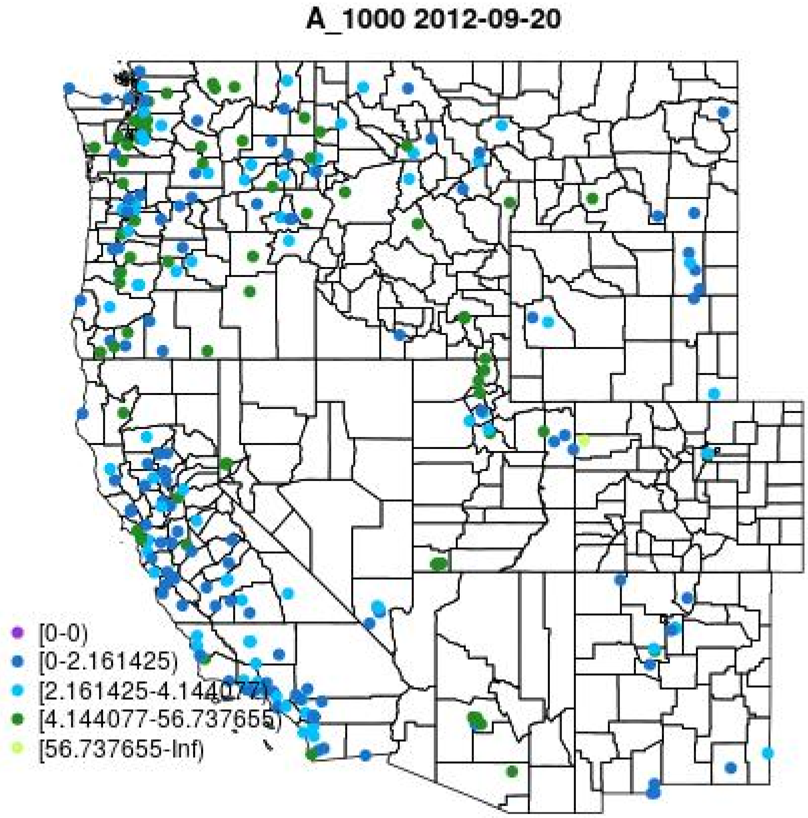

Figure 145: A-1000 2012-09-20

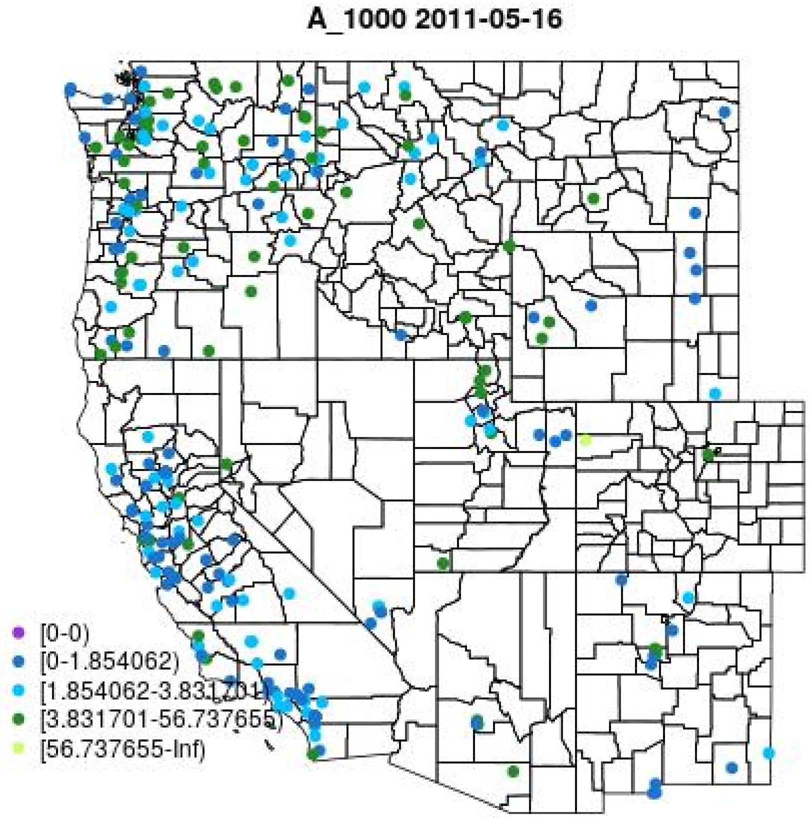

Figure 146: A-1000 2011-05-16

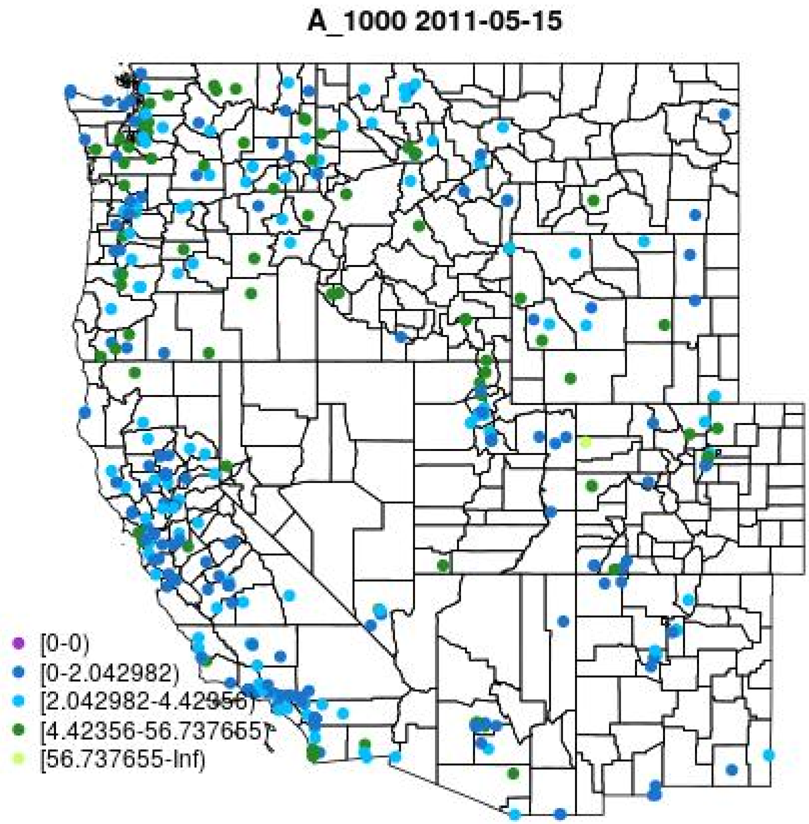

Figure 147: A-1000 2011-05-15

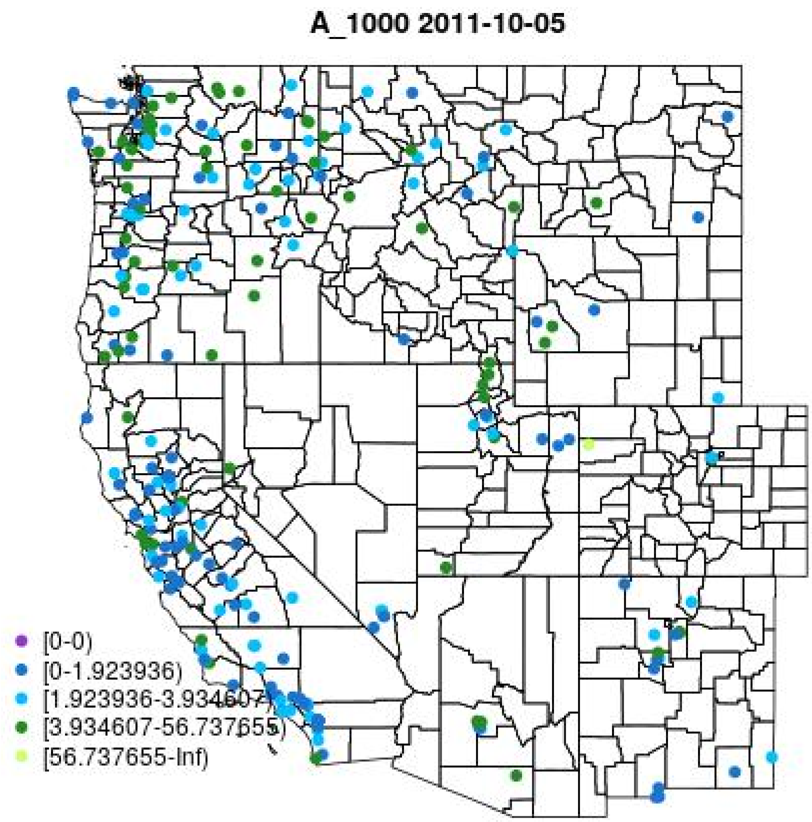

Figure 148: A-1000 2011-10-05

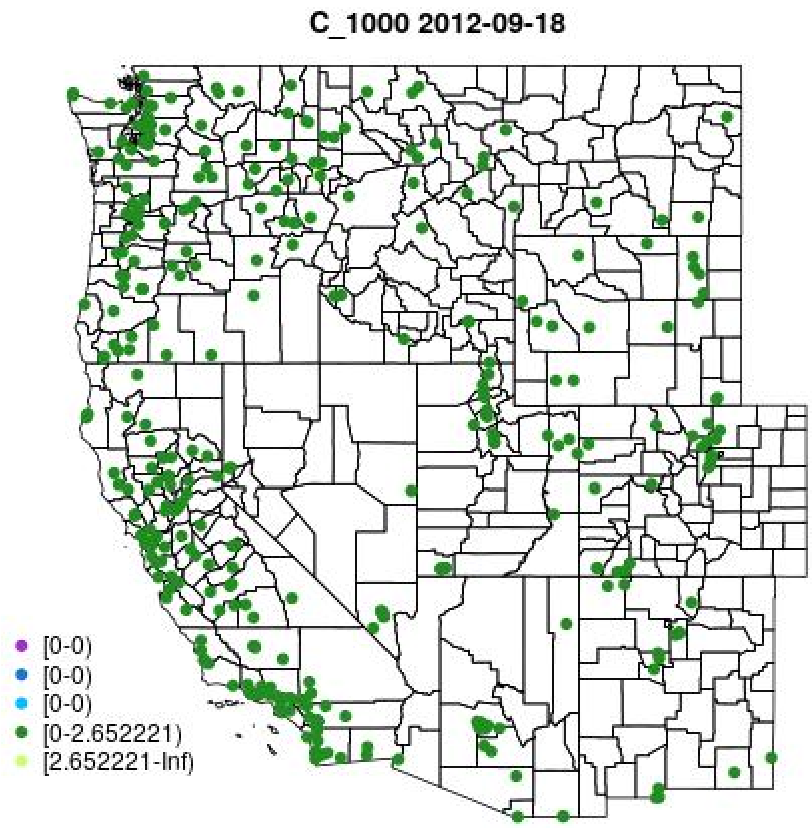

Figure 149: C-1000 2012-09-18

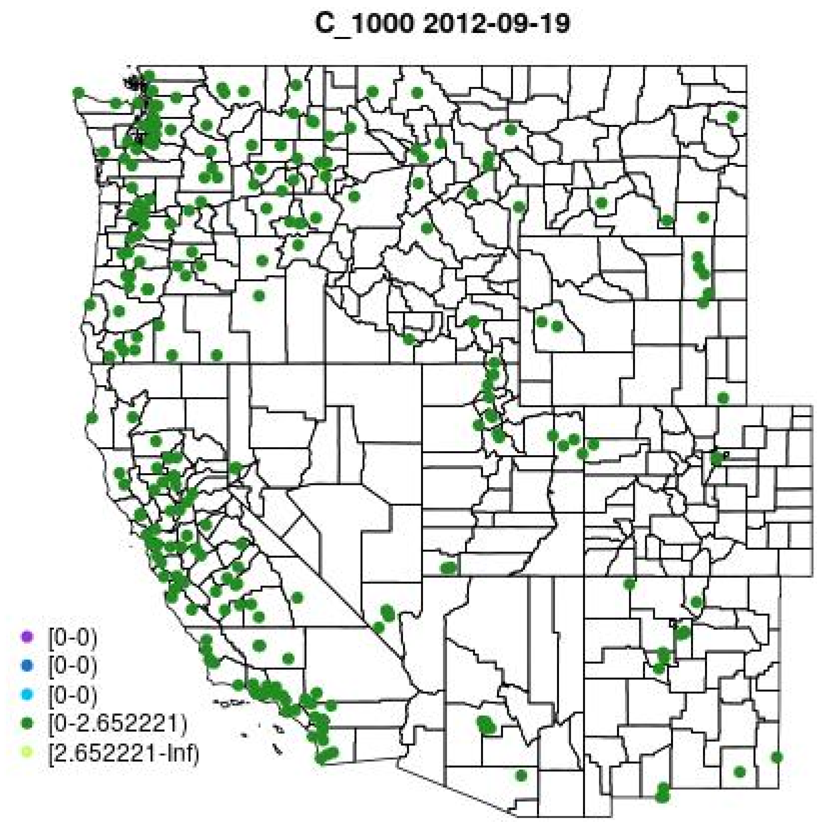

Figure 150: C-1000 2012-09-19

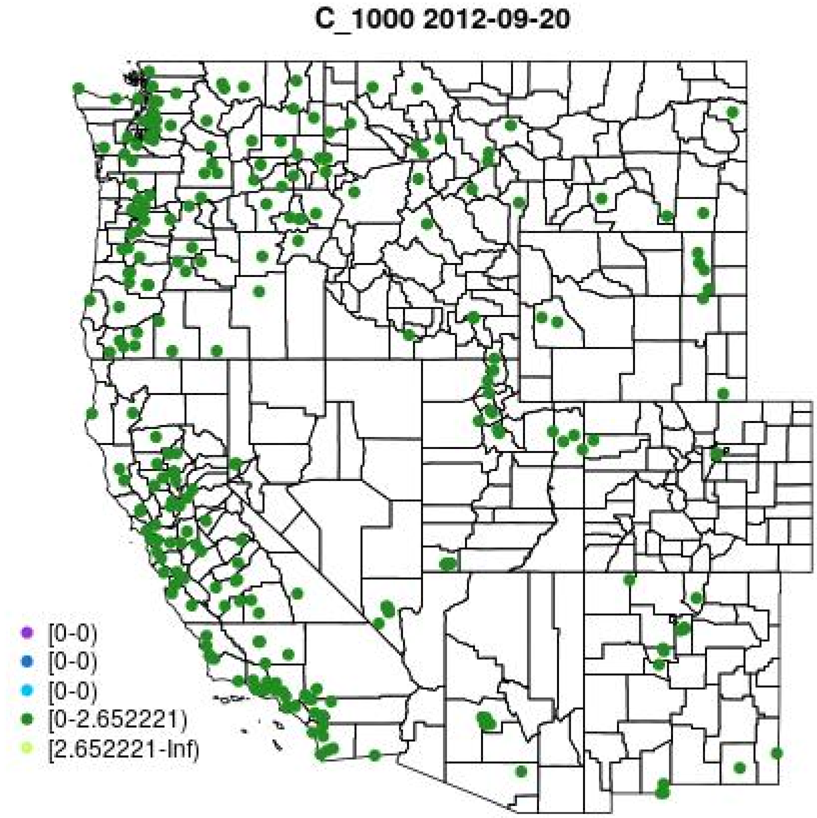

Figure 151: C-1000 2012-09-20

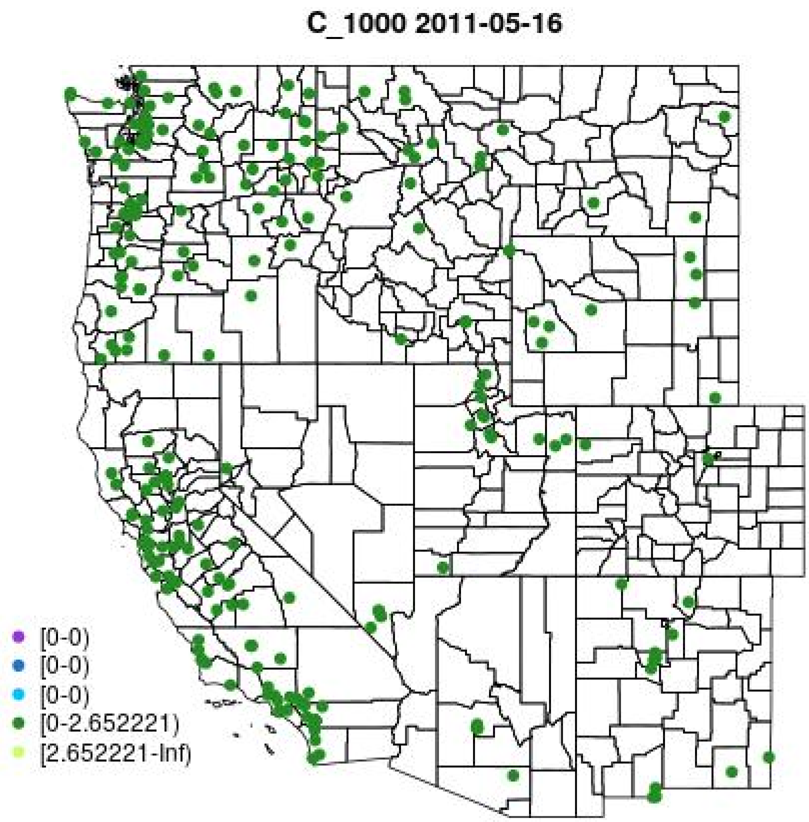

Figure 152: C-1000 2011-05-16

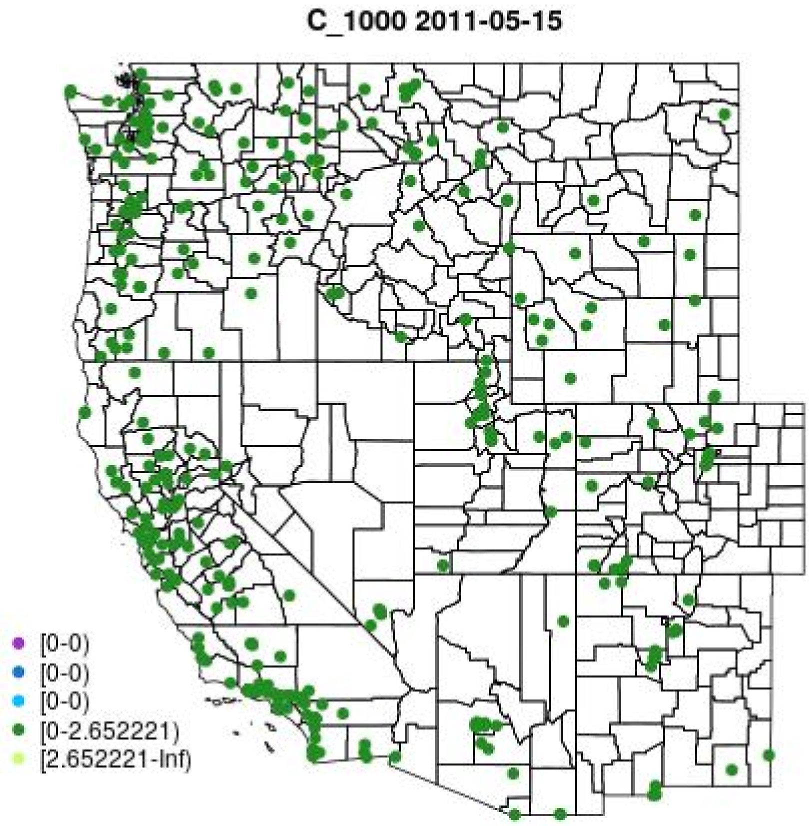

Figure 153: C-1000 2011-05-15

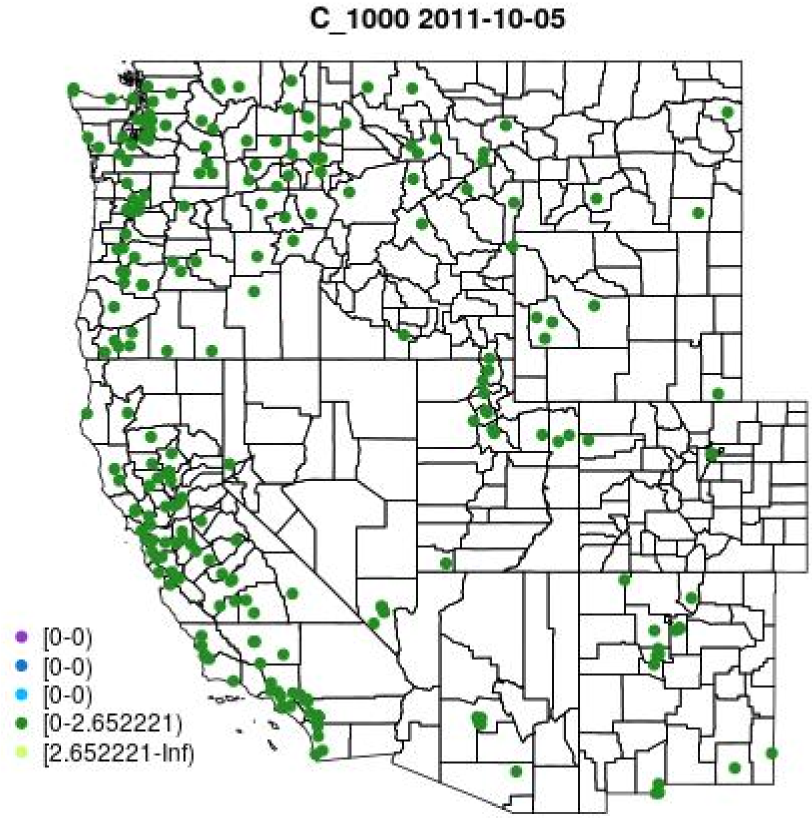

Figure 154: C-1000 2011-10-05

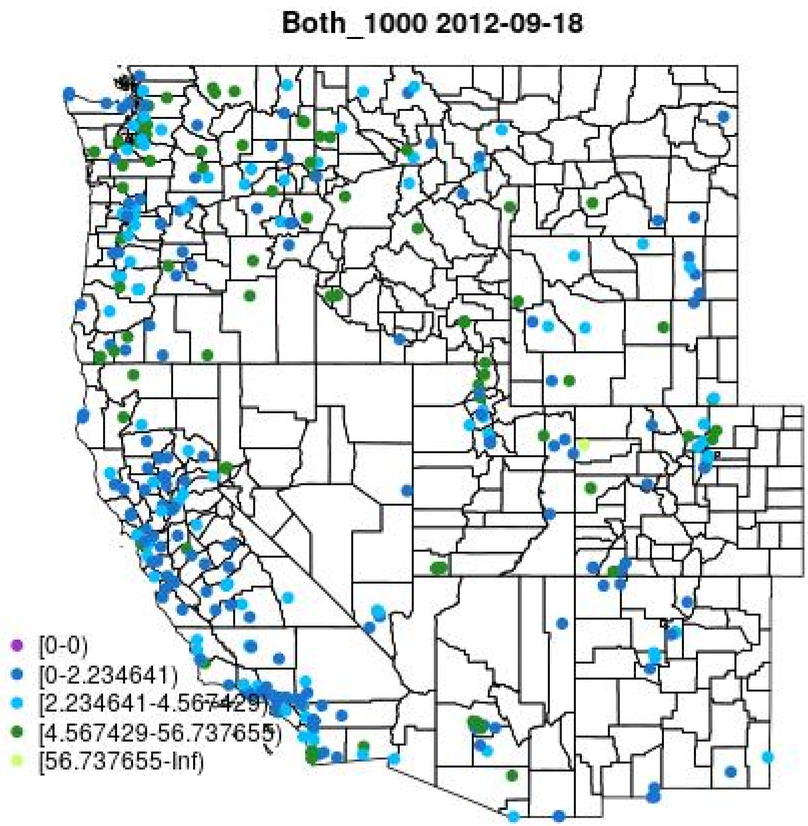

Figure 155: Both-1000 2012-09-18

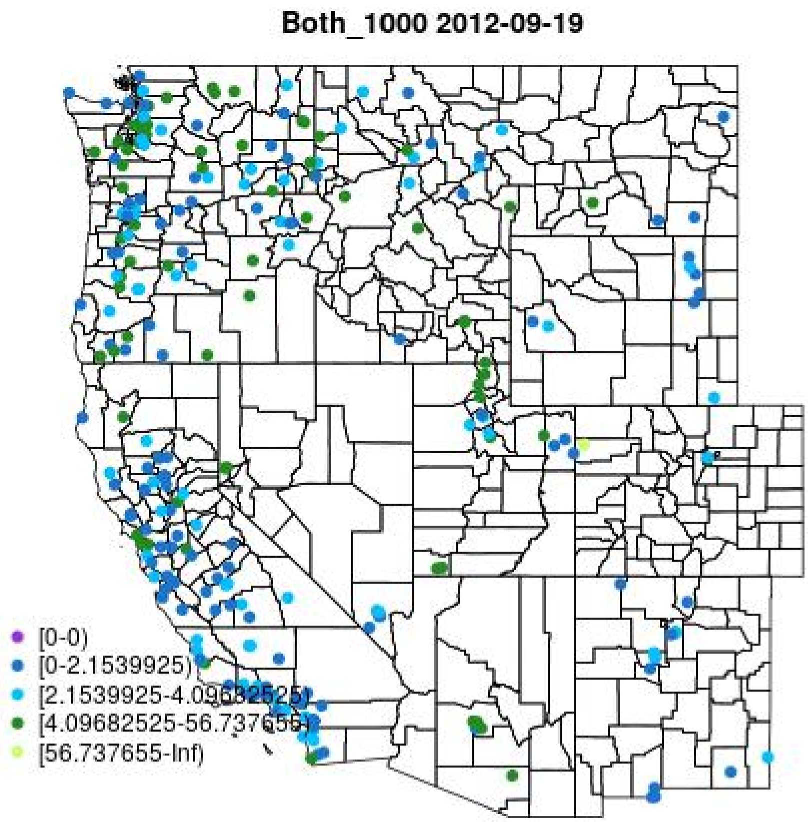

Figure 156: Both-1000 2012-09-19

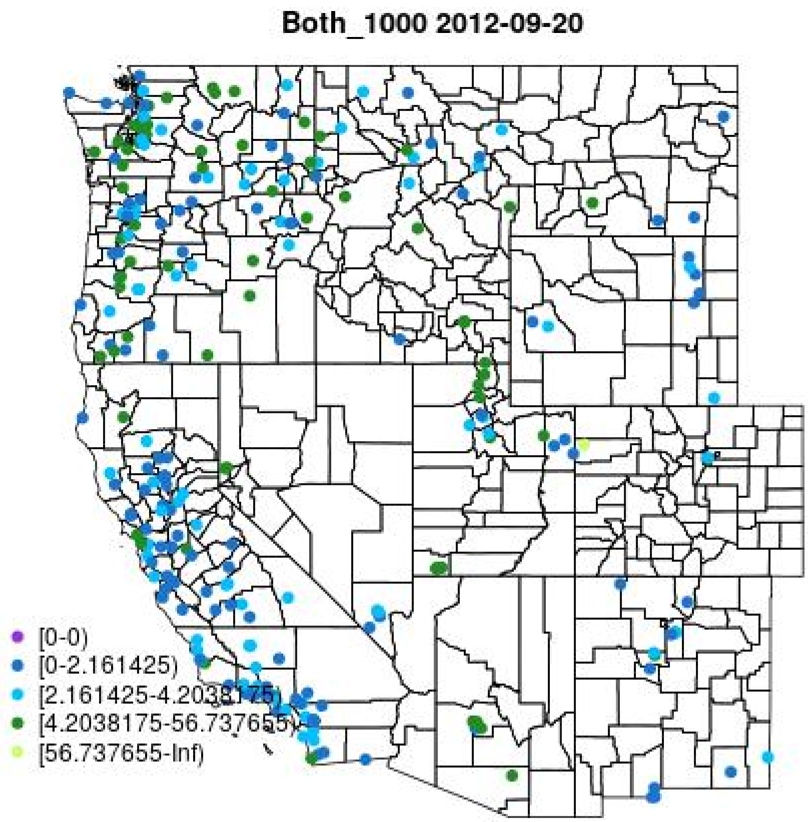

Figure 157: Both-1000 2012-09-20

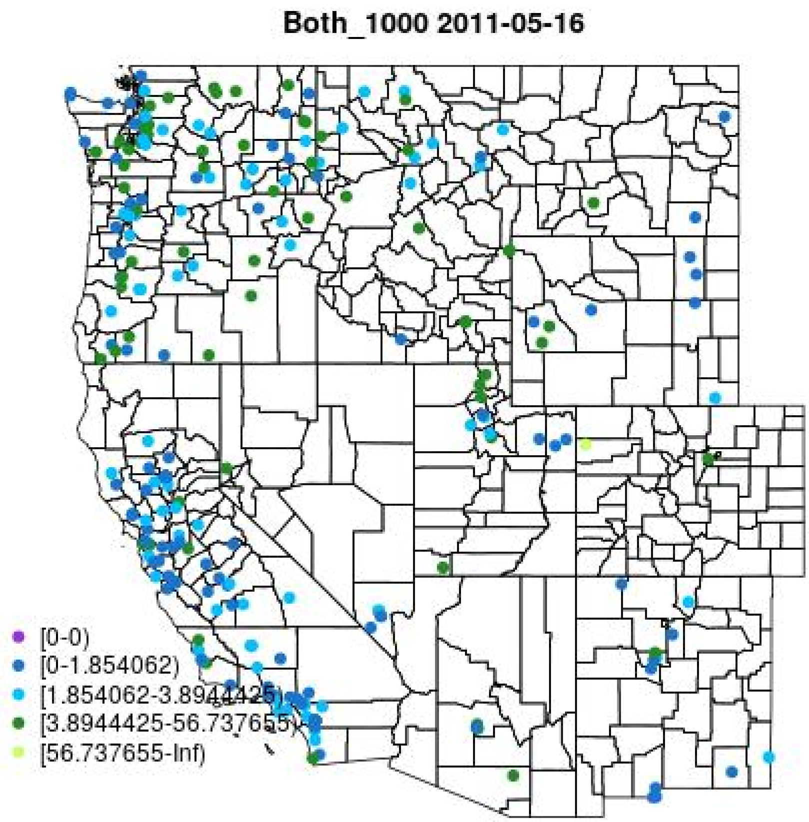

Figure 158: Both-1000 2011-05-16

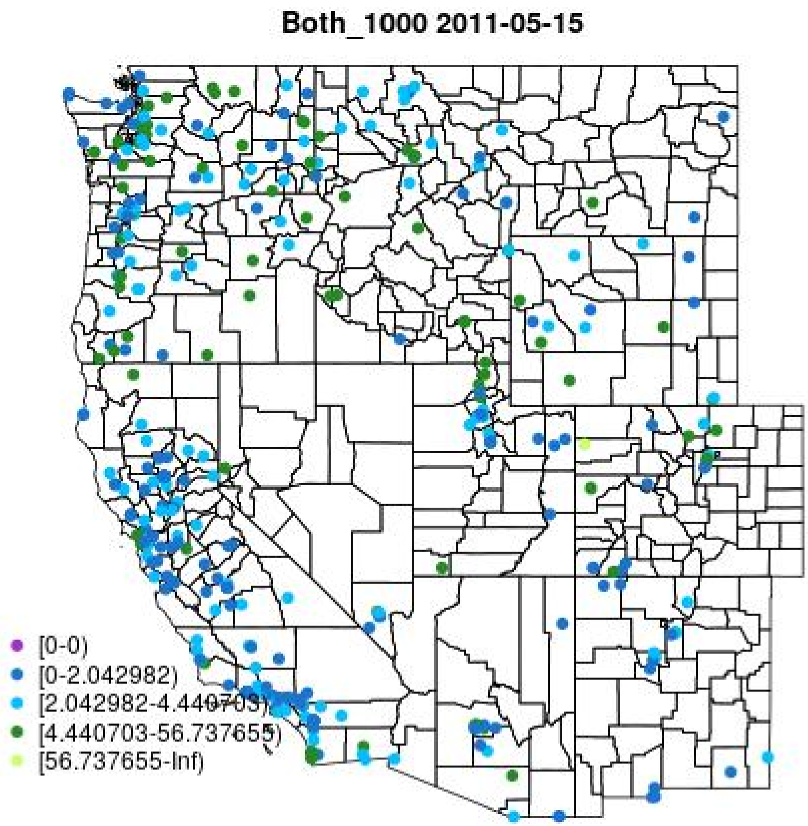

Figure 159: Both-1000 2011-05-15

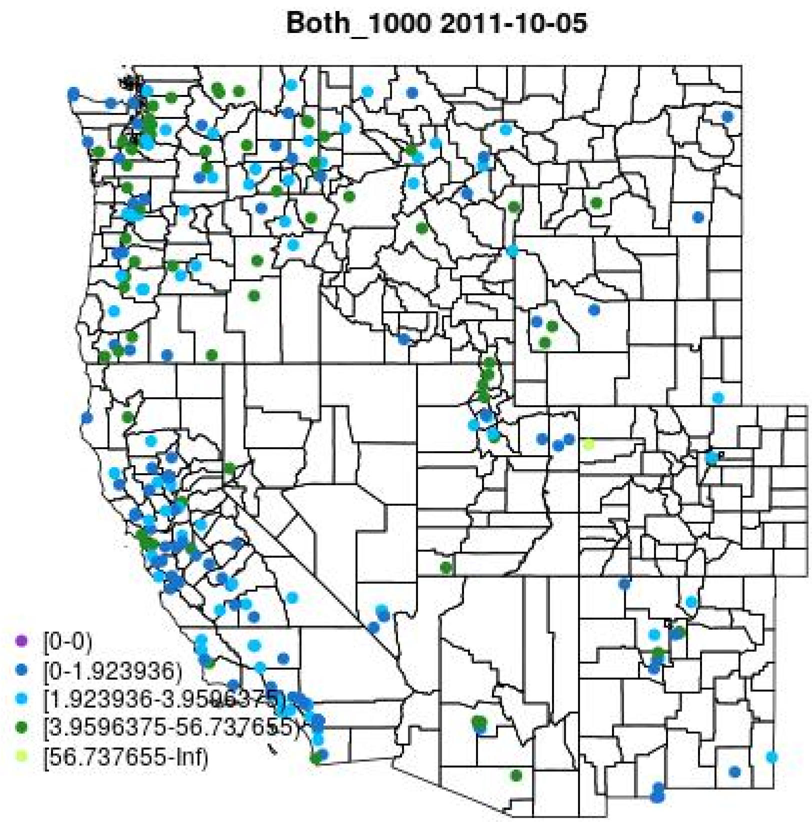

Figure 160: Both-1000 2011-10-05

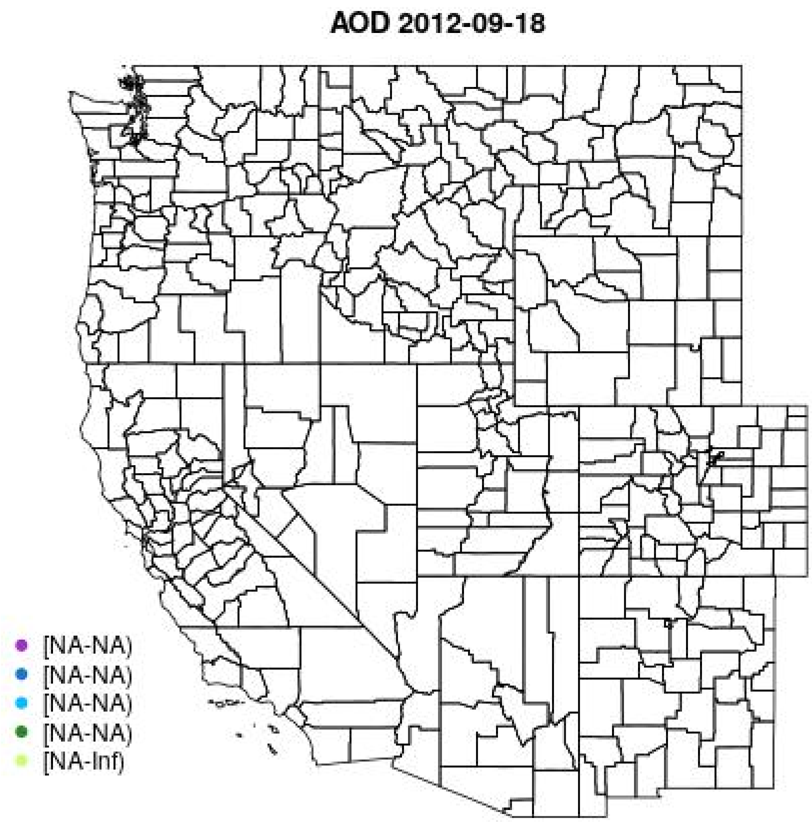

Figure 161: AOD 2012-09-18

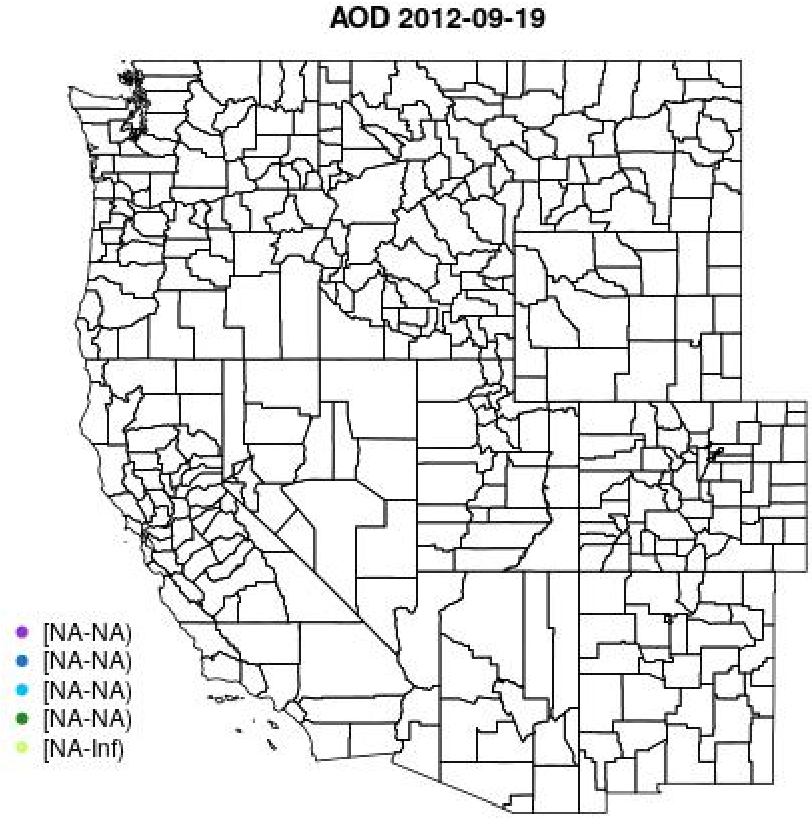

Figure 162: AOD 2012-09-19

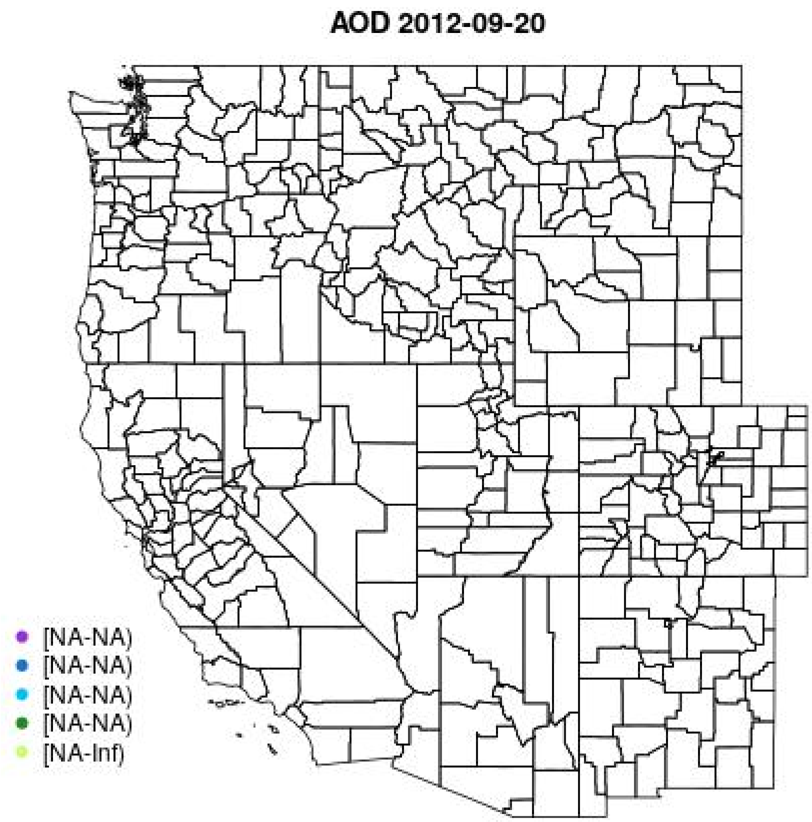

Figure 163: AOD 2012-09-20

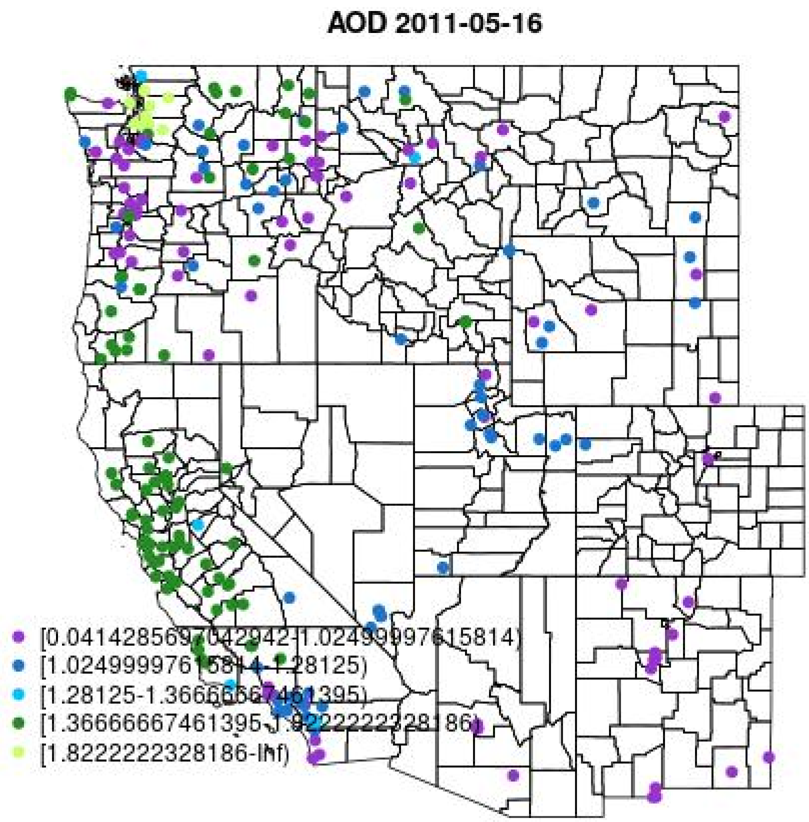

Figure 164: AOD 2011-05-16

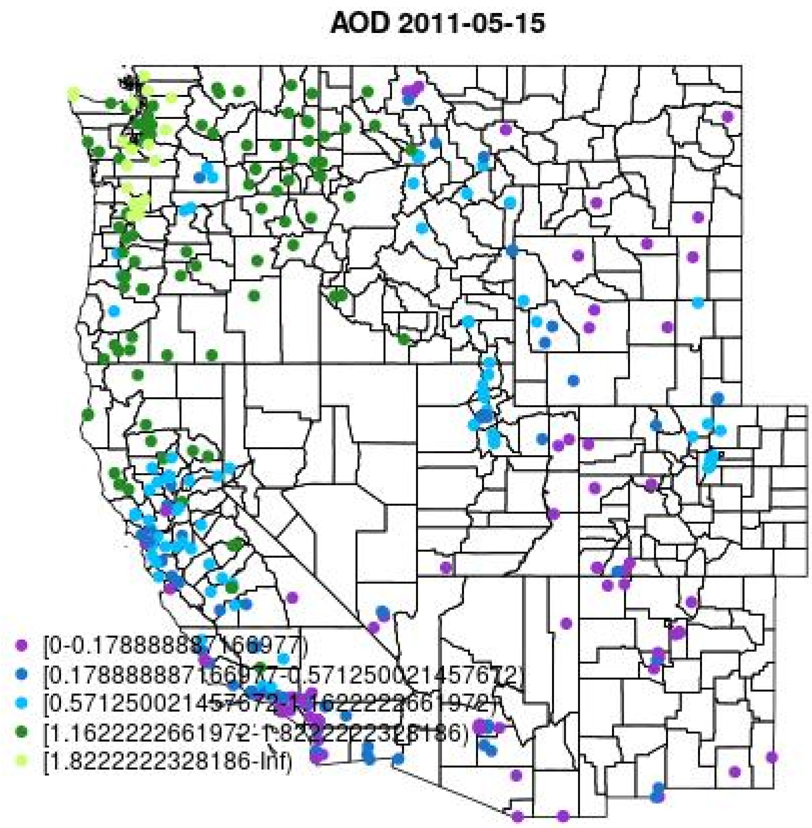

Figure 165: AOD 2011-05-15

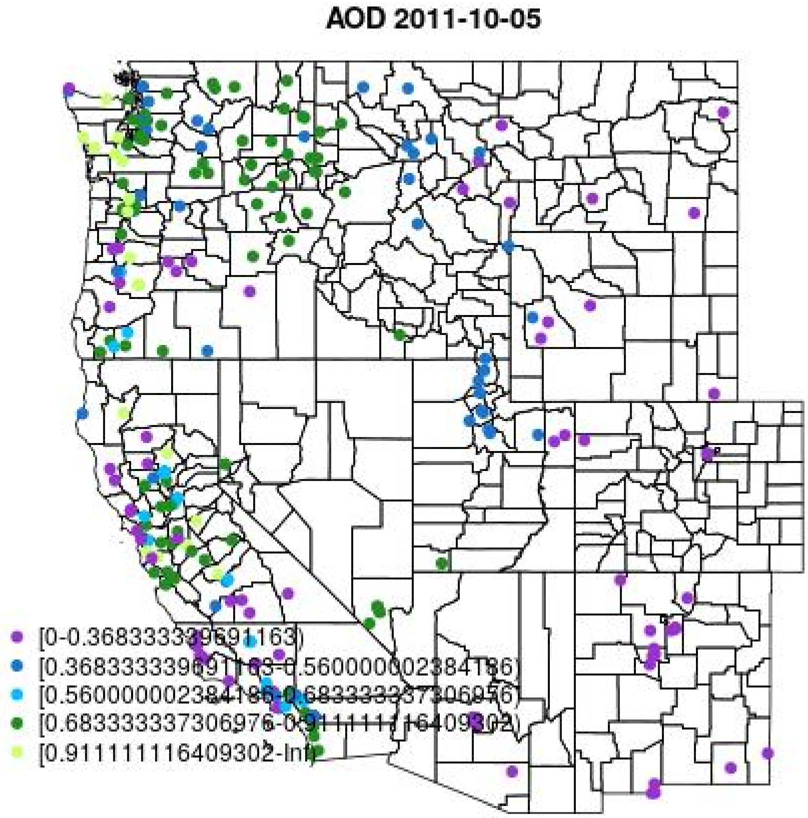

Figure 166: AOD 2011-10-05

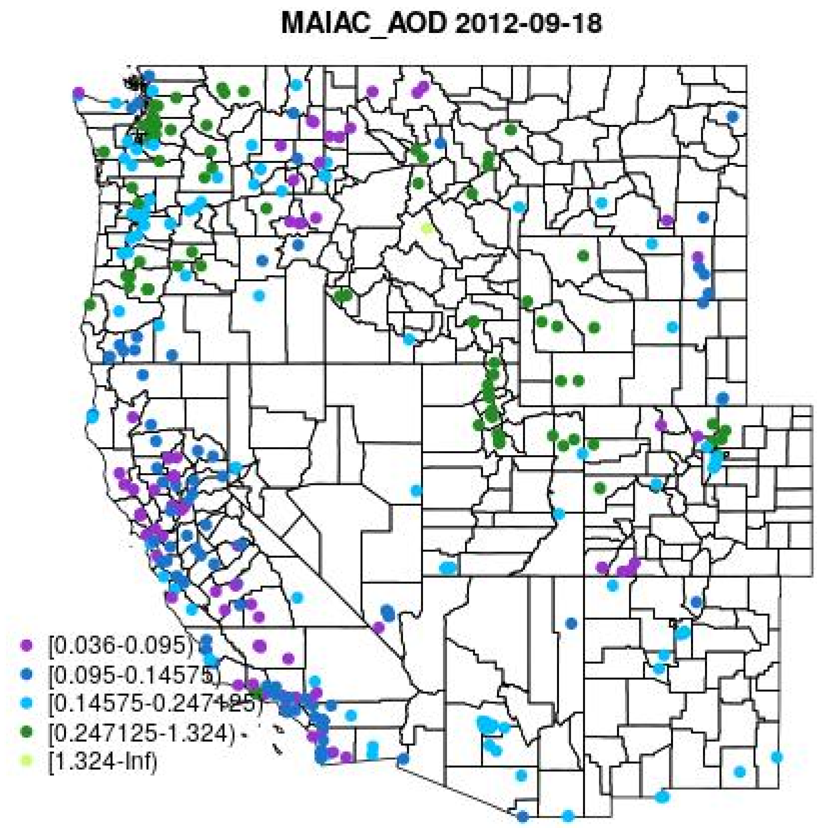

Figure 167: MAIAC-AOD 2012-09-18

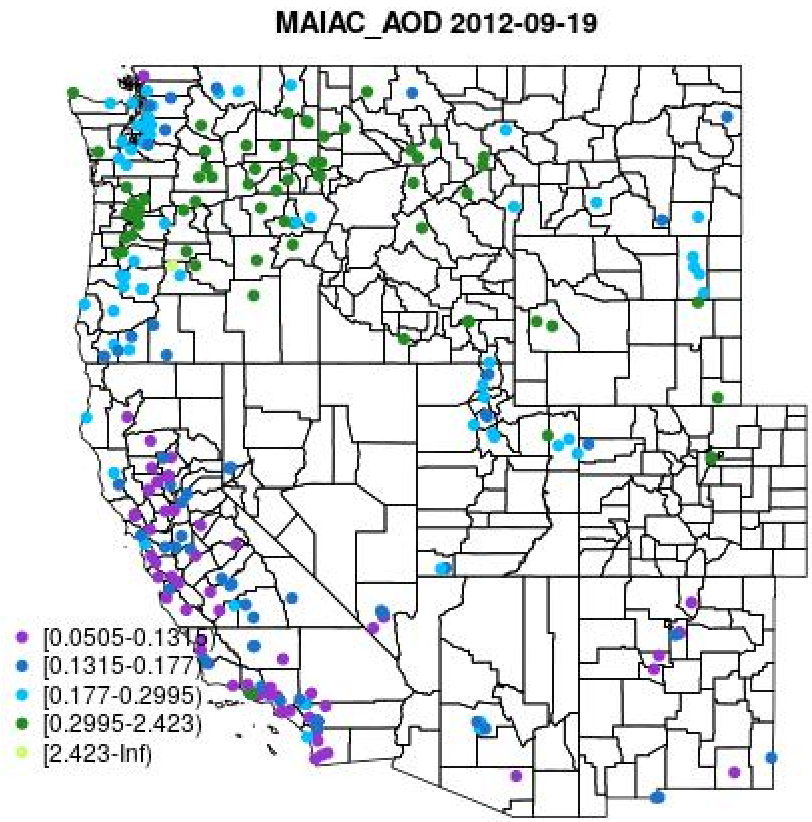

Figure 168: MAIAC-AOD 2012-09-19

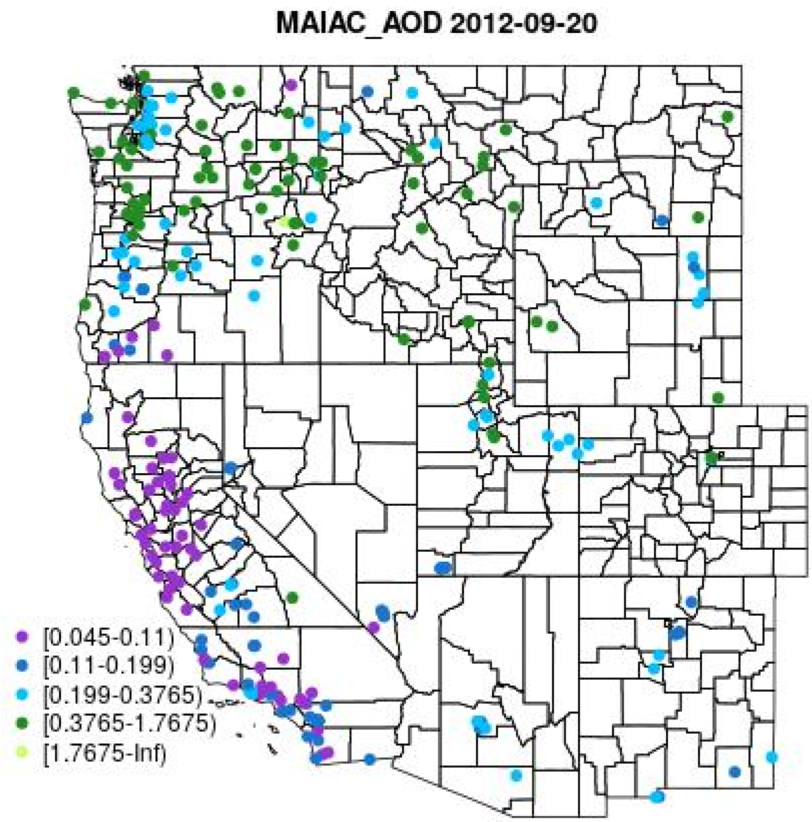

Figure 169: MAIAC-AOD 2012-09-20

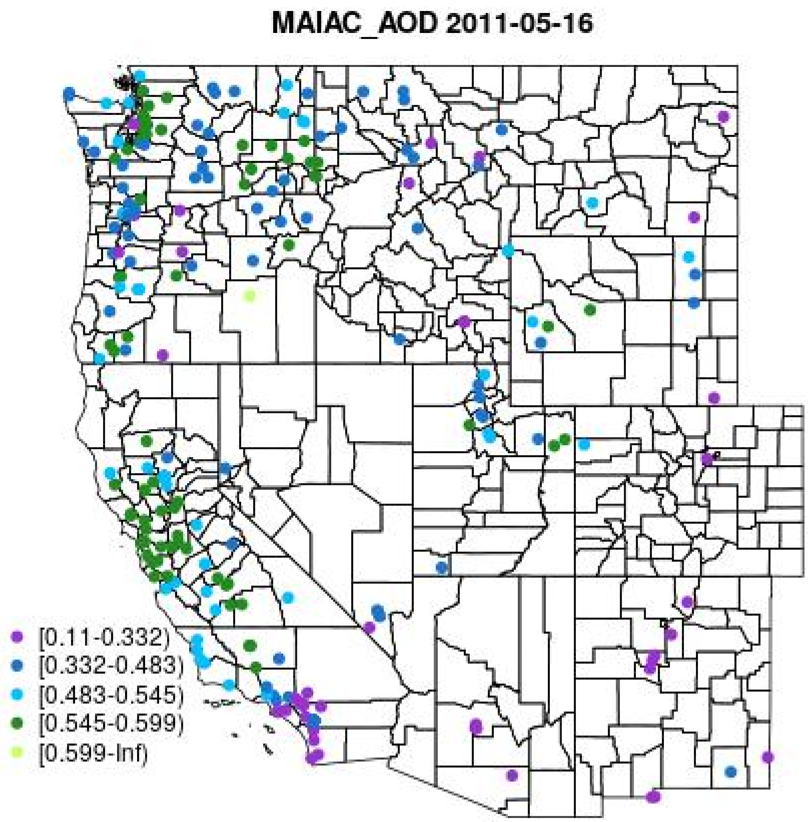

Figure 170: MAIAC-AOD 2011-05-16

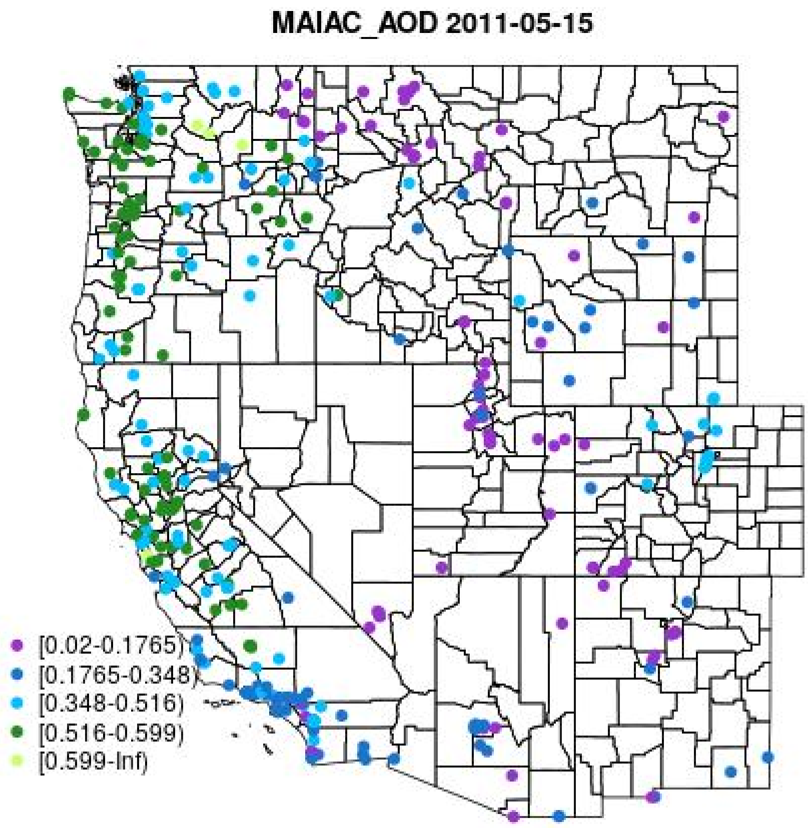

Figure 171: MAIAC-AOD 2011-05-15

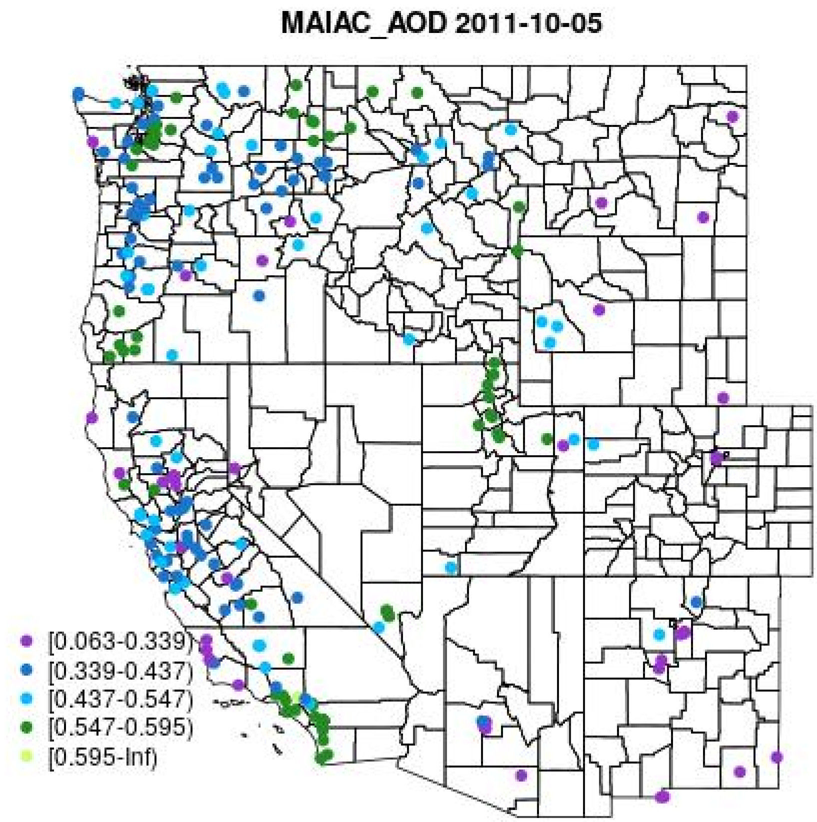

Figure 172: MAIAC-AOD 2011-10-05

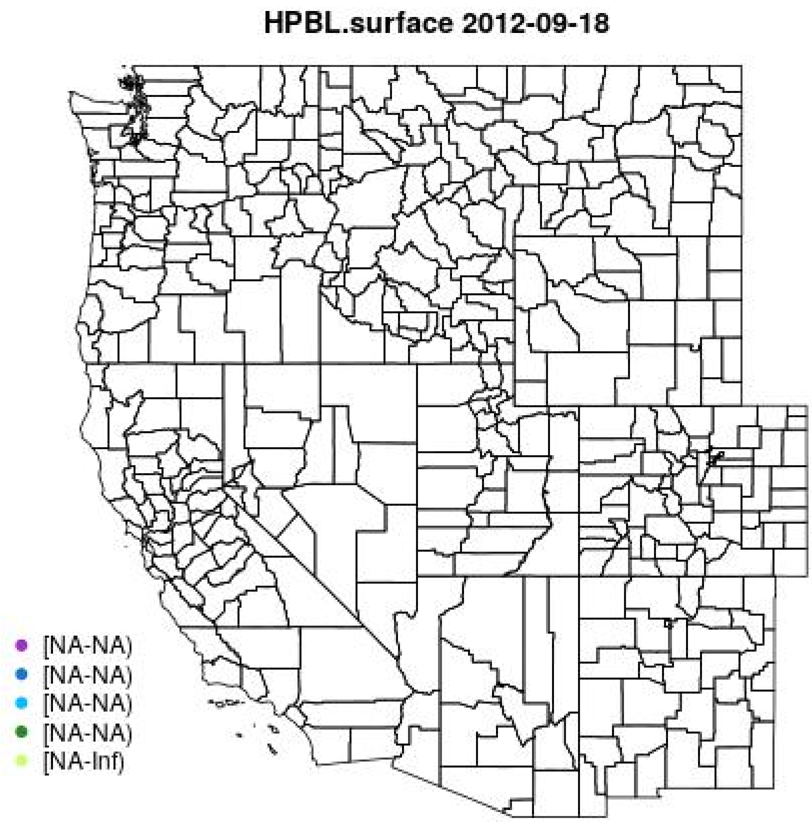

Figure 173: HPBL.surface 2012-09-18

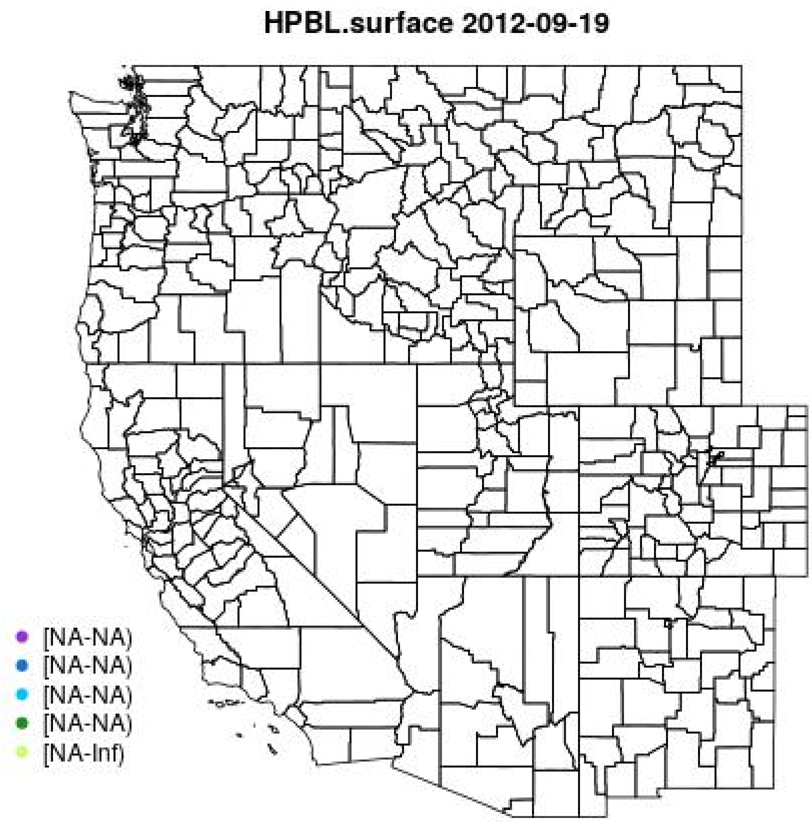

Figure 174: HPBL.surface 2012-09-19

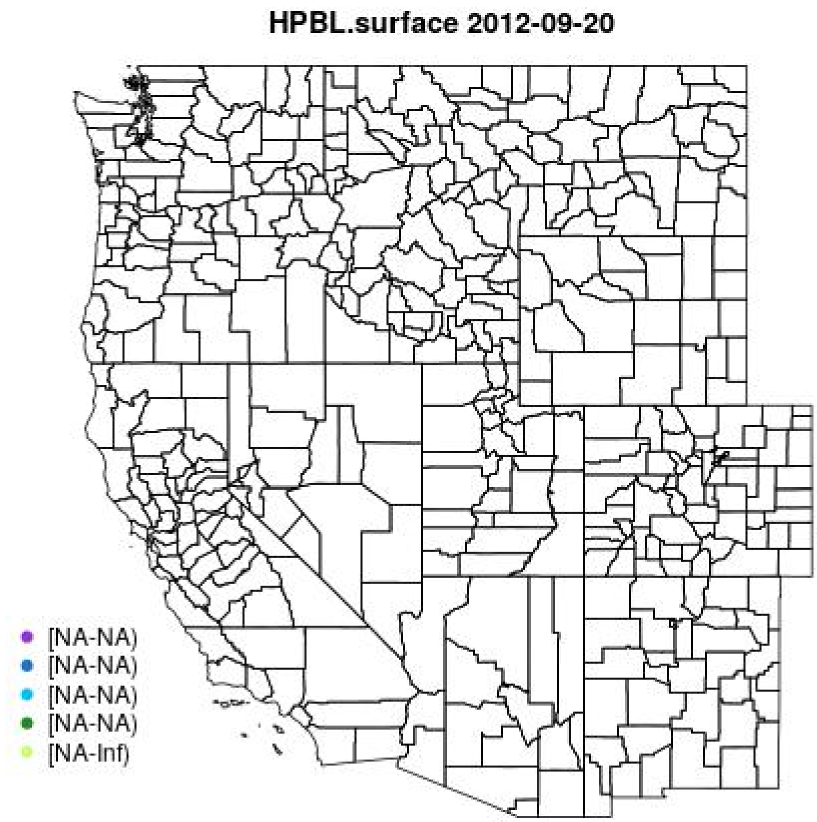

Figure 175: HPBL.surface 2012-09-20

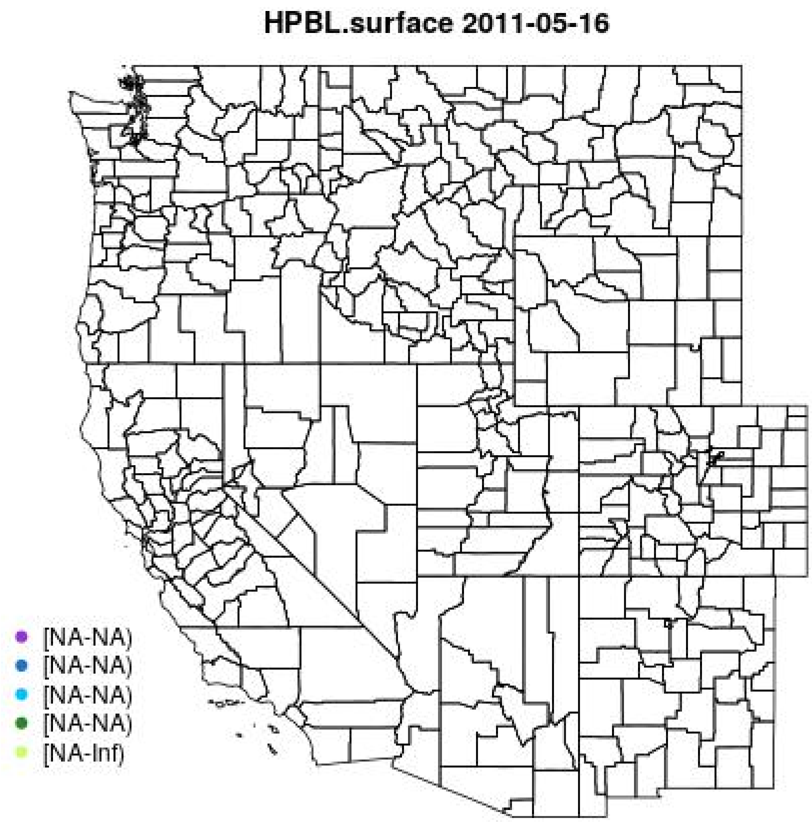

Figure 176: HPBL.surface 2011-05-16

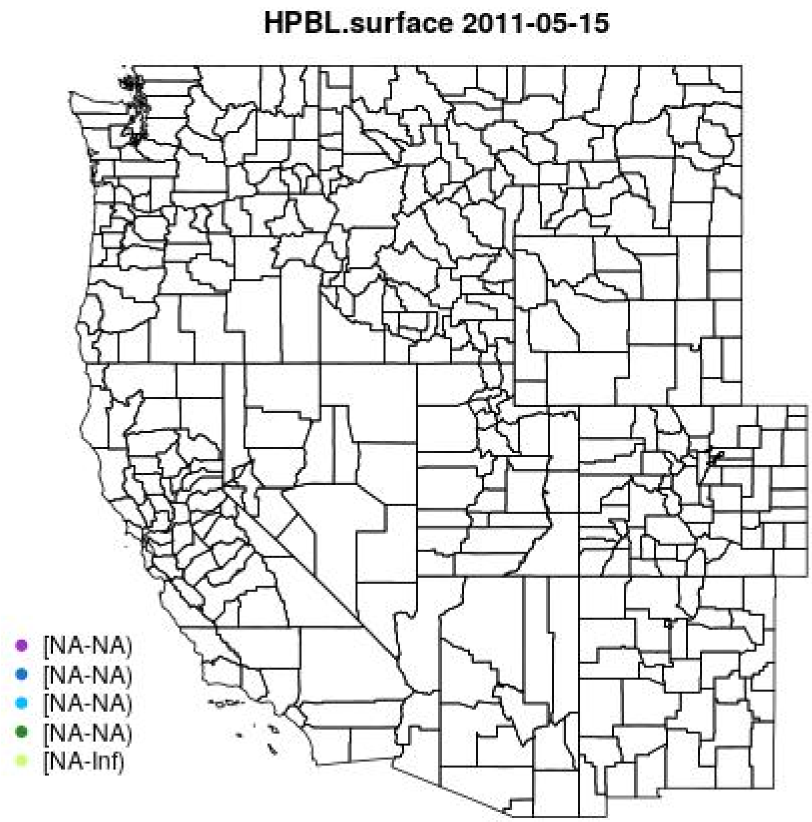

Figure 177: HPBL.surface 2011-05-15

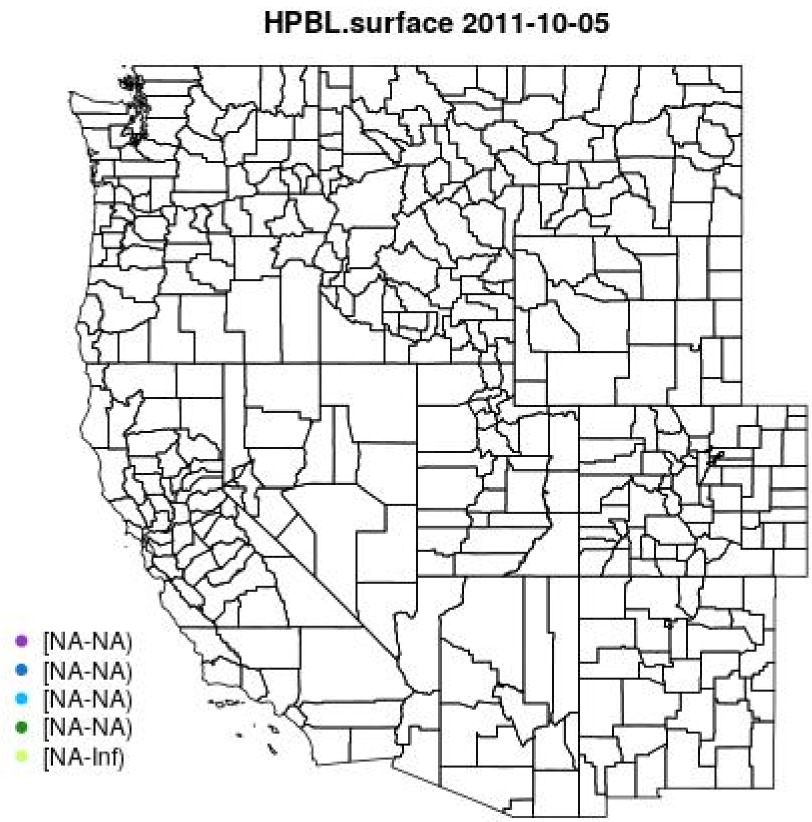

Figure 178: HPBL.surface 2011-10-05

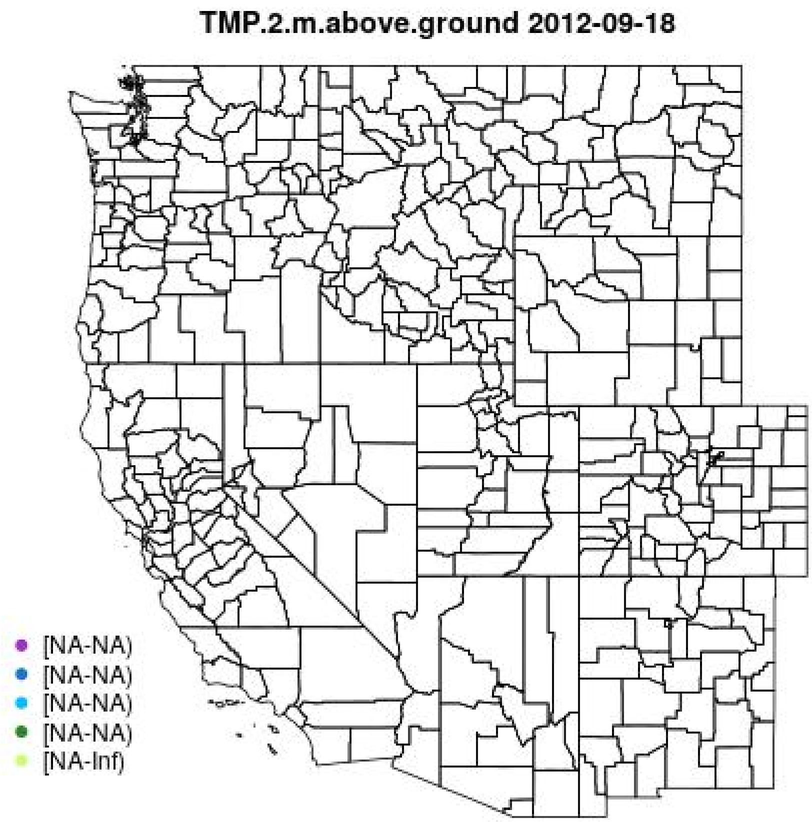

Figure 179: TMP.2.m.above.ground 2012-09-18

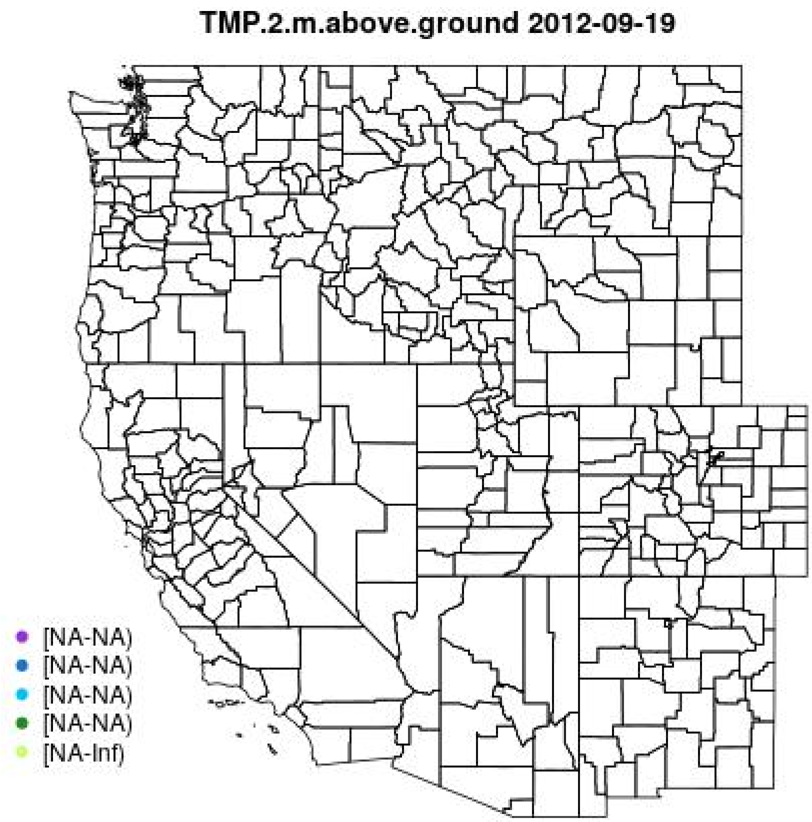

Figure 180: TMP.2.m.above.ground 2012-09-19

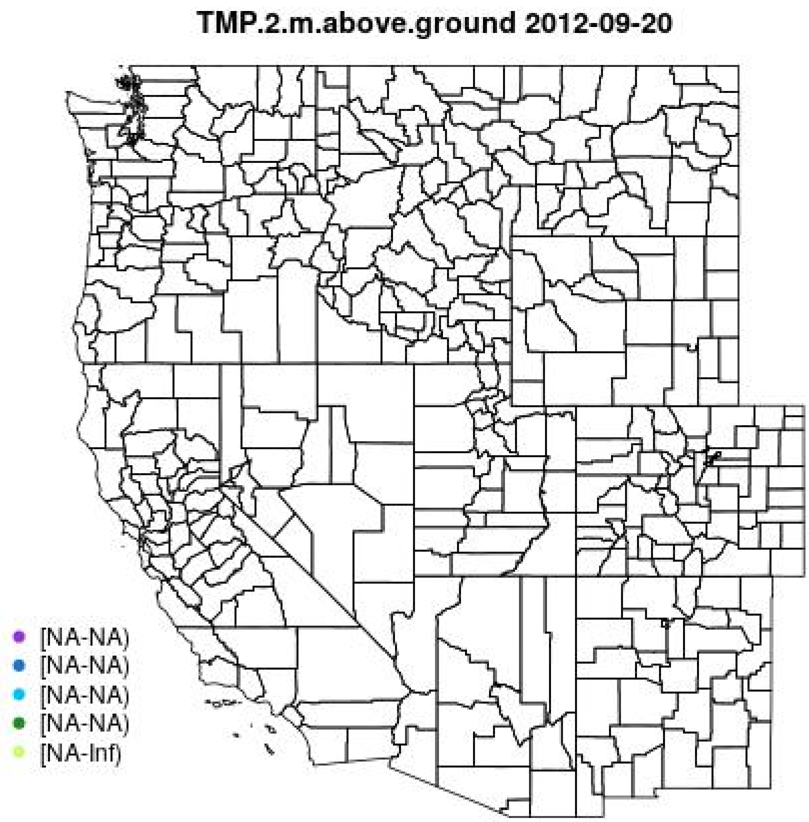

Figure 181: TMP.2.m.above.ground 2012-09-20

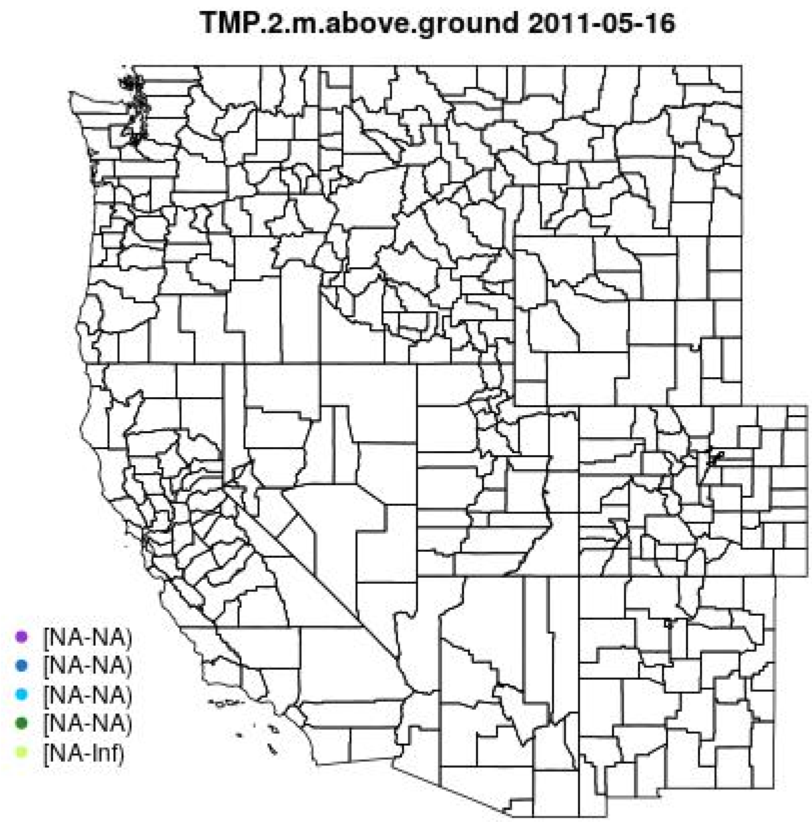

Figure 182: TMP.2.m.above.ground 2011-05-16

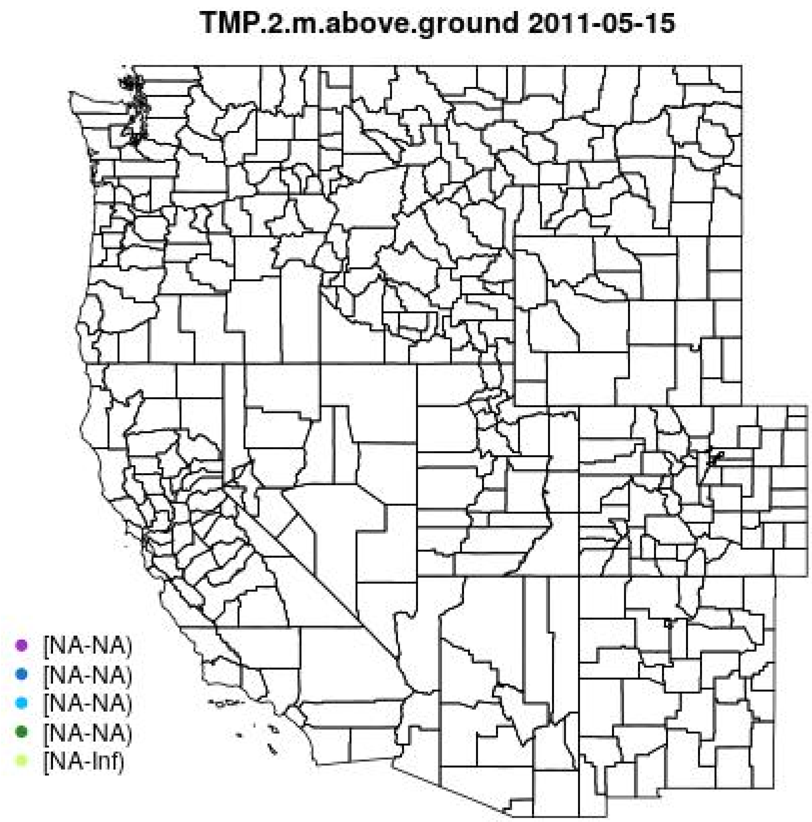

Figure 183: TMP.2.m.above.ground 2011-05-15

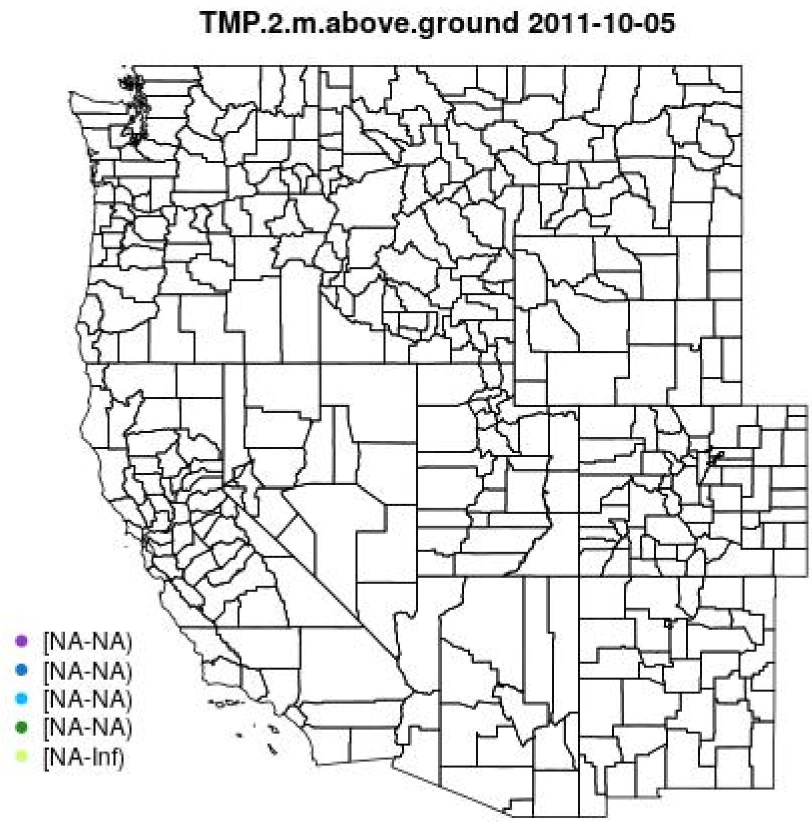

Figure 184: TMP.2.m.above.ground 2011-10-05

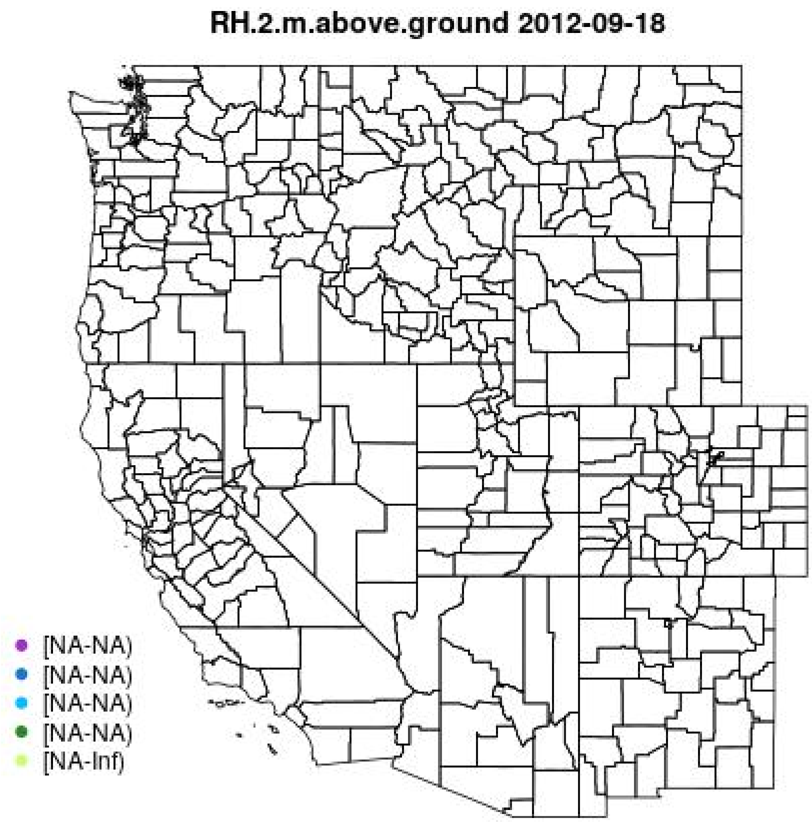

Figure 185: RH.2.m.above.ground 2012-09-18

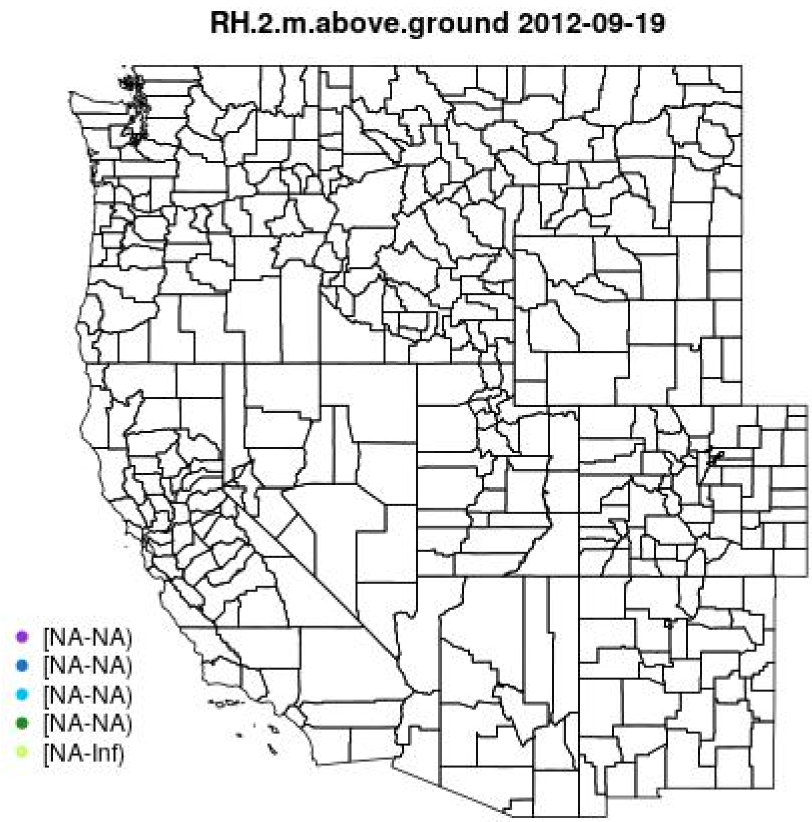

Figure 186: RH.2.m.above.ground 2012-09-19

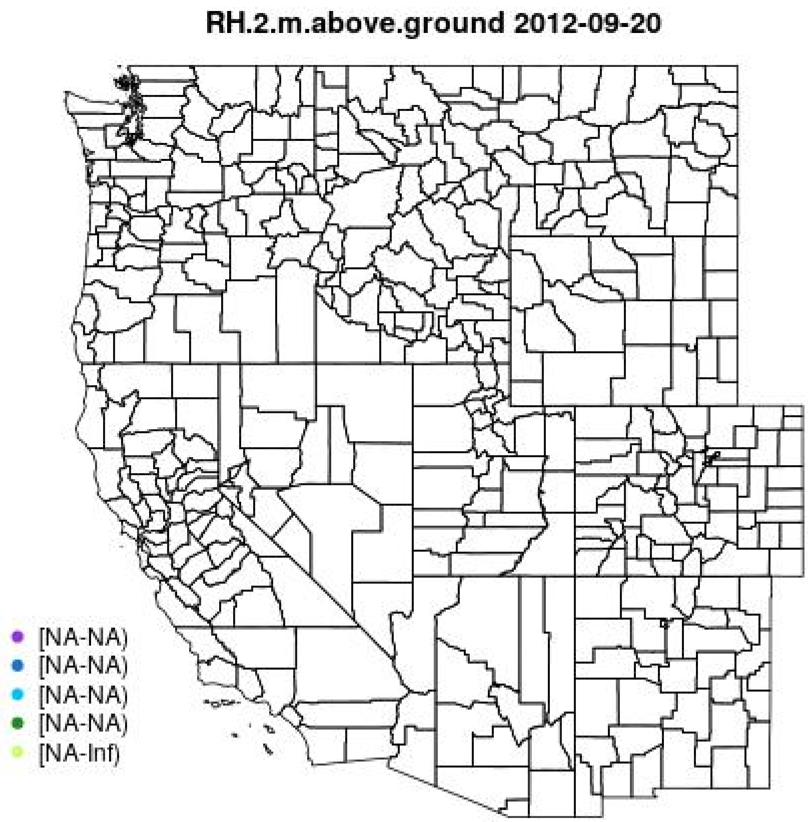

Figure 187: RH.2.m.above.ground 2012-09-20

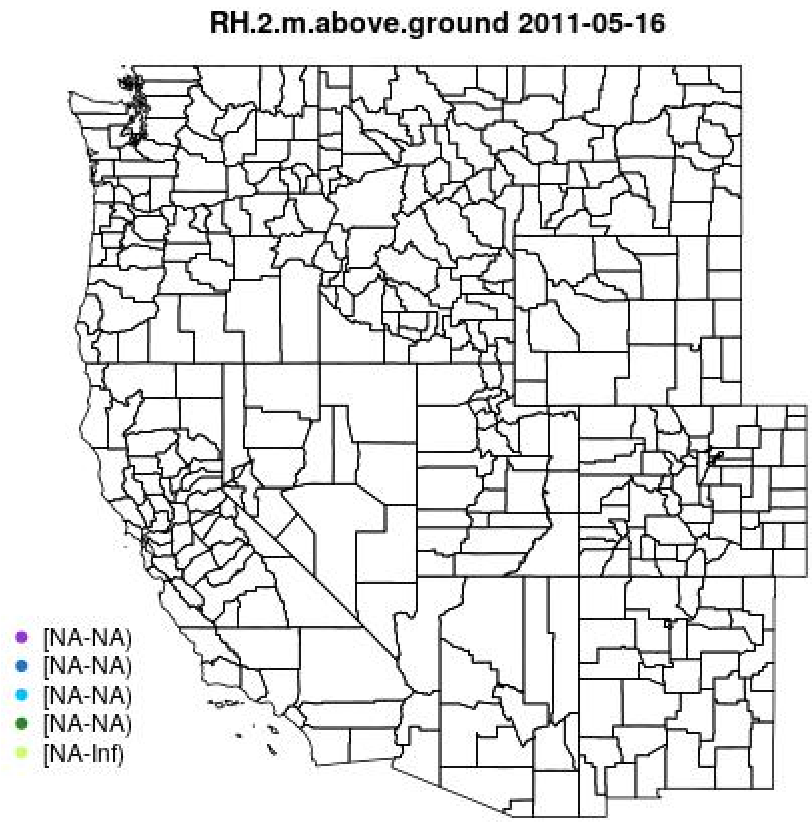

Figure 188: RH.2.m.above.ground 2011-05-16

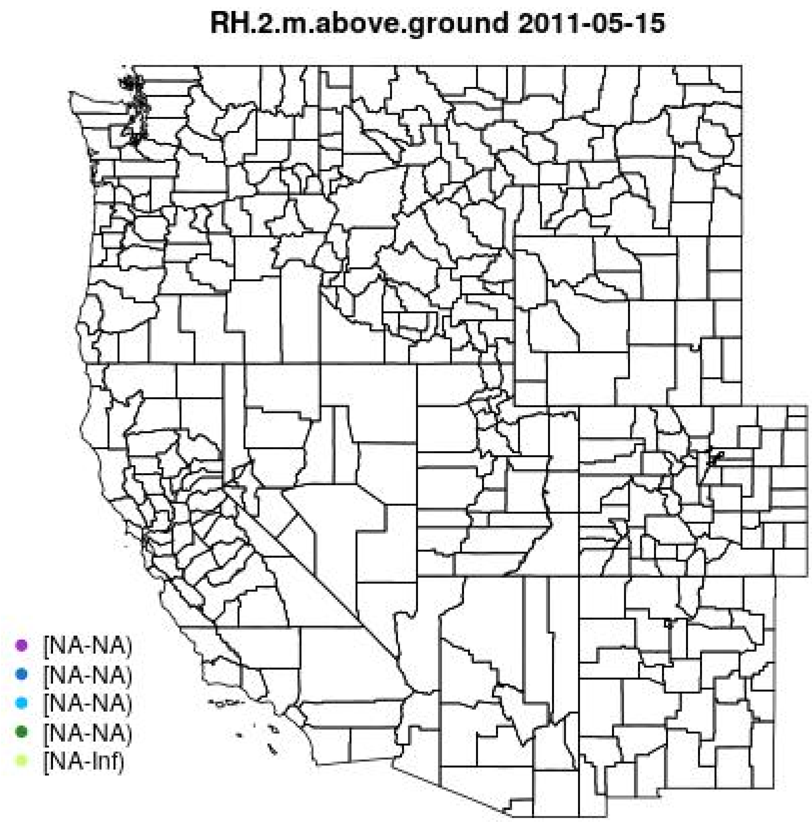

Figure 189: RH.2.m.above.ground 2011-05-15

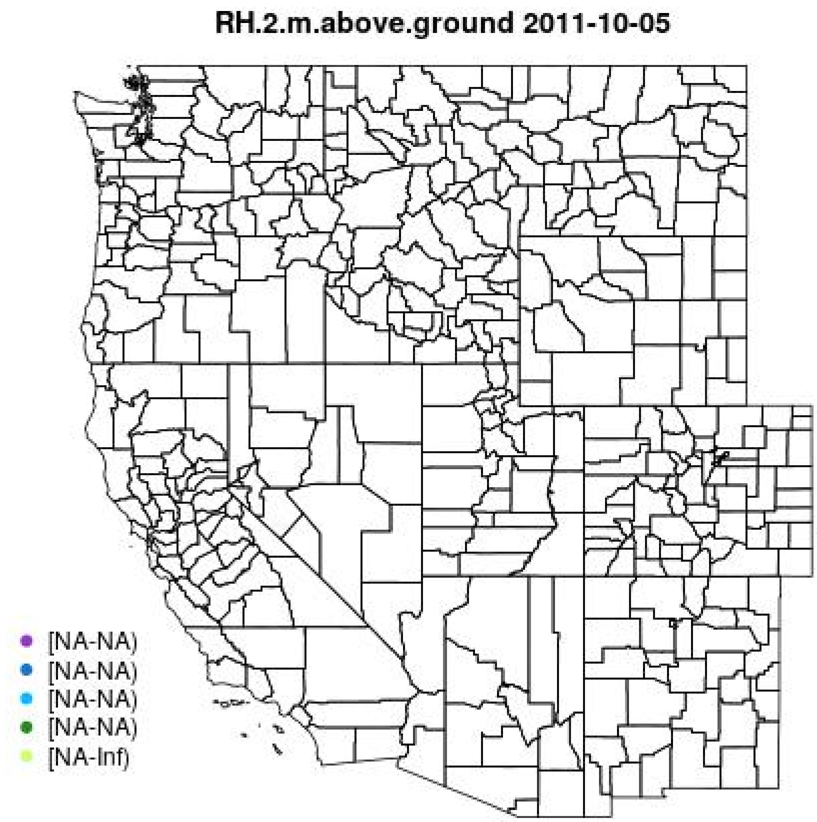

Figure 190: RH.2.m.above.ground 2011-10-05

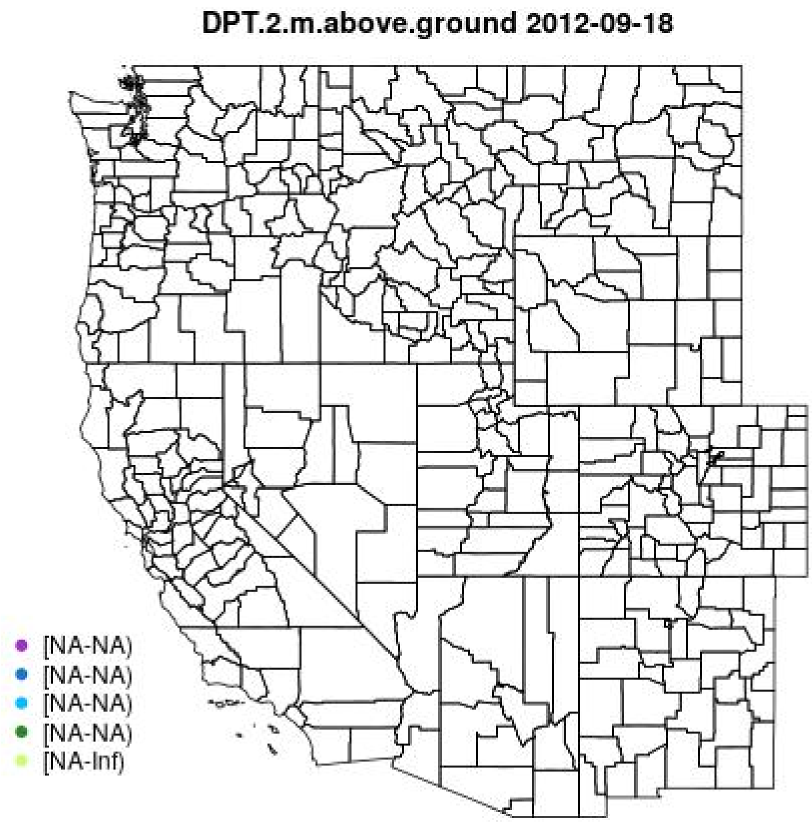

Figure 191: DPT.2.m.above.ground 2012-09-18

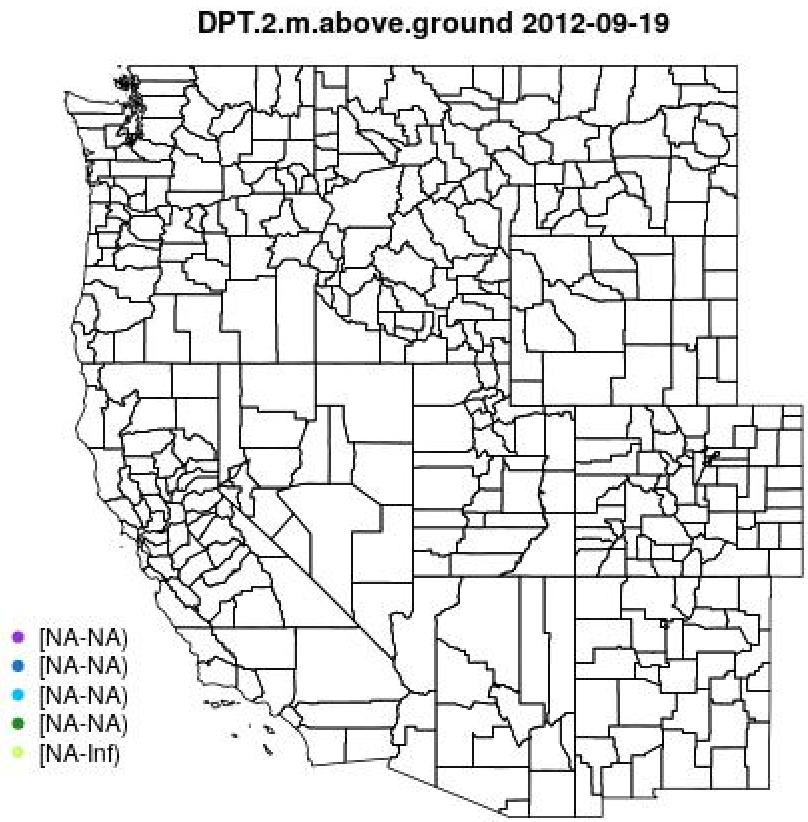

Figure 192: DPT.2.m.above.ground 2012-09-19

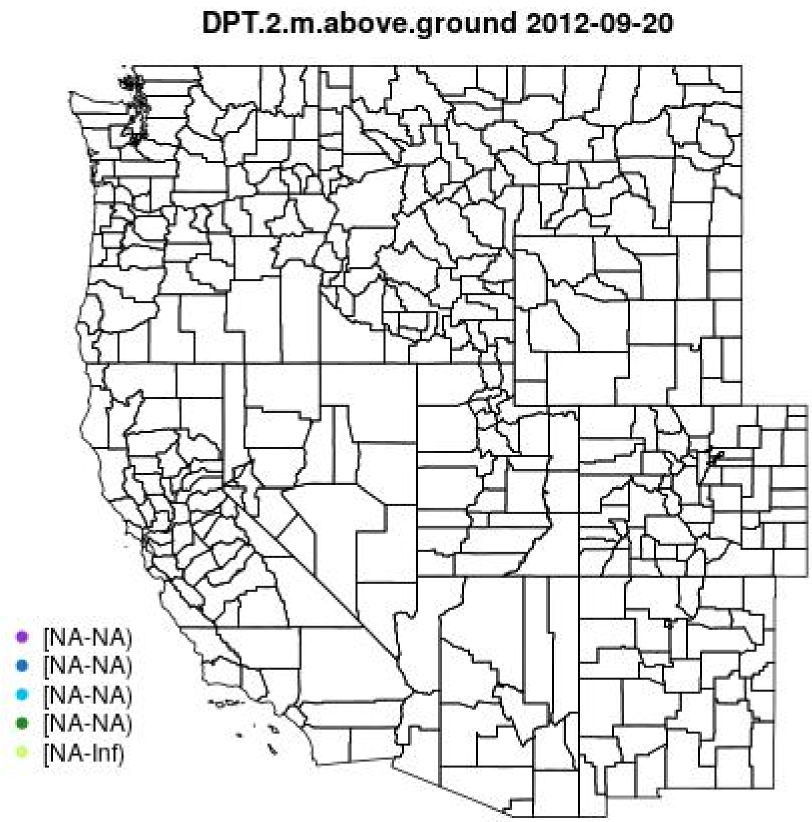

Figure 193: DPT.2.m.above.ground 2012-09-20

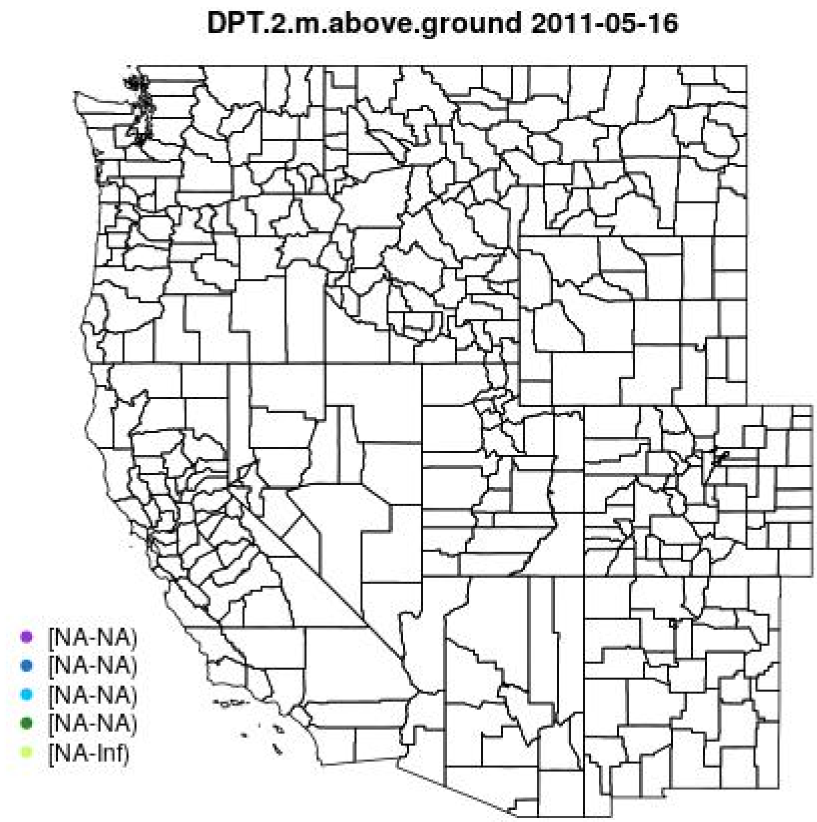

Figure 194: DPT.2.m.above.ground 2011-05-16

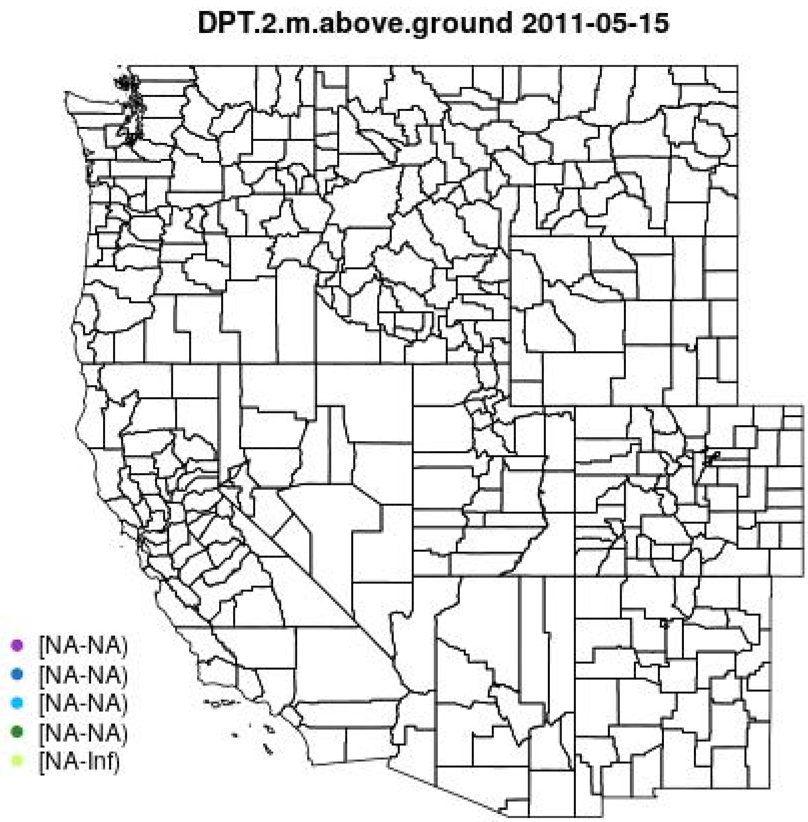

Figure 195: DPT.2.m.above.ground 2011-05-15

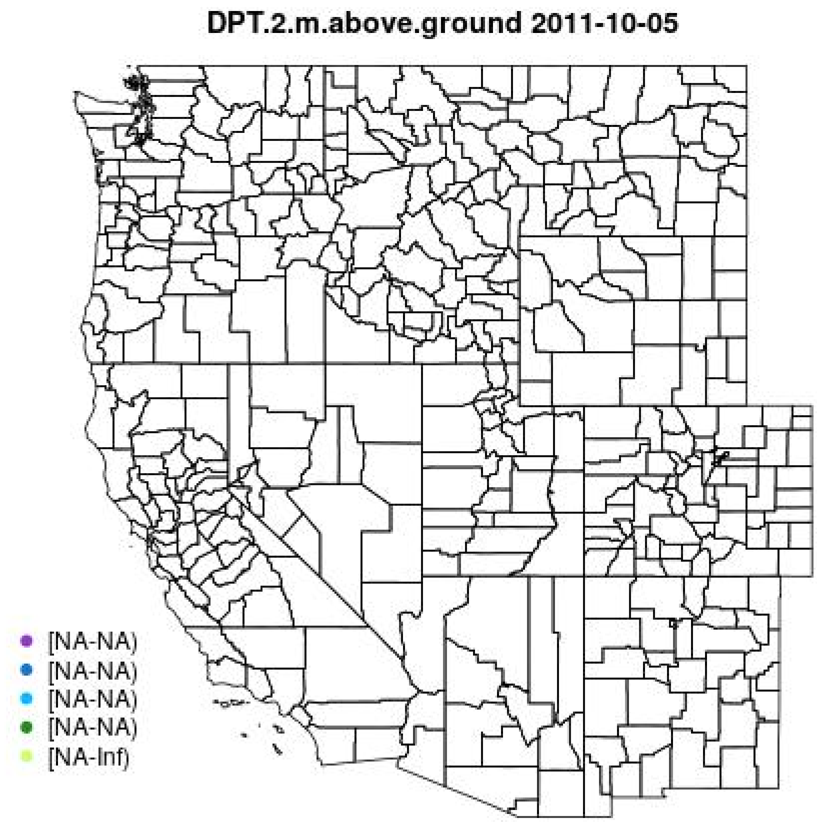

Figure 196: DPT.2.m.above.ground 2011-10-05

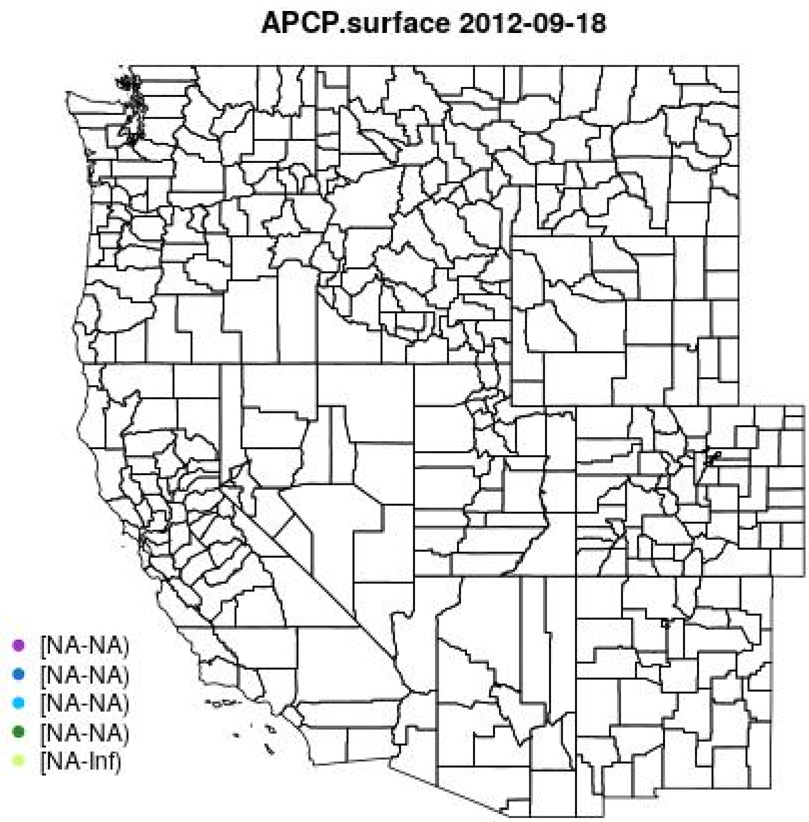

Figure 197: APCP.surface 2012-09-18

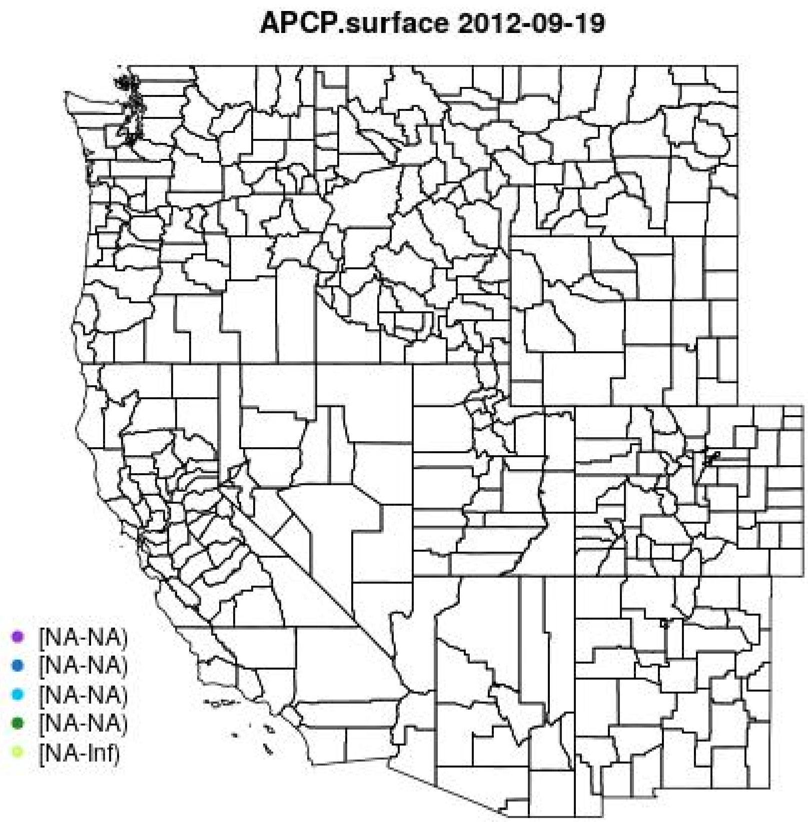

Figure 198: APCP.surface 2012-09-19

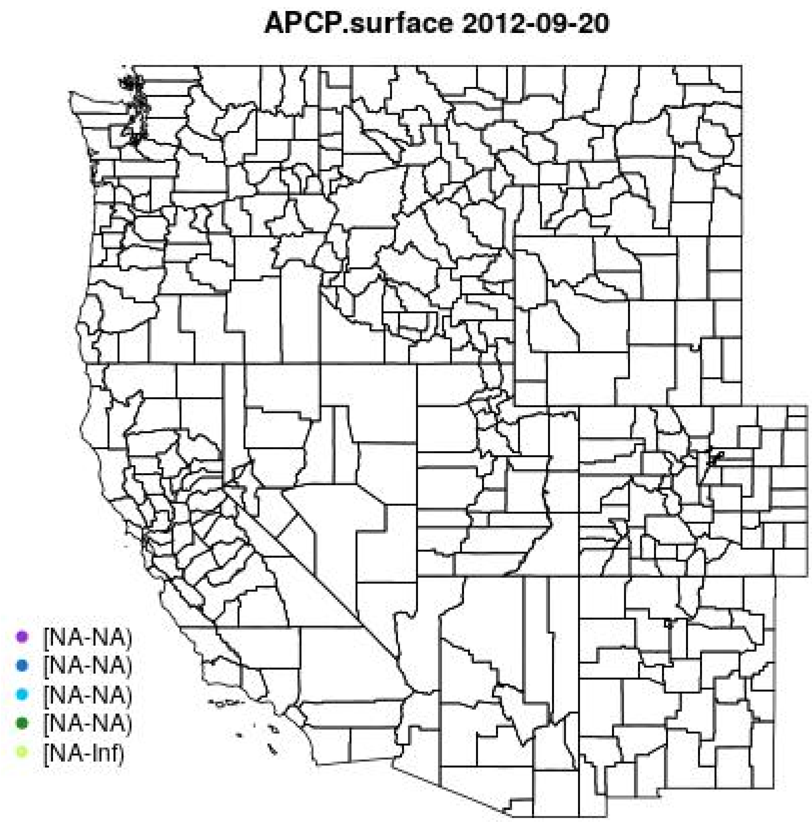

Figure 199: APCP.surface 2012-09-20

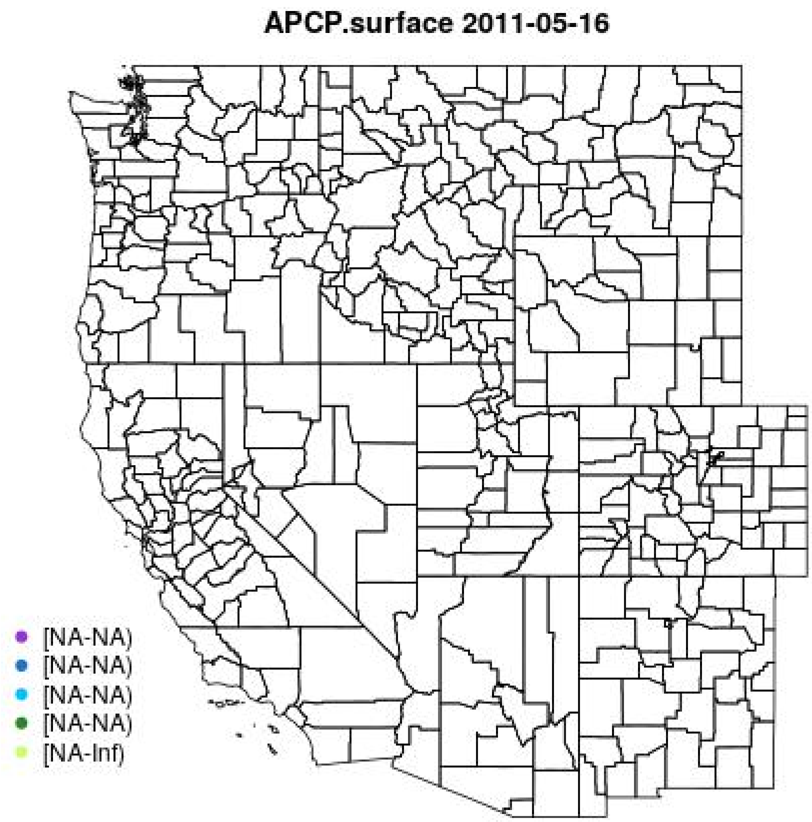

Figure 200: APCP.surface 2011-05-16

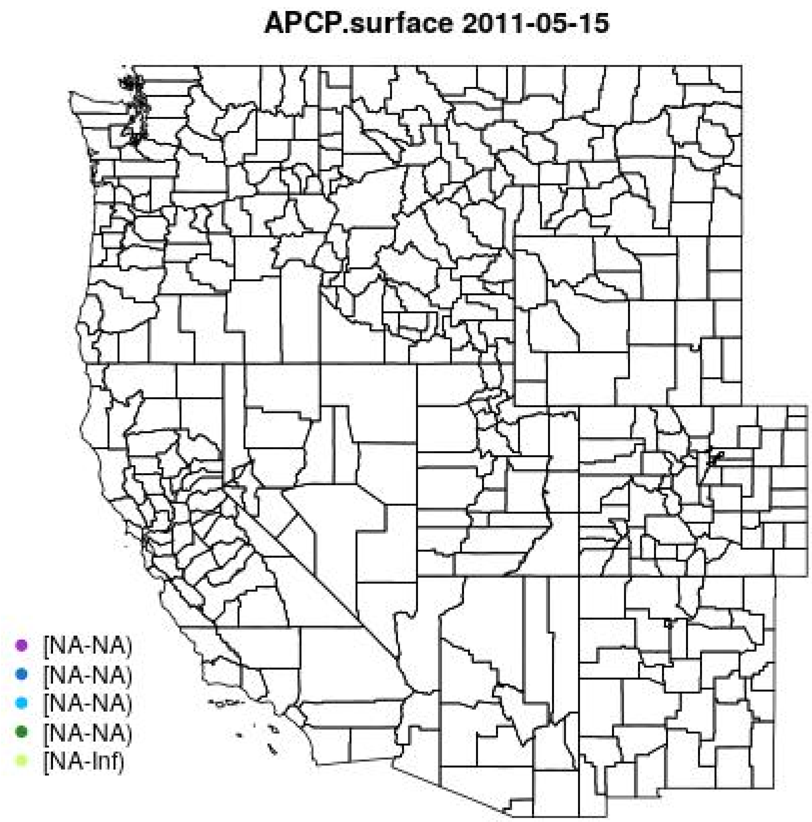

Figure 201: APCP.surface 2011-05-15

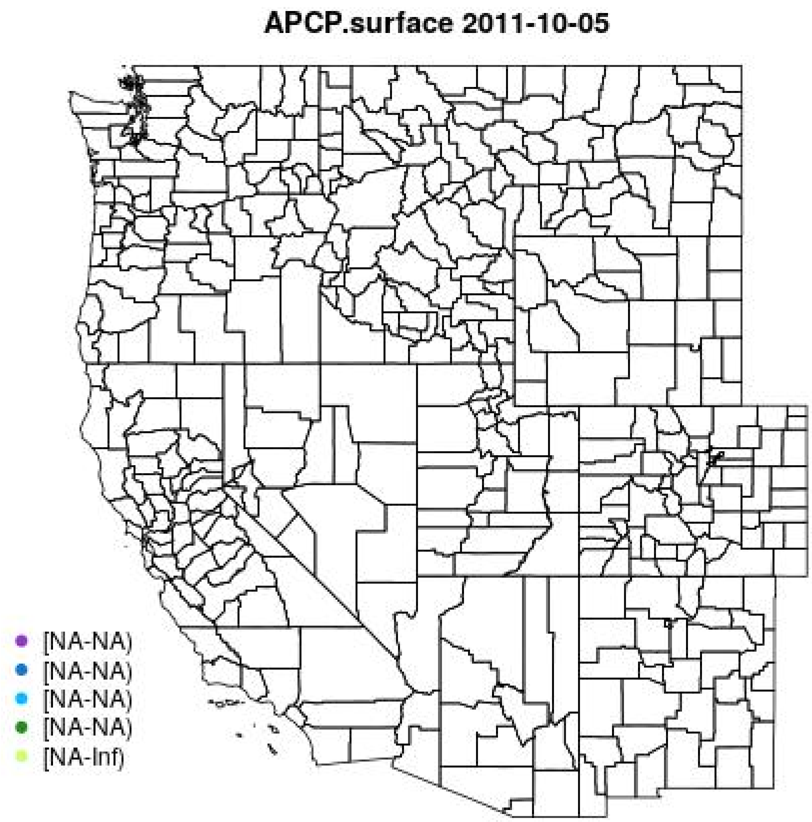

Figure 202: APCP.surface 2011-10-05

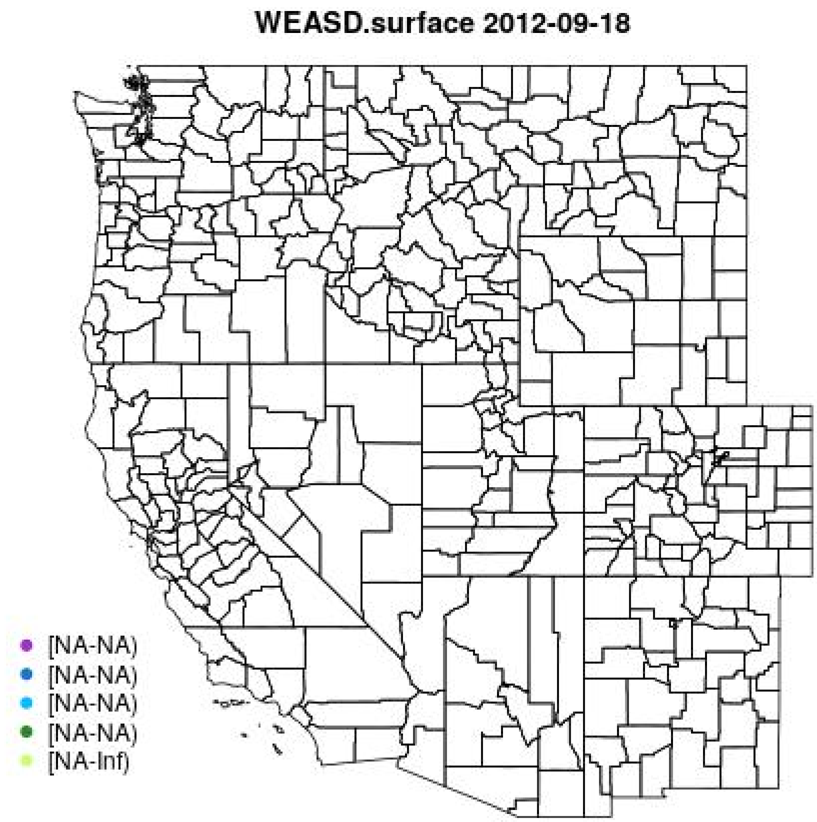

Figure 203: WEASD.surface 2012-09-18

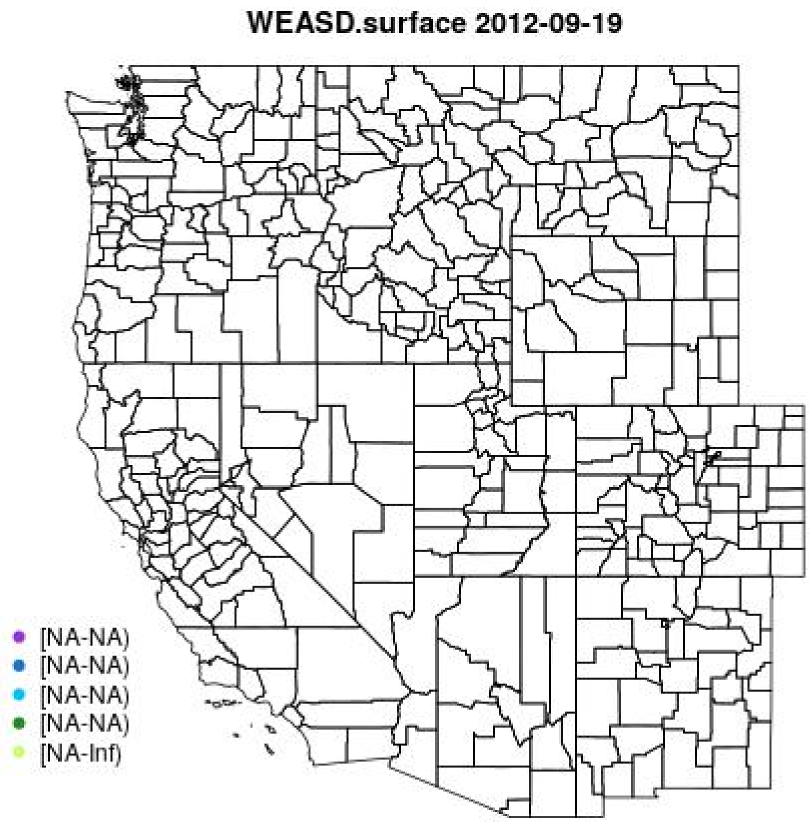

Figure 204: WEASD.surface 2012-09-19

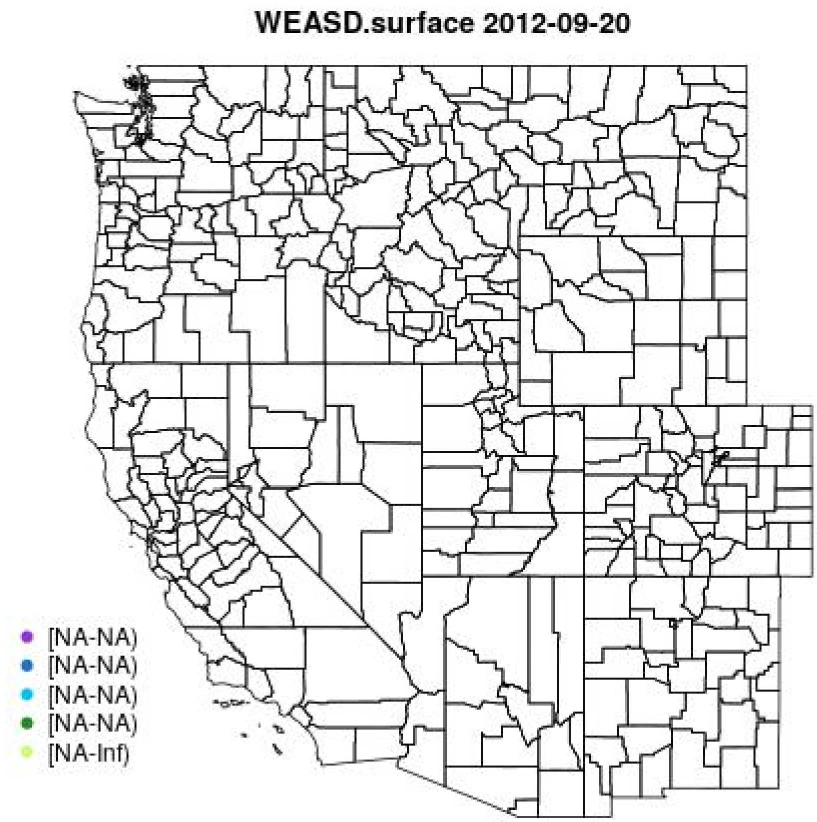

Figure 205: WEASD.surface 2012-09-20

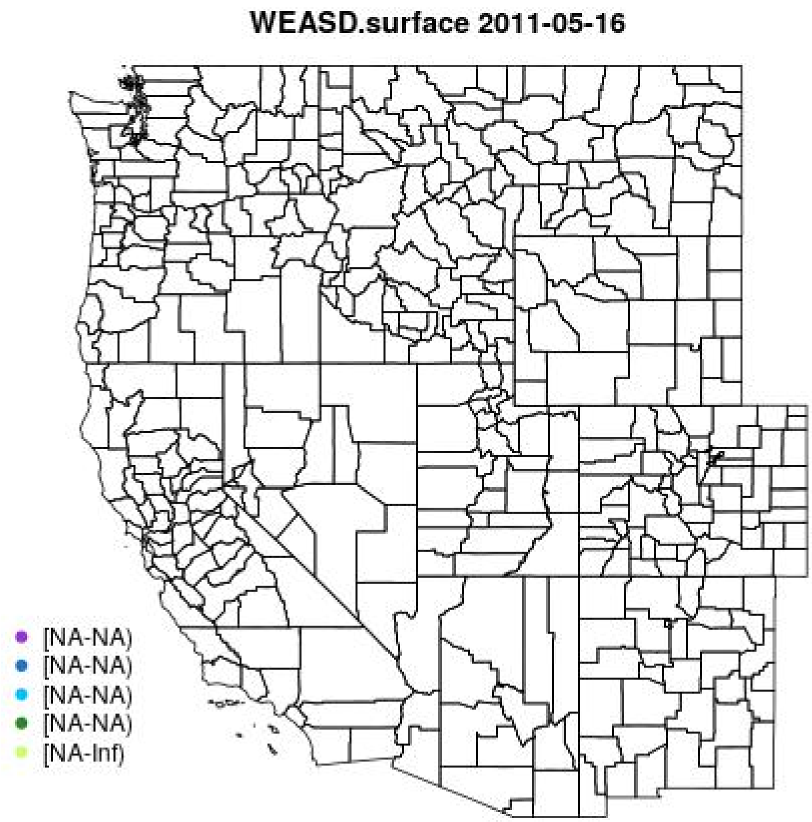

Figure 206: WEASD.surface 2011-05-16

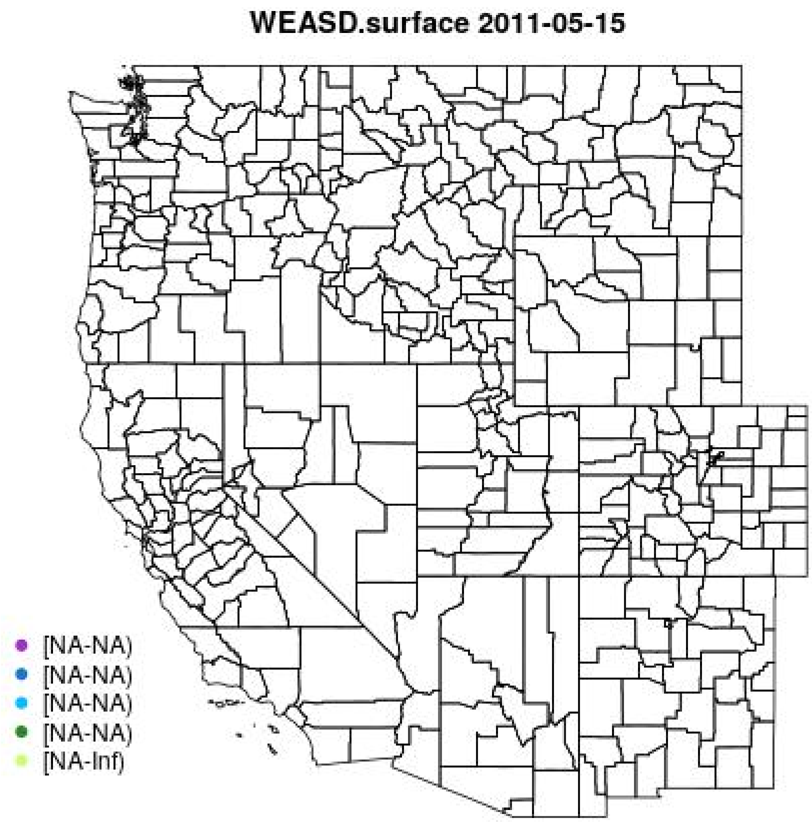

Figure 207: WEASD.surface 2011-05-15

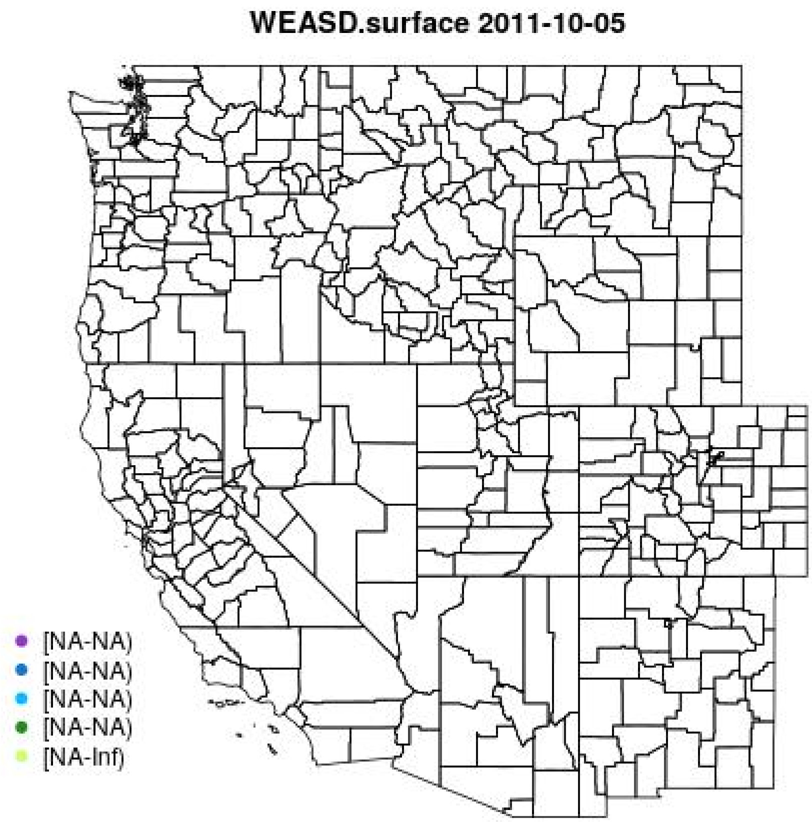

Figure 208: WEASD.surface 2011-10-05

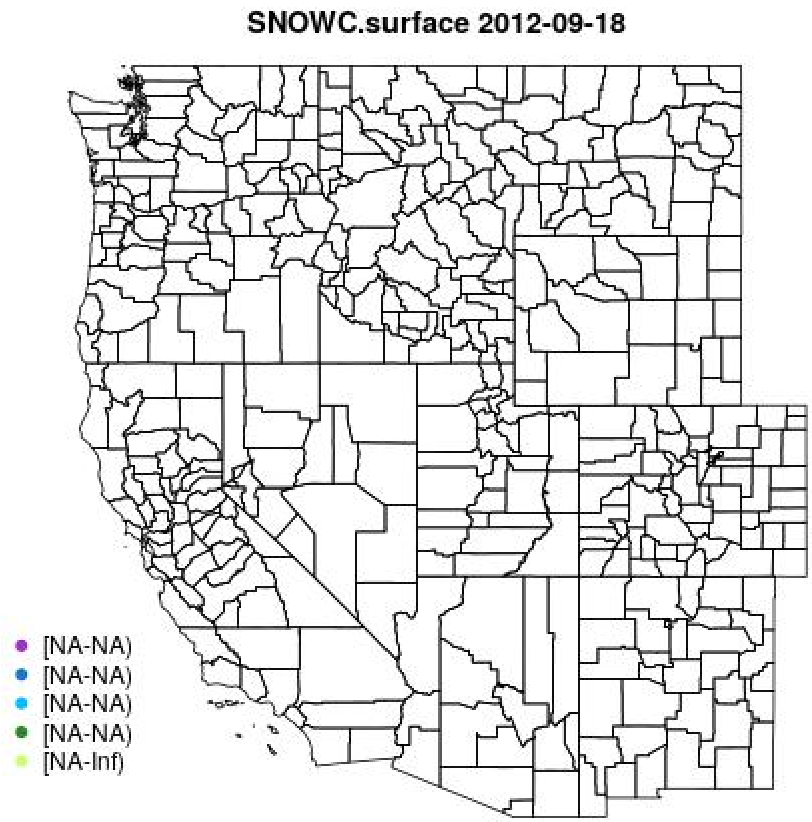

Figure 209: SNOWC.surface 2012-09-18

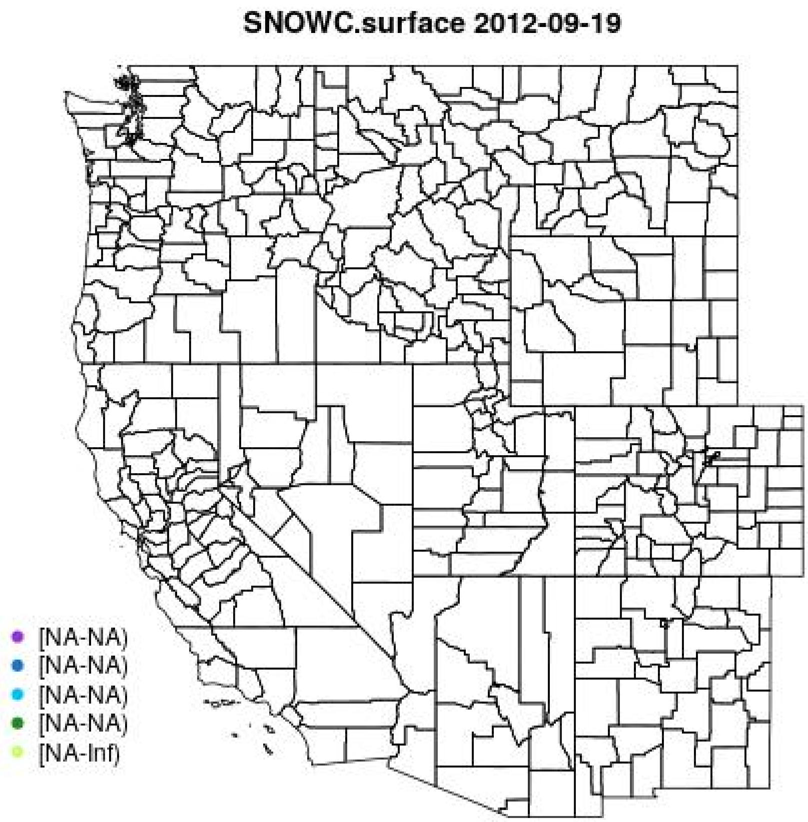

Figure 210: SNOWC.surface 2012-09-19

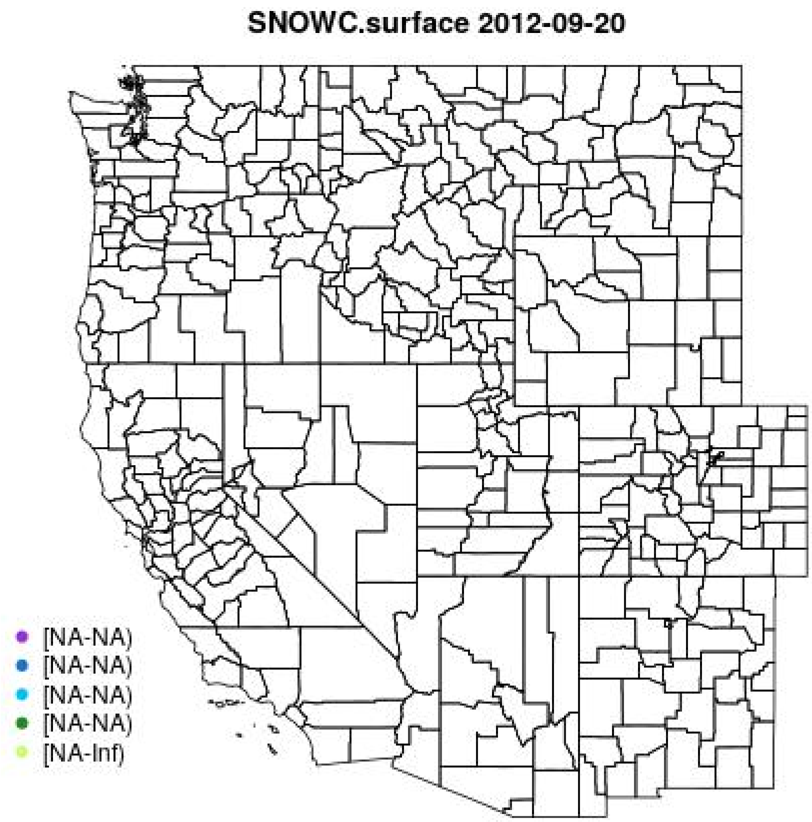

Figure 211: SNOWC.surface 2012-09-20

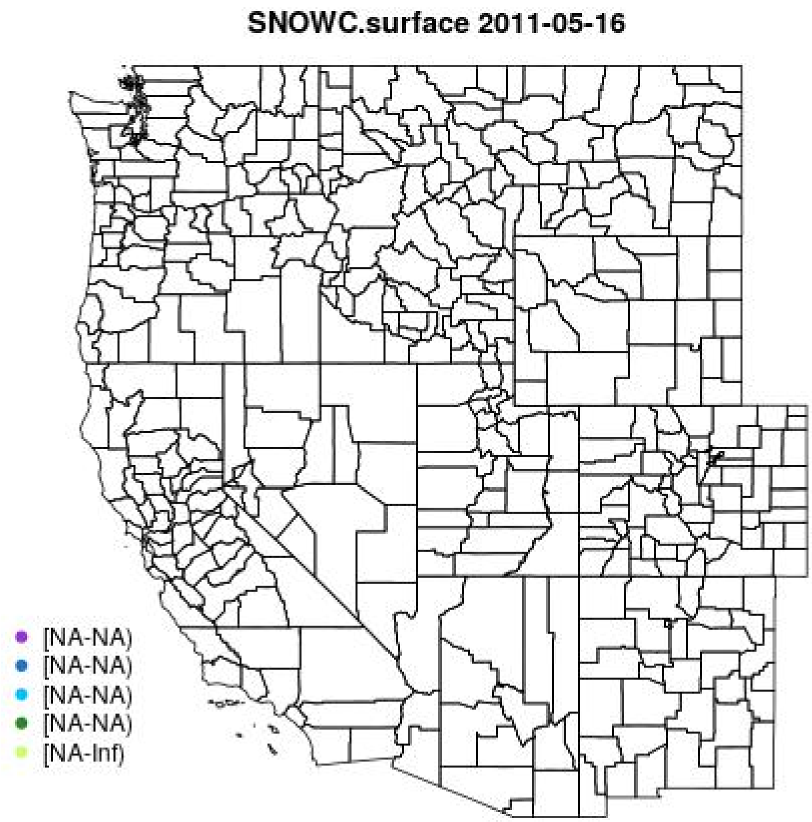

Figure 212: SNOWC.surface 2011-05-16

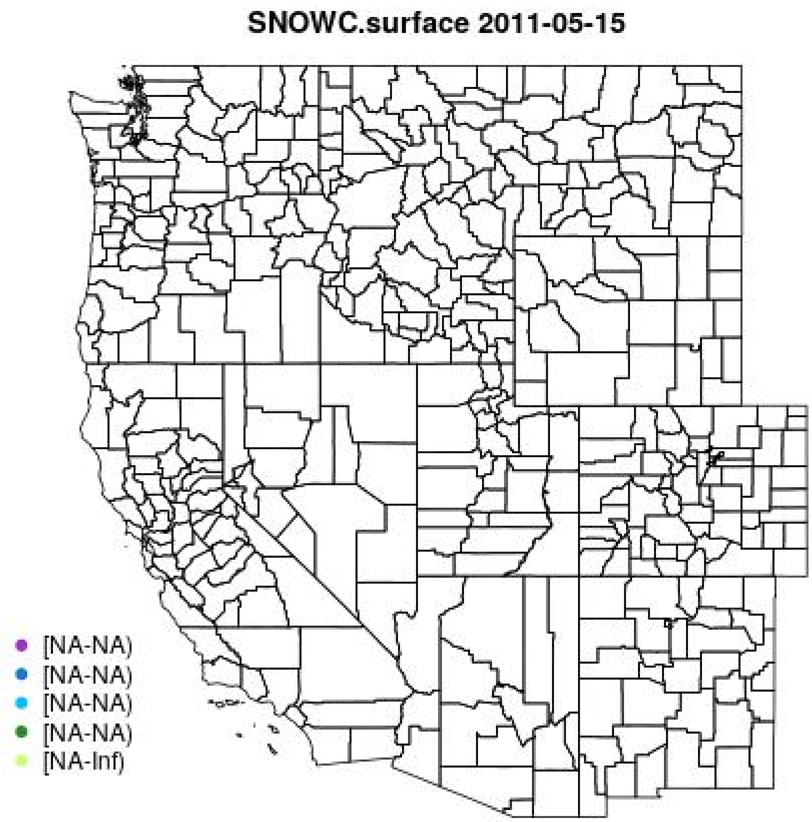

Figure 213: SNOWC.surface 2011-05-15

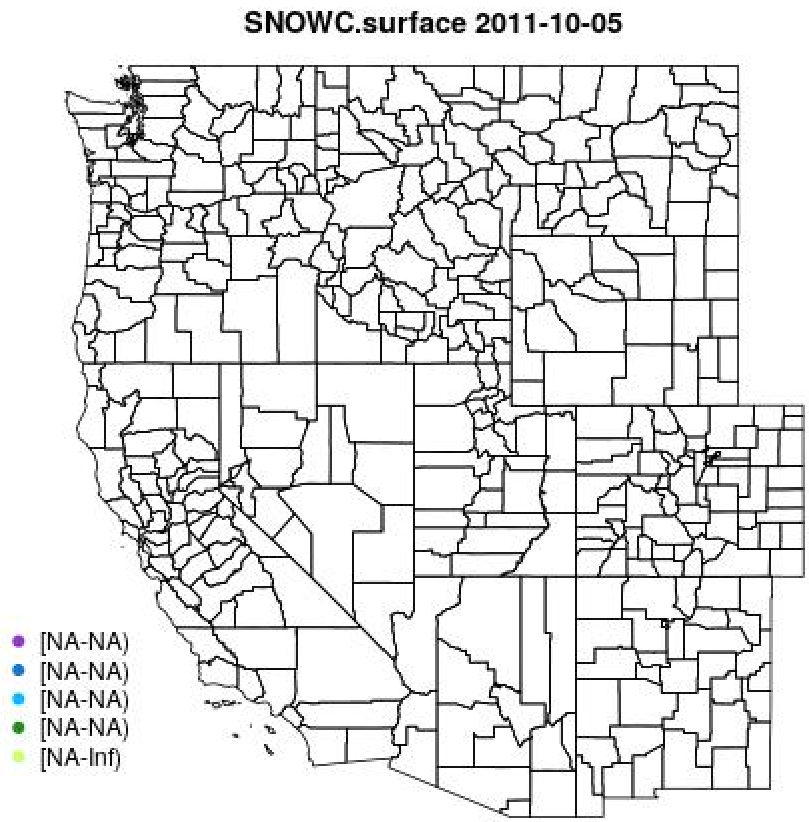

Figure 214: SNOWC.surface 2011-10-05

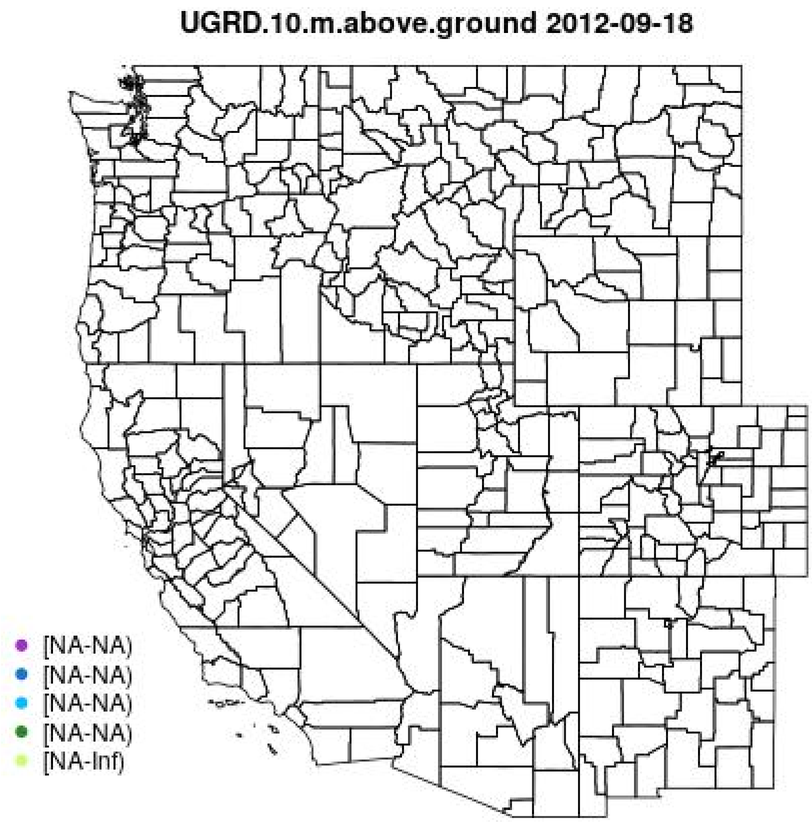

Figure 215: UGRD.10.m.above.ground 2012-09-18

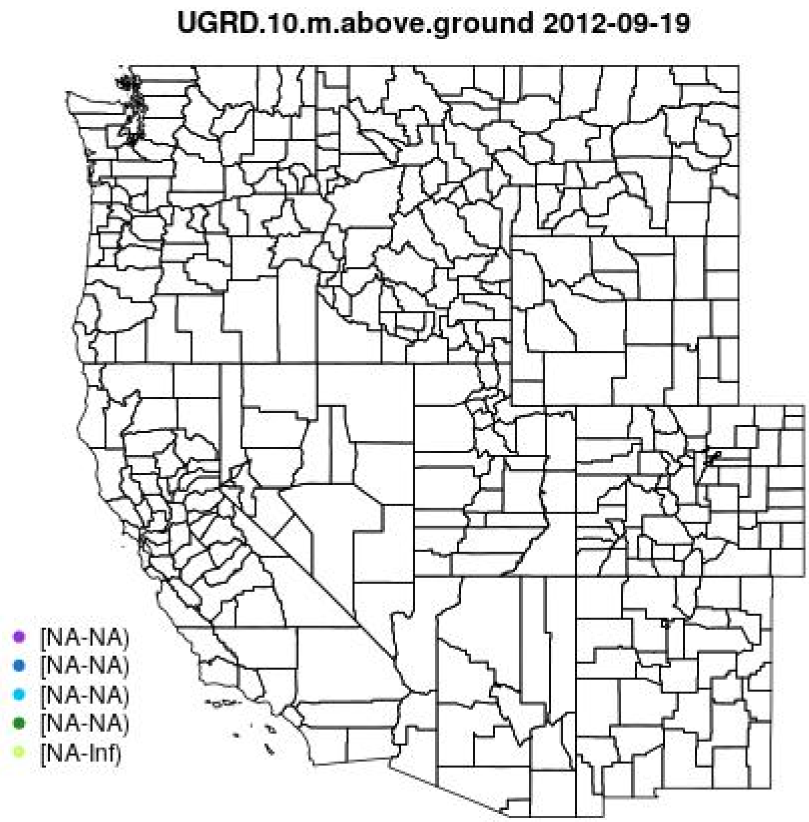

Figure 216: UGRD.10.m.above.ground 2012-09-19

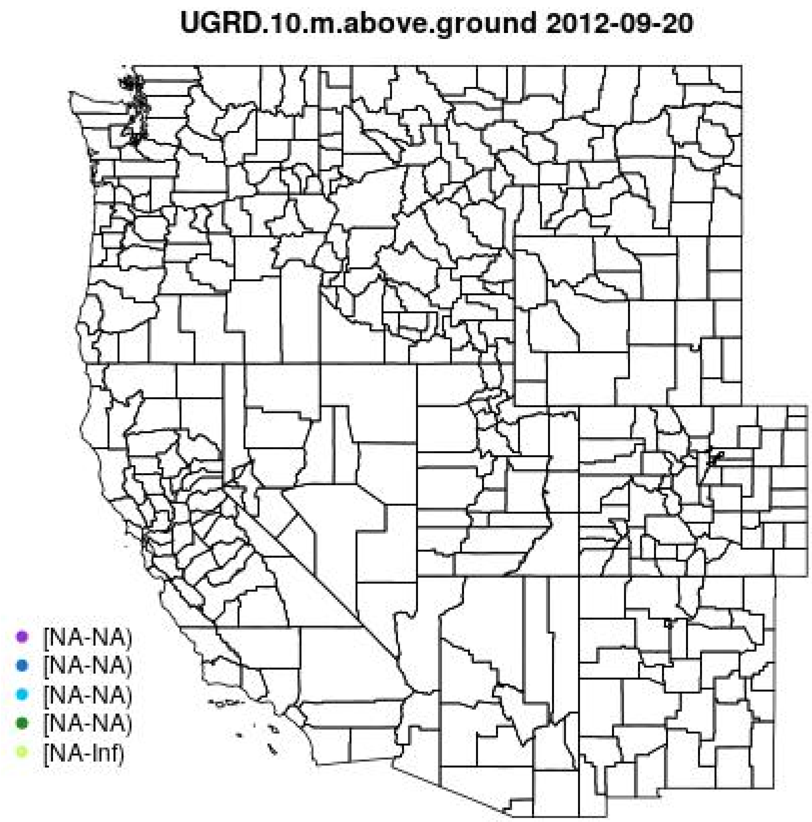

Figure 217: UGRD.10.m.above.ground 2012-09-20

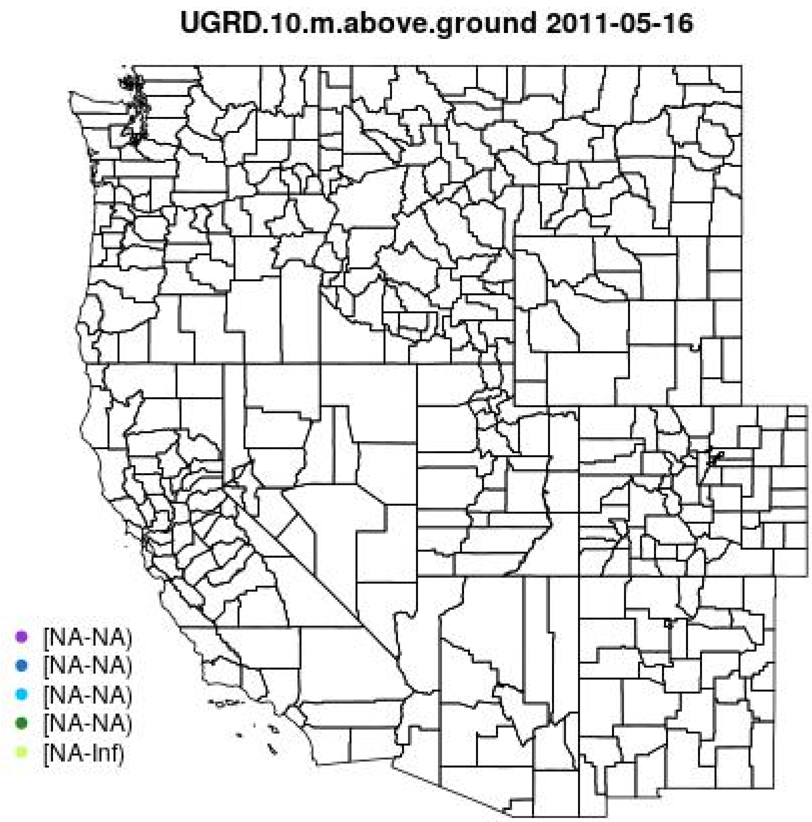

Figure 218: UGRD.10.m.above.ground 2011-05-16

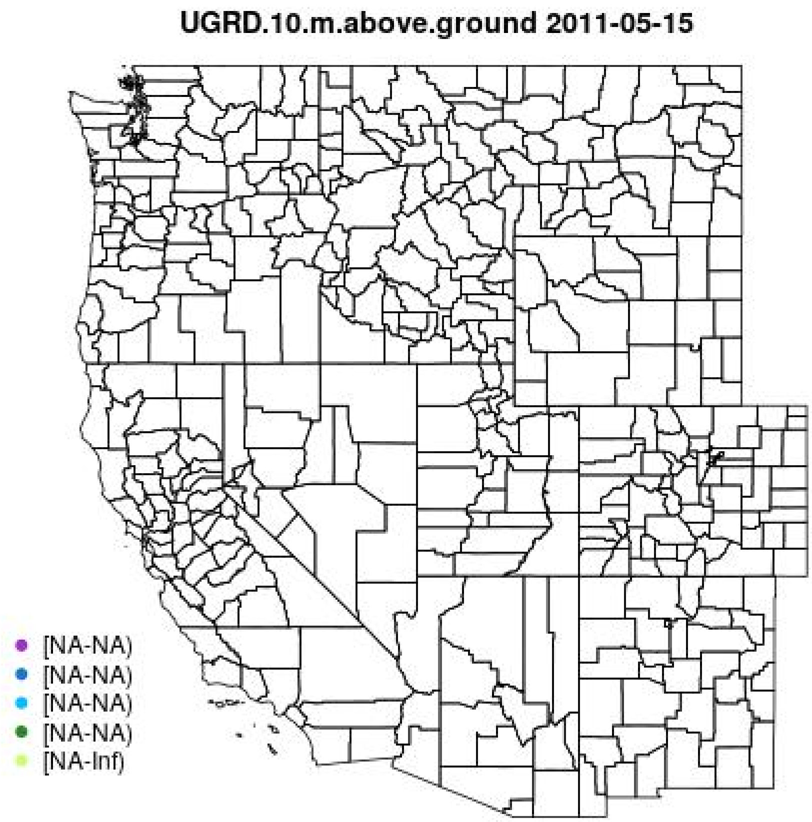

Figure 219: UGRD.10.m.above.ground 2011-05-15

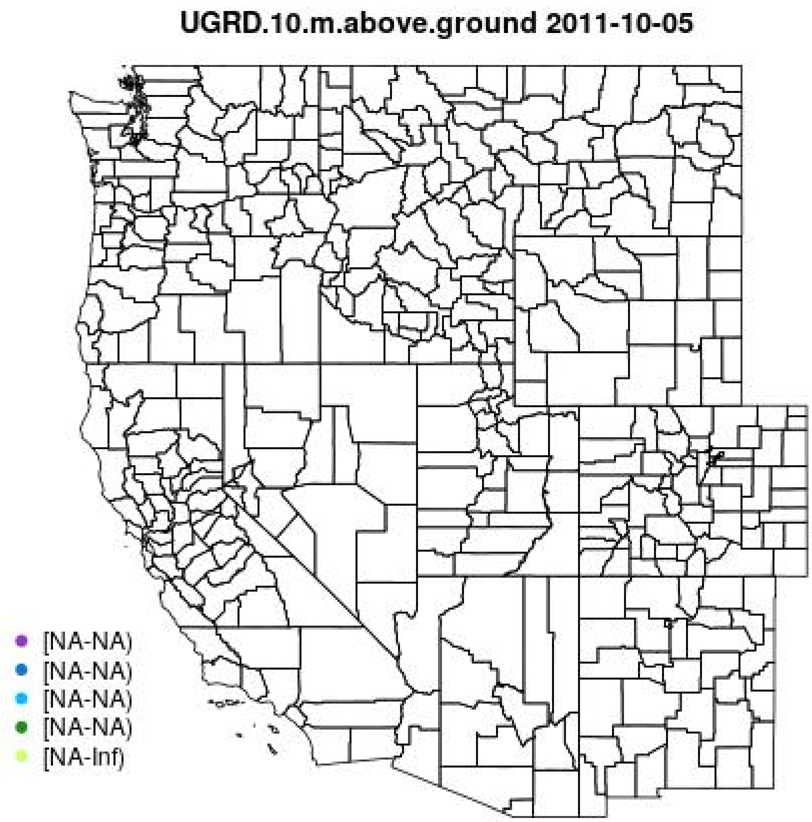

Figure 220: UGRD.10.m.above.ground 2011-10-05

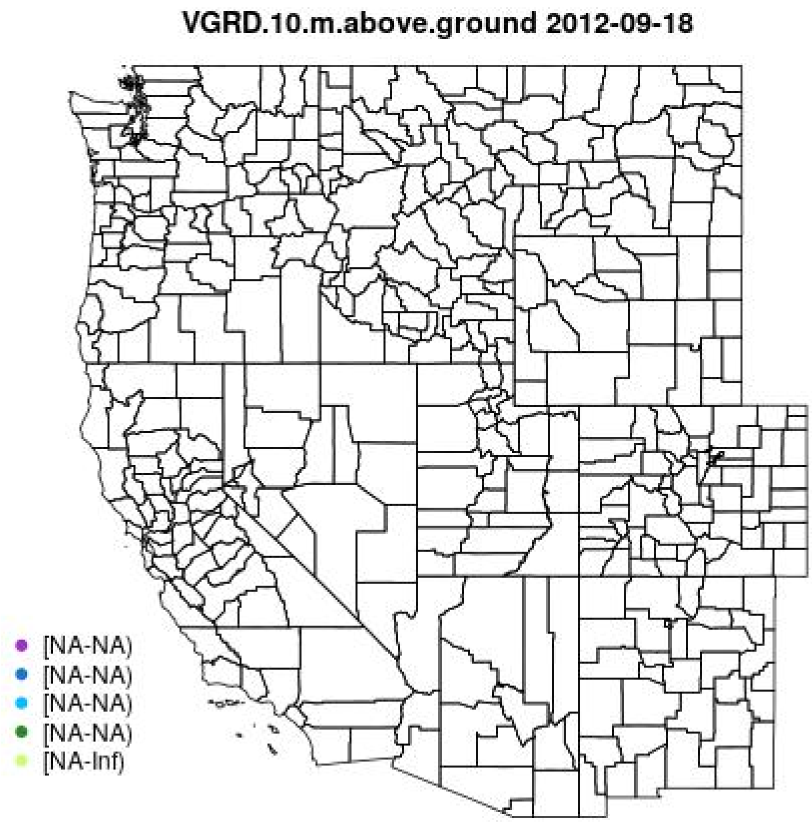

Figure 221: VGRD.10.m.above.ground 2012-09-18

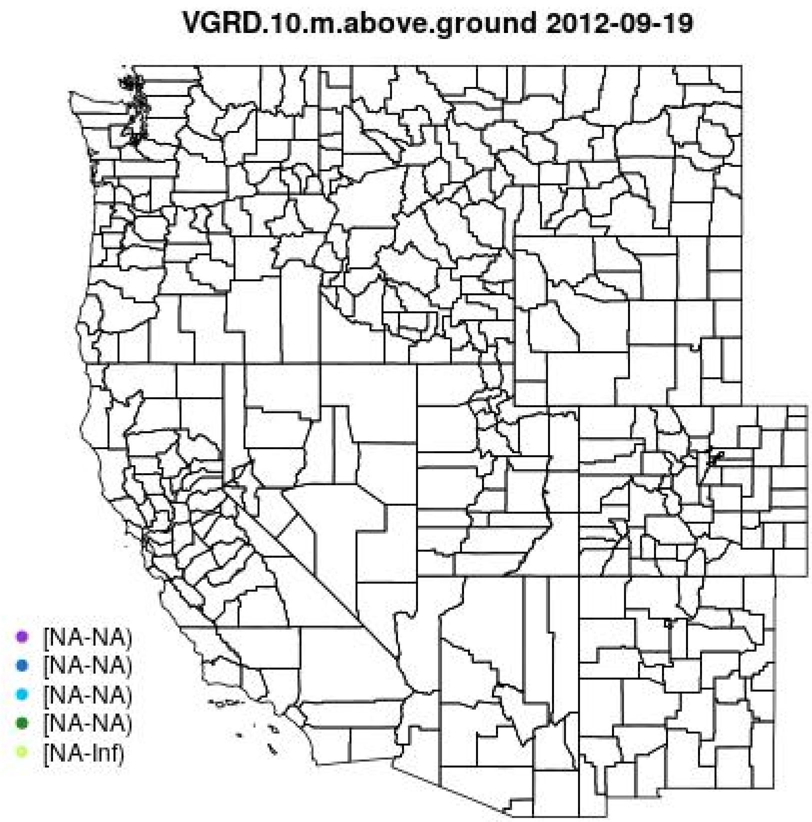

Figure 222: VGRD.10.m.above.ground 2012-09-19

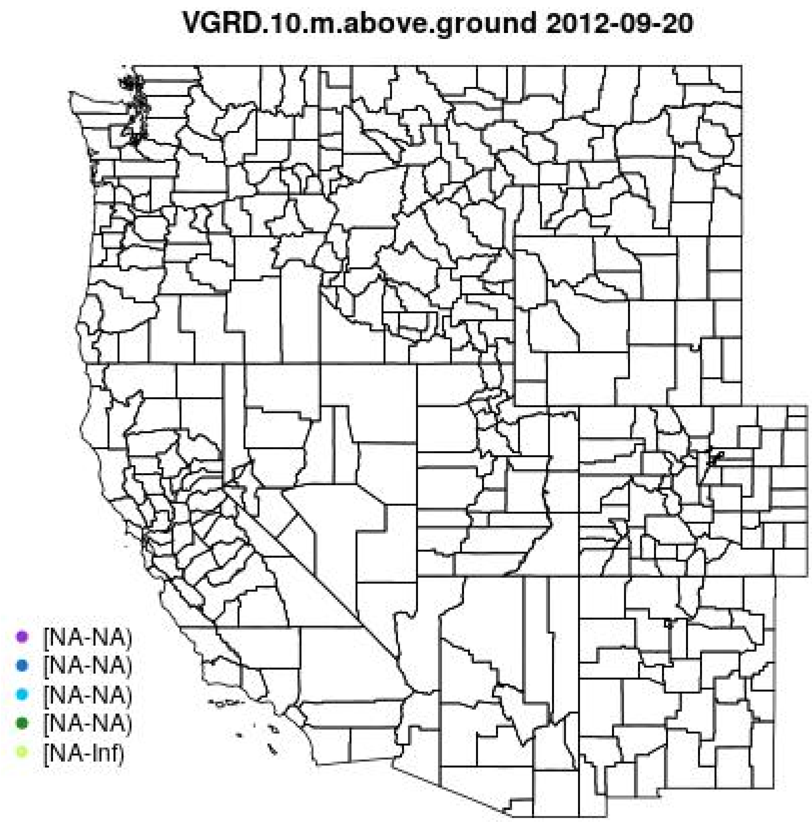

Figure 223: VGRD.10.m.above.ground 2012-09-20

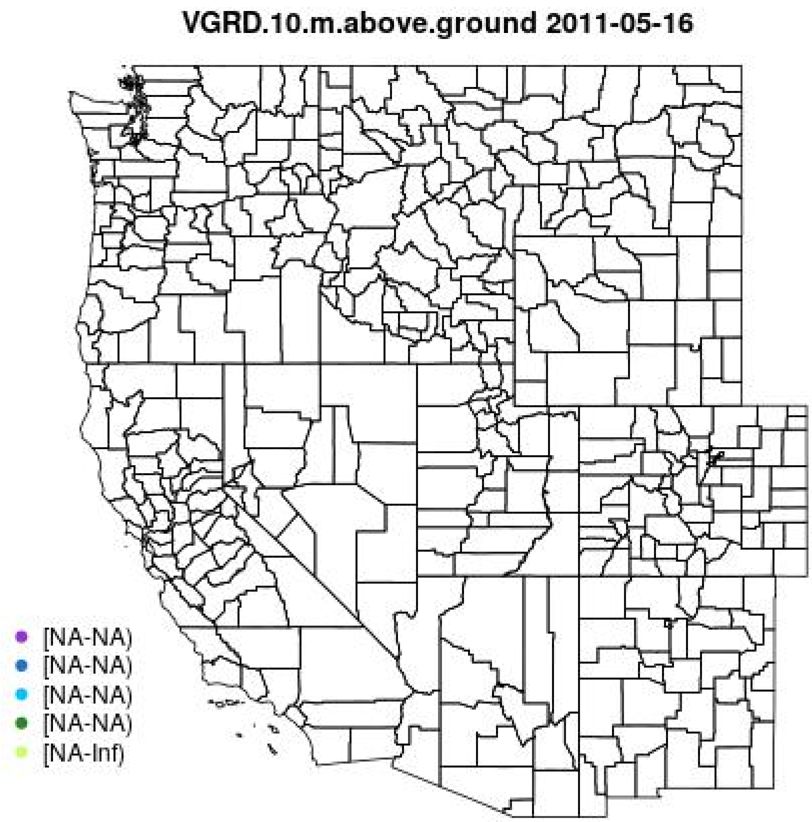

Figure 224: VGRD.10.m.above.ground 2011-05-16

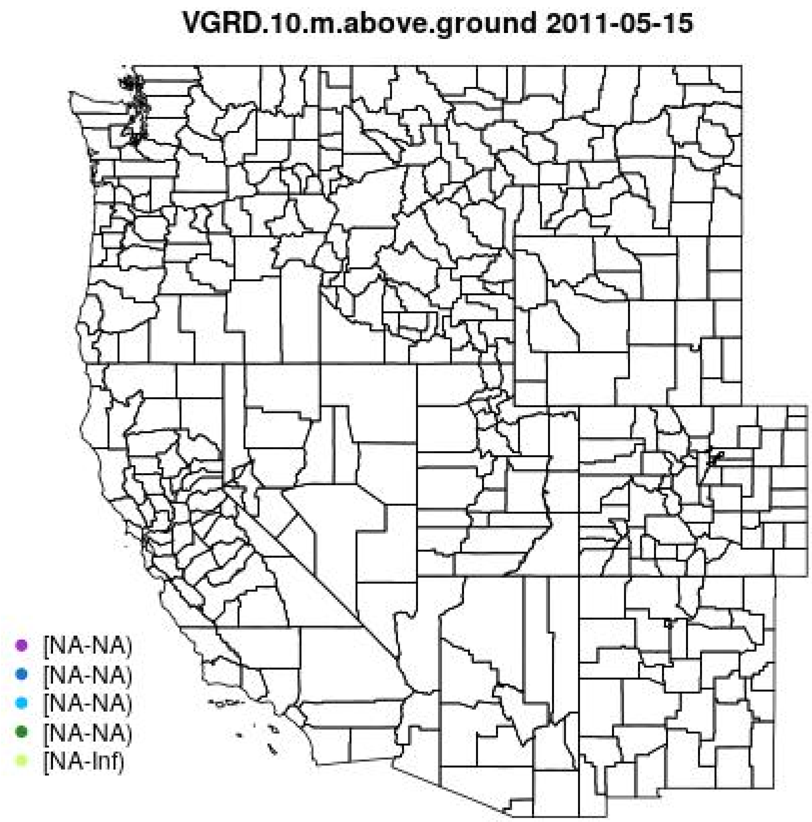

Figure 225: VGRD.10.m.above.ground 2011-05-15

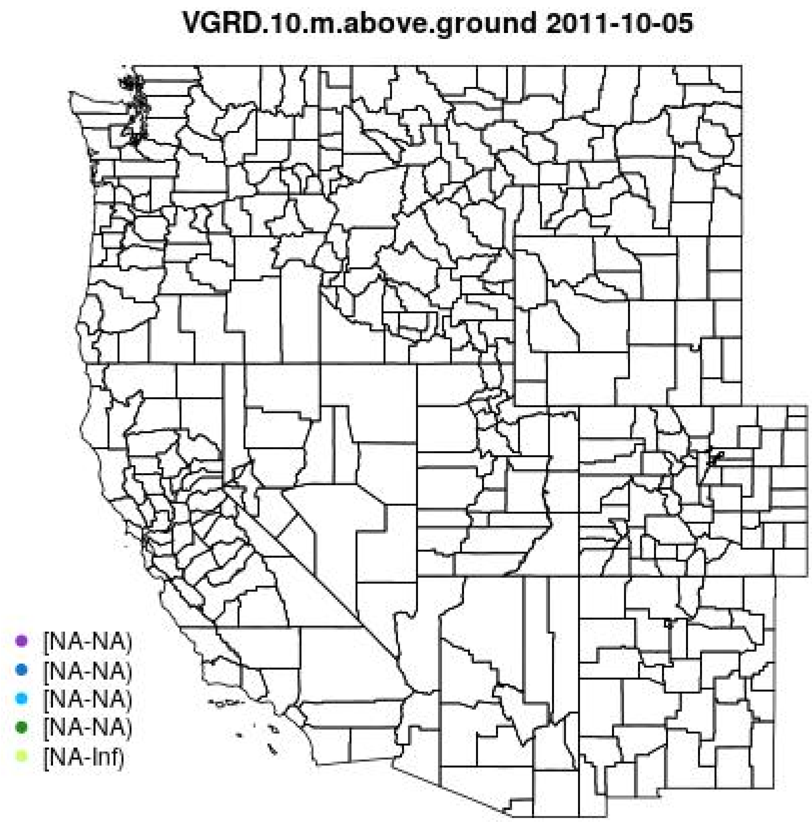

Figure 226: VGRD.10.m.above.ground 2011-10-05

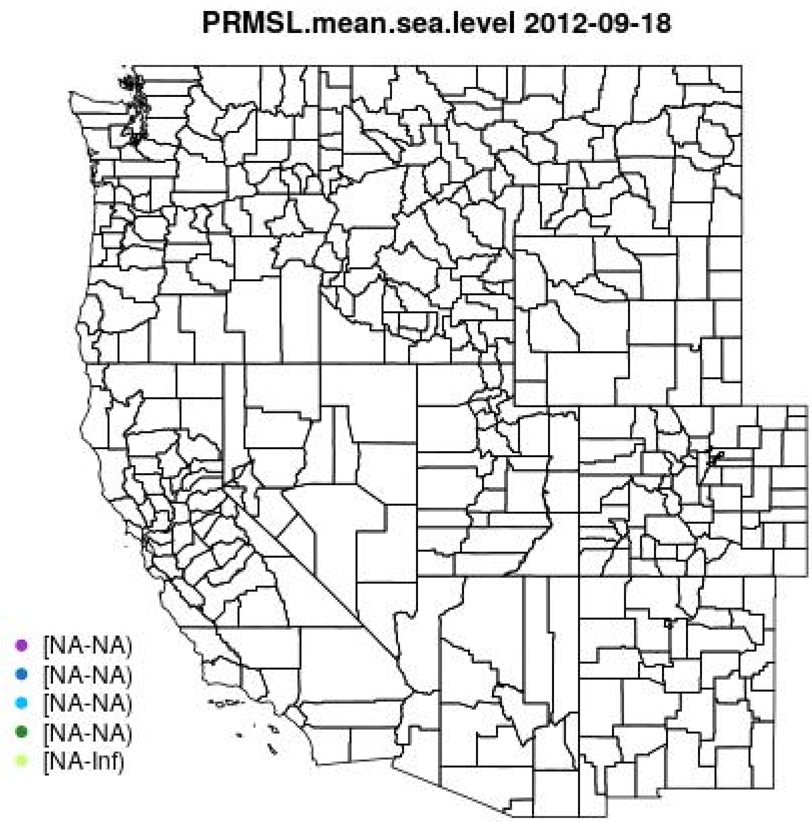

Figure 227: PRMSL.mean.sea.level 2012-09-18

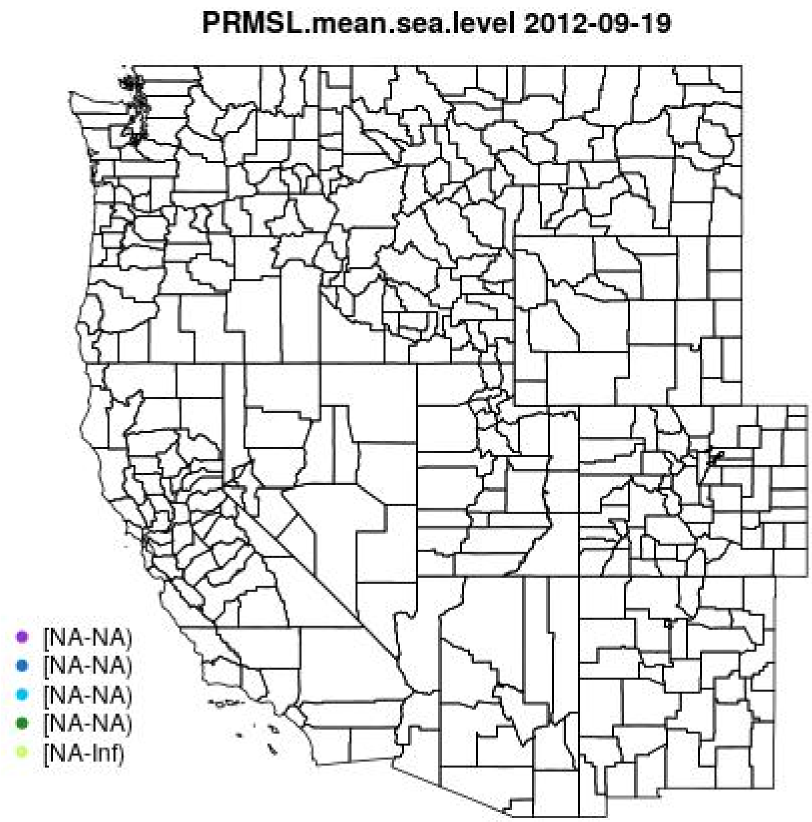

Figure 228: PRMSL.mean.sea.level 2012-09-19

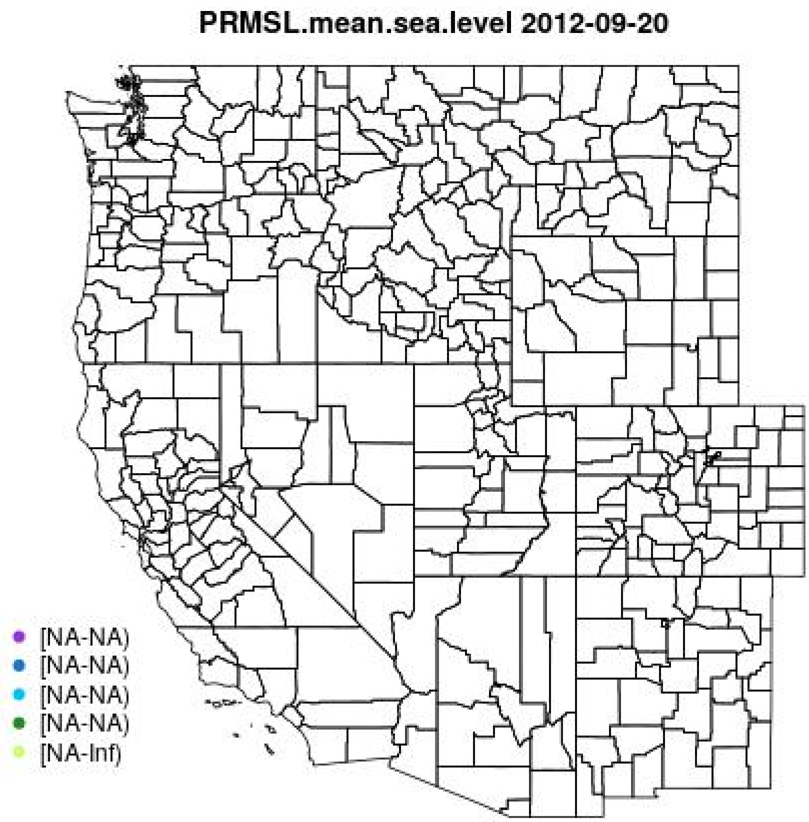

Figure 229: PRMSL.mean.sea.level 2012-09-20

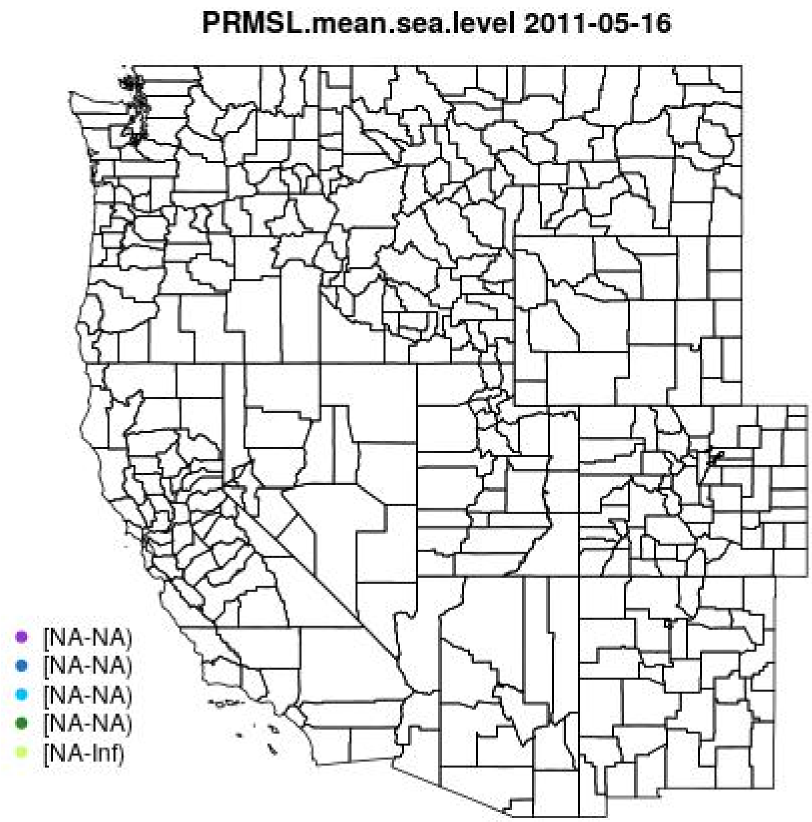

Figure 230: PRMSL.mean.sea.level 2011-05-16

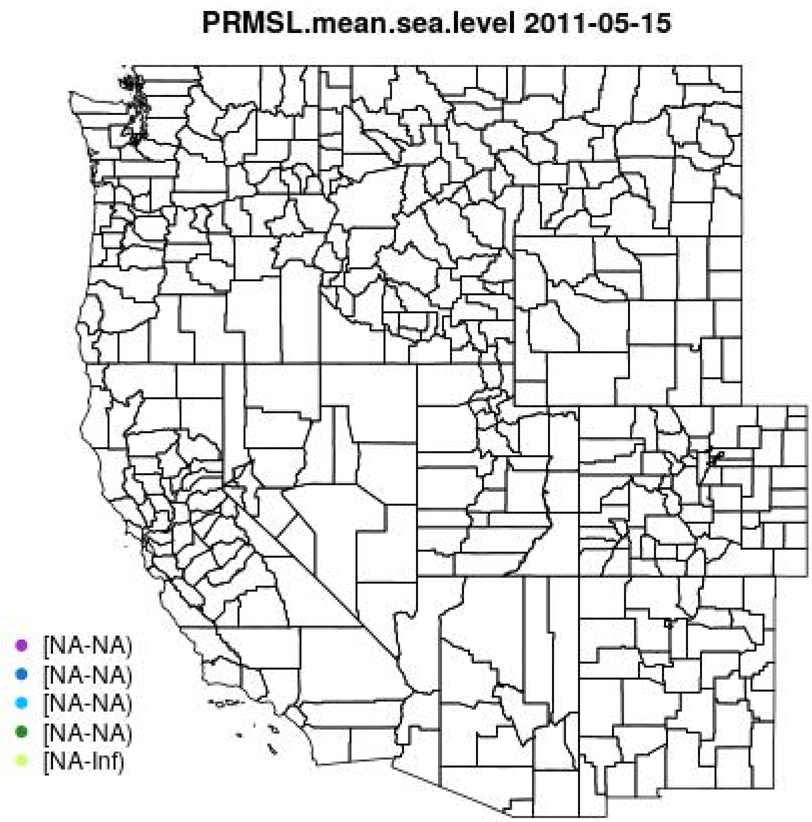

Figure 231: PRMSL.mean.sea.level 2011-05-15

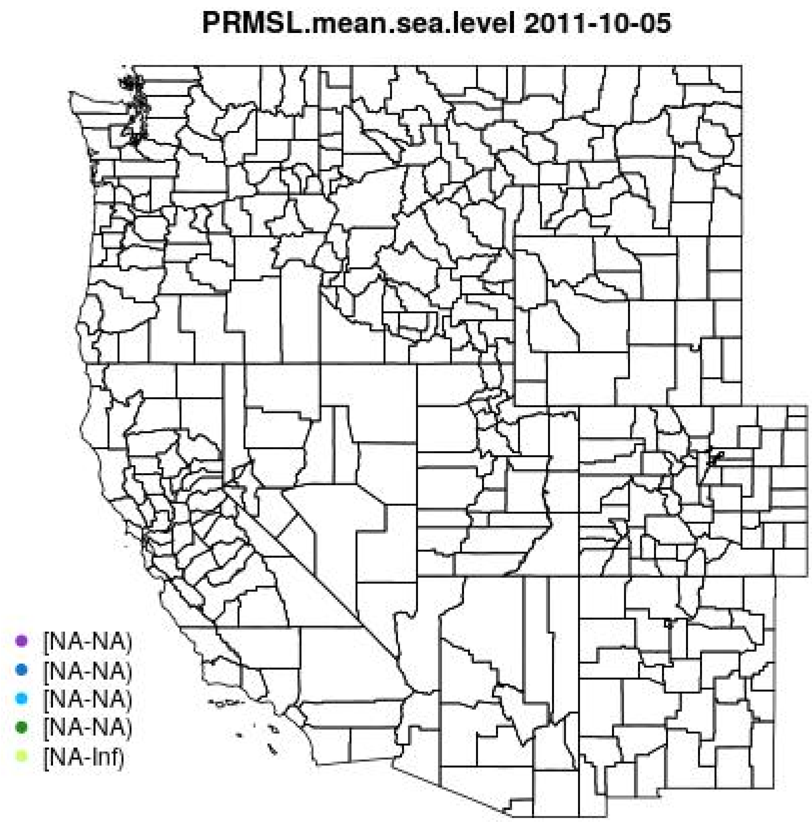

Figure 232: PRMSL.mean.sea.level 2011-10-05

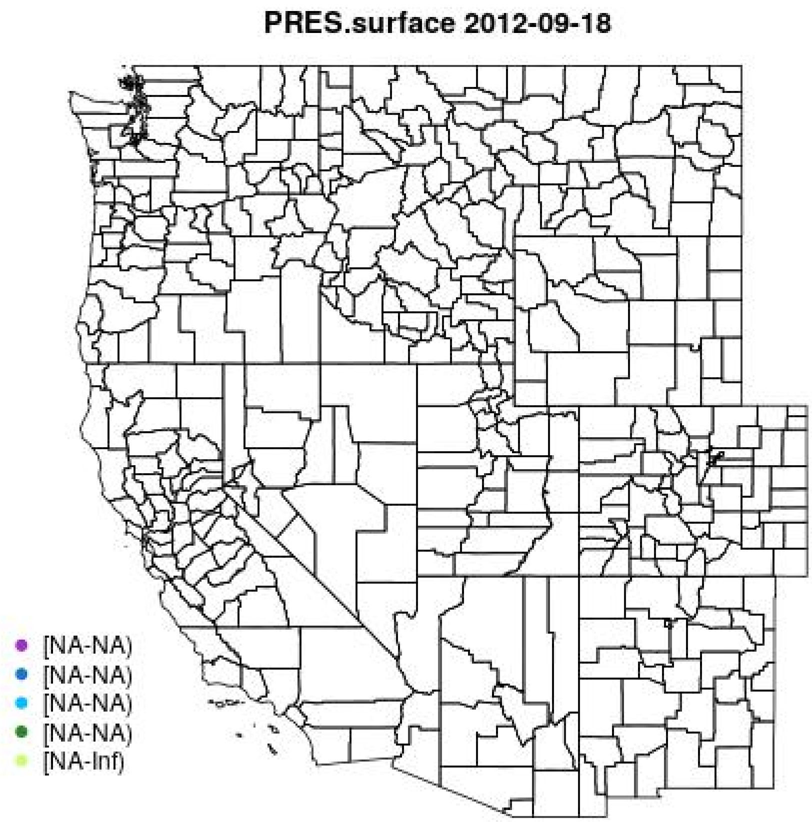

Figure 233: PRES.surface 2012-09-18

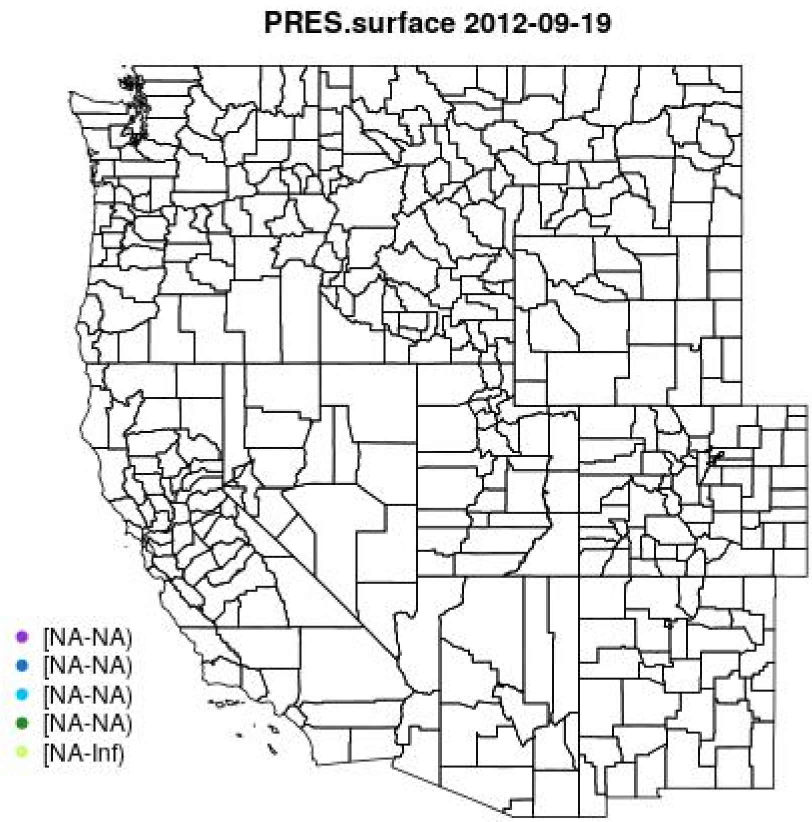

Figure 234: PRES.surface 2012-09-19

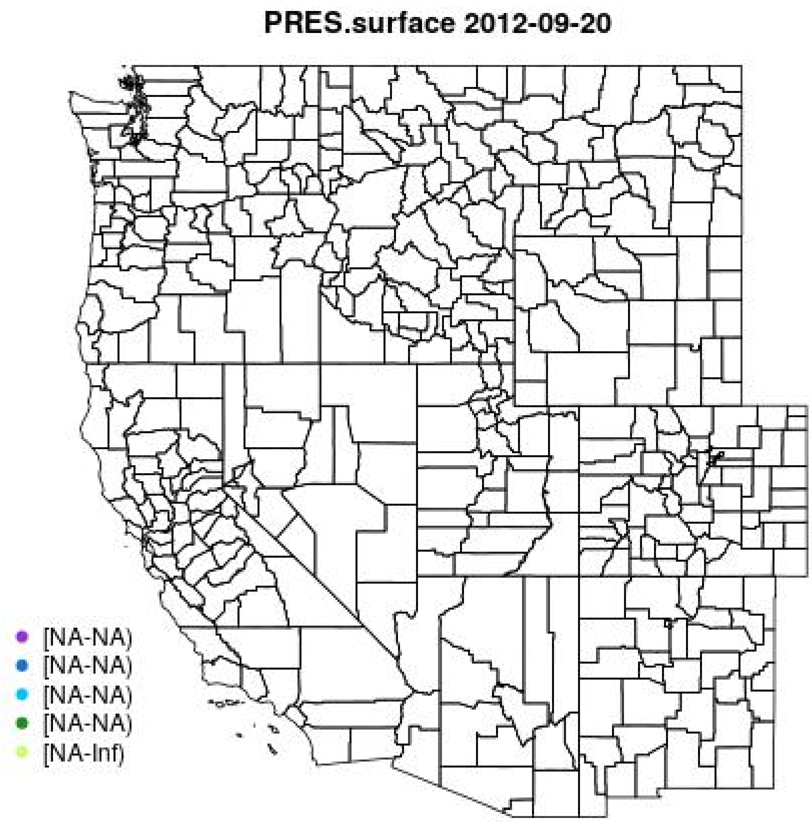

Figure 235: PRES.surface 2012-09-20

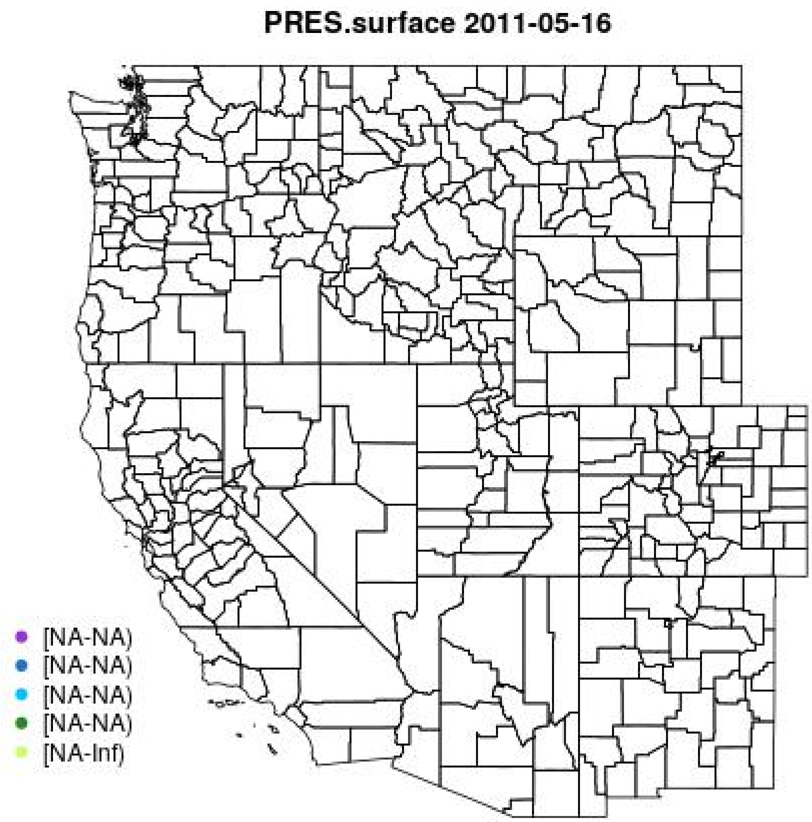

Figure 236: PRES.surface 2011-05-16

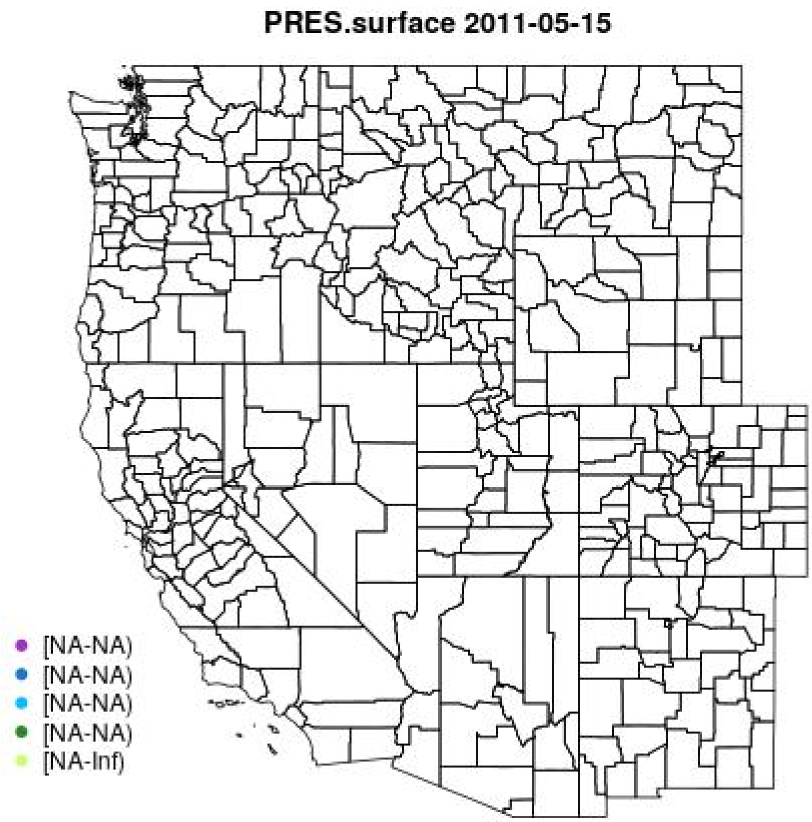

Figure 237: PRES.surface 2011-05-15

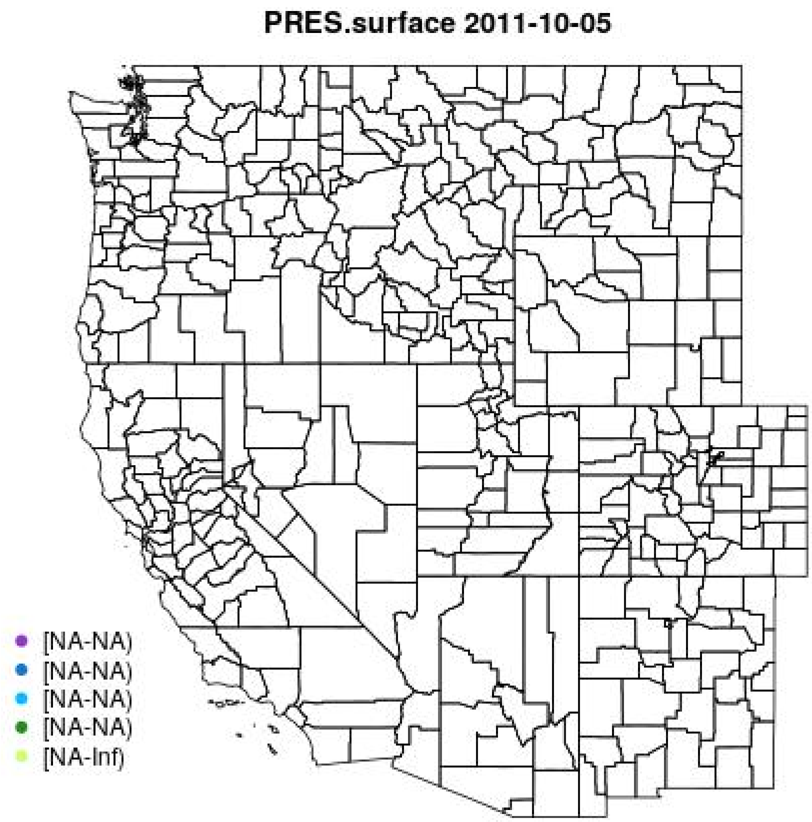

Figure 238: PRES.surface 2011-10-05

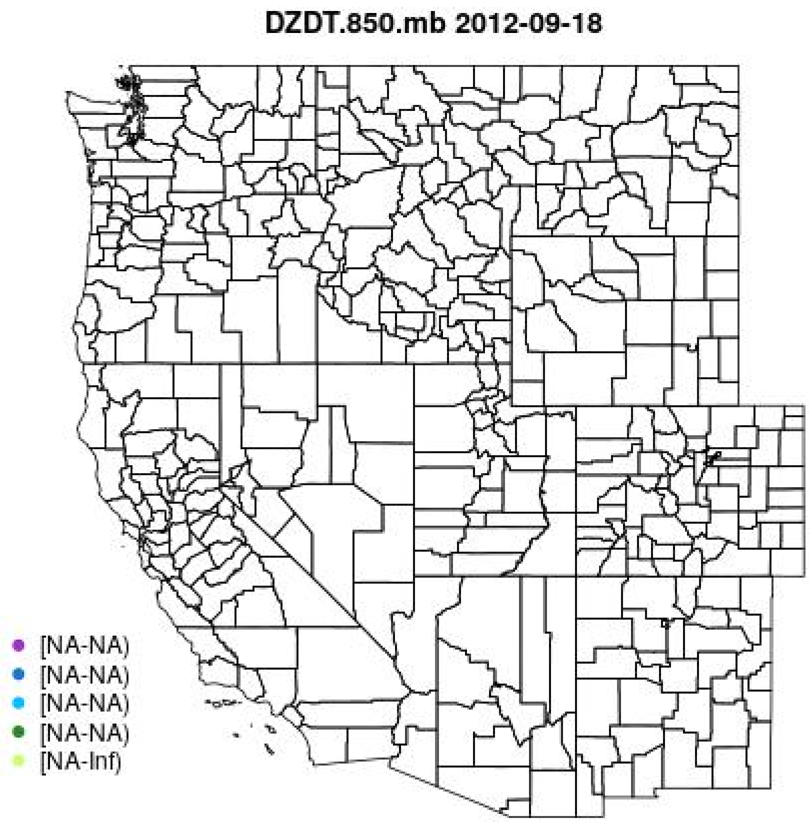

Figure 239: DZDT.850.mb 2012-09-18

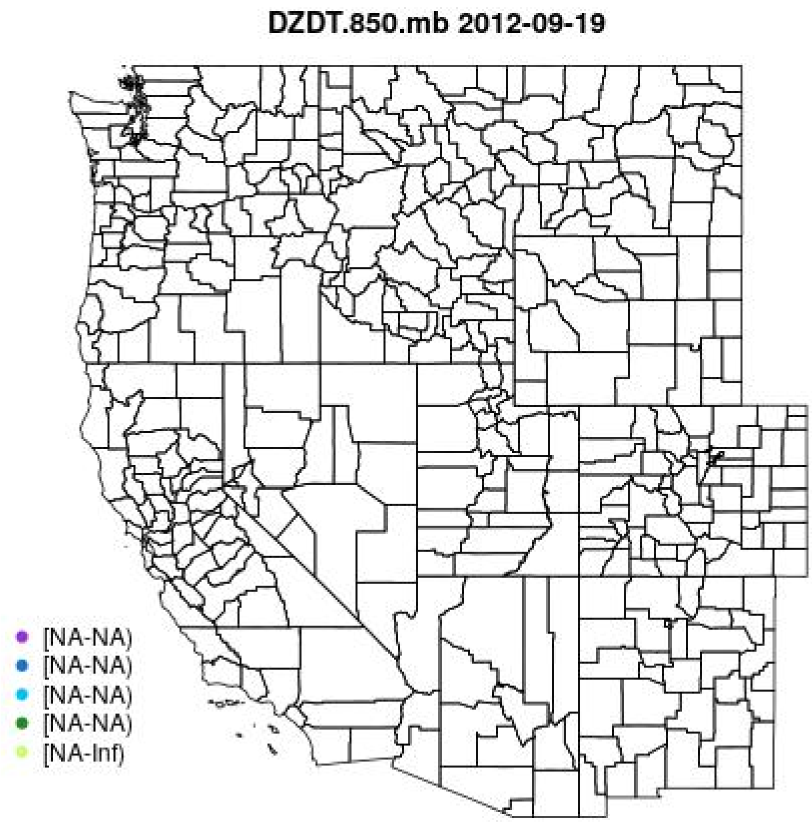

Figure 240: DZDT.850.mb 2012-09-19

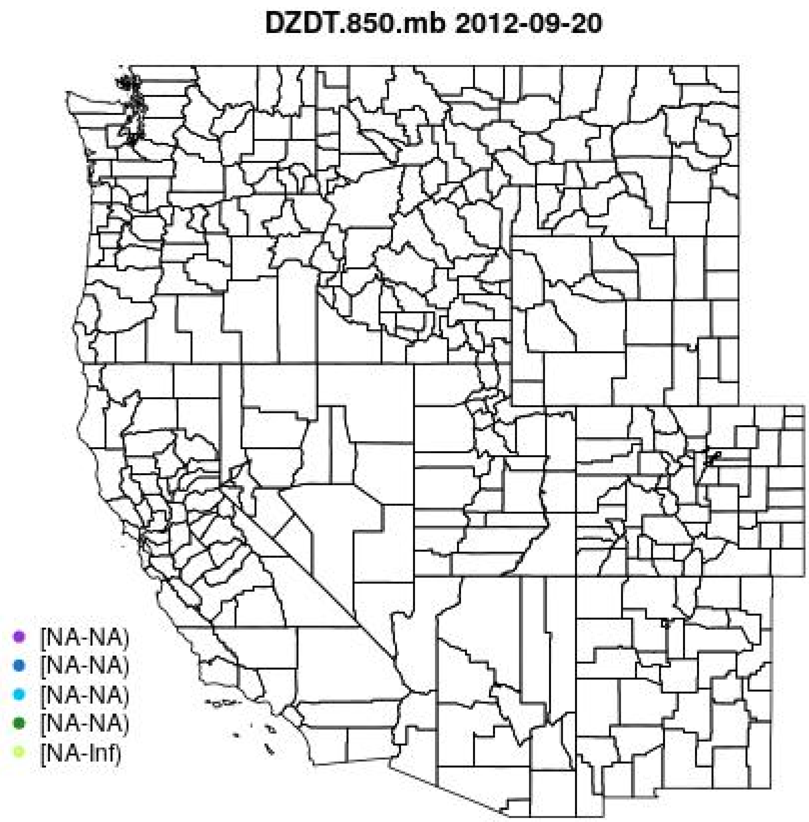

Figure 241: DZDT.850.mb 2012-09-20

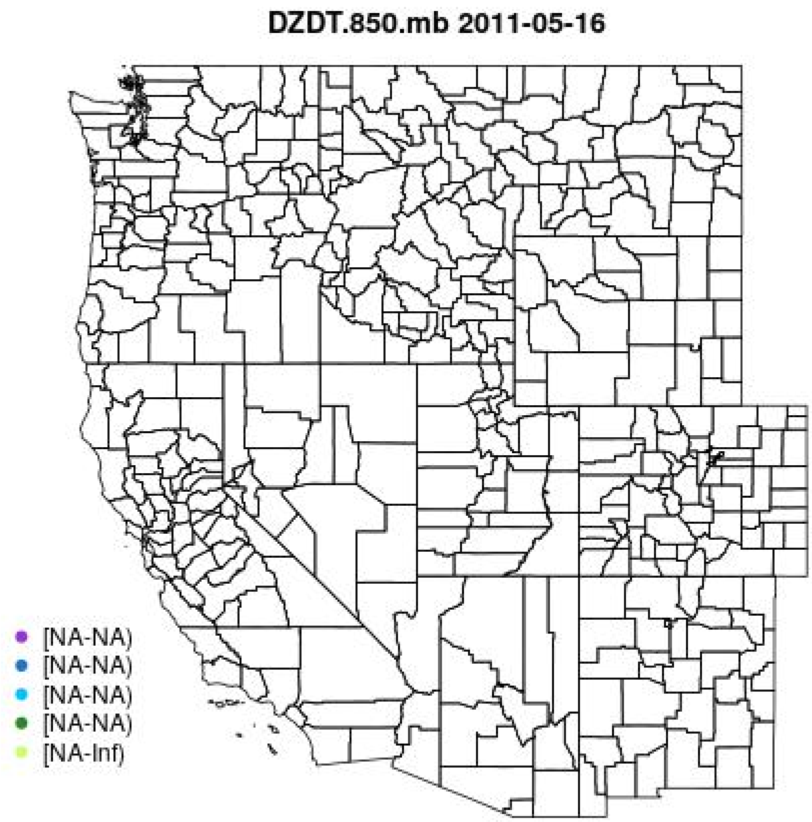

Figure 242: DZDT.850.mb 2011-05-16

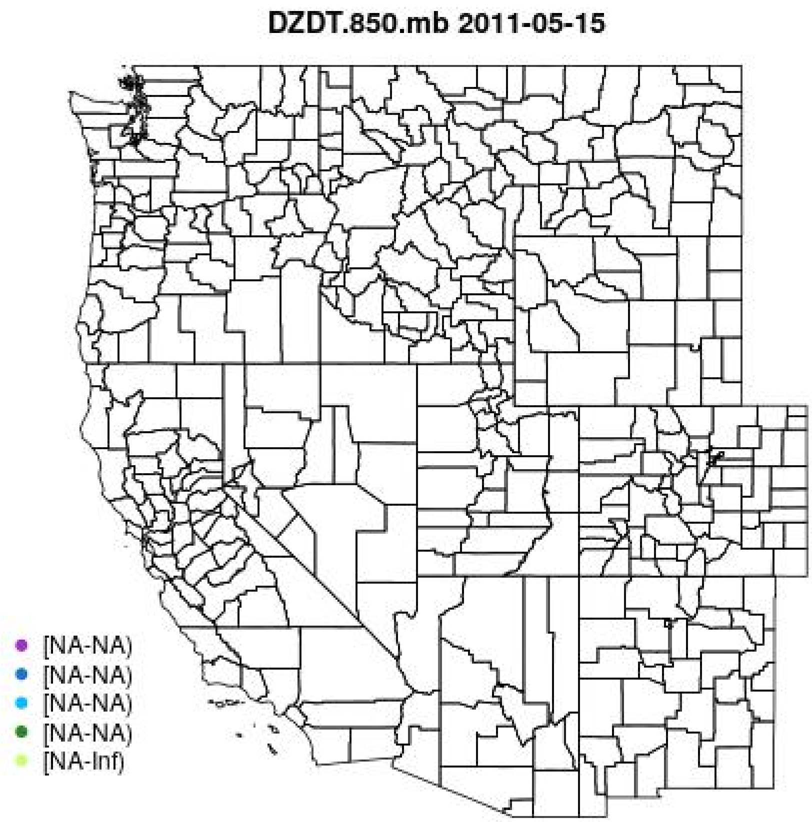

Figure 243: DZDT.850.mb 2011-05-15

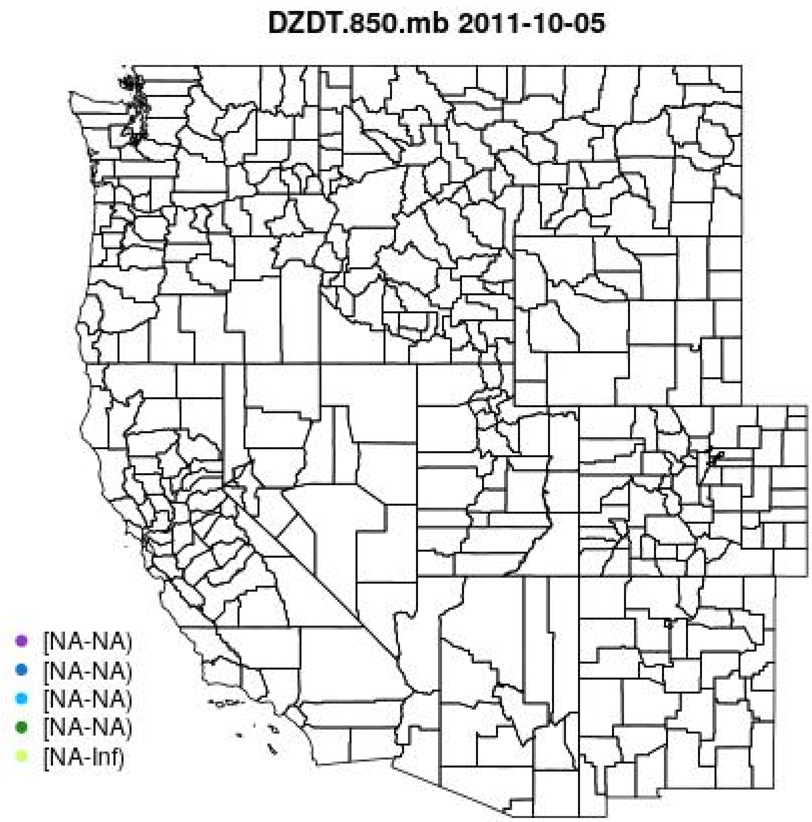

Figure 244: DZDT.850.mb 2011-10-05

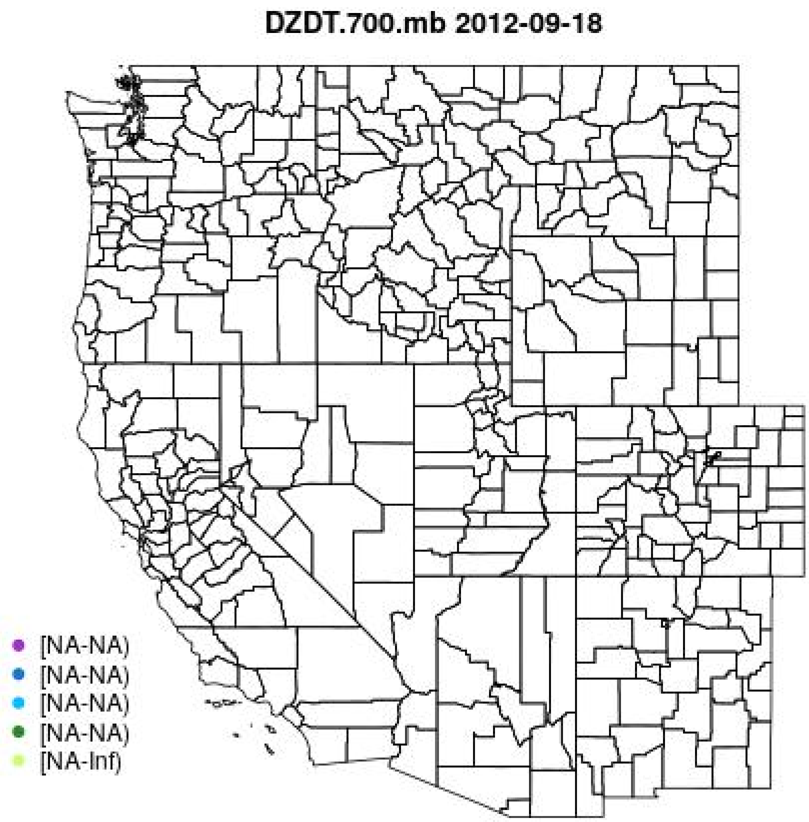

Figure 245: DZDT.700.mb 2012-09-18

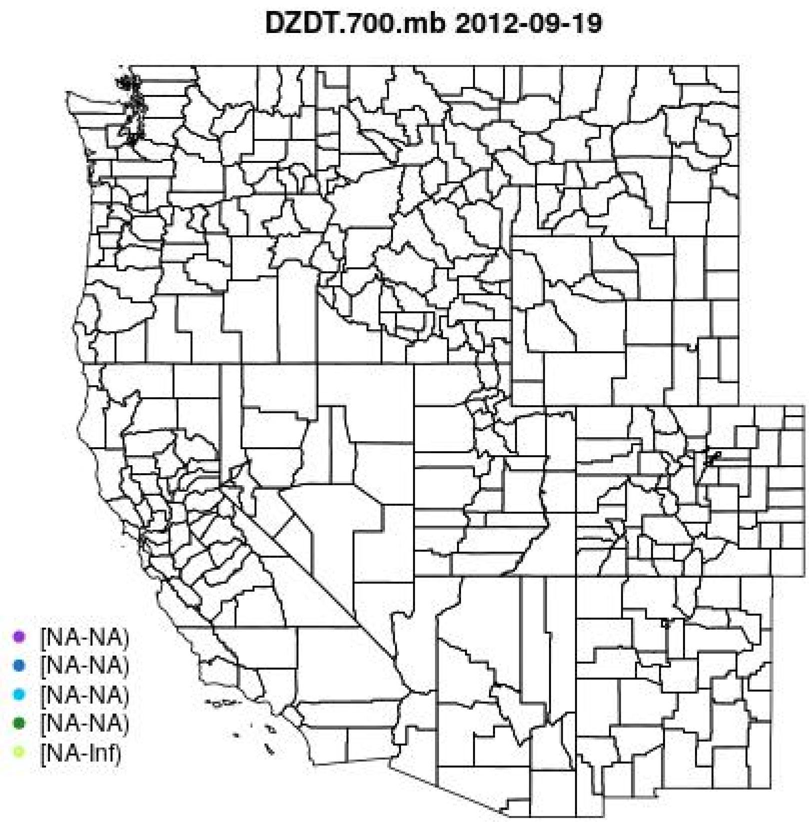

Figure 246: DZDT.700.mb 2012-09-19

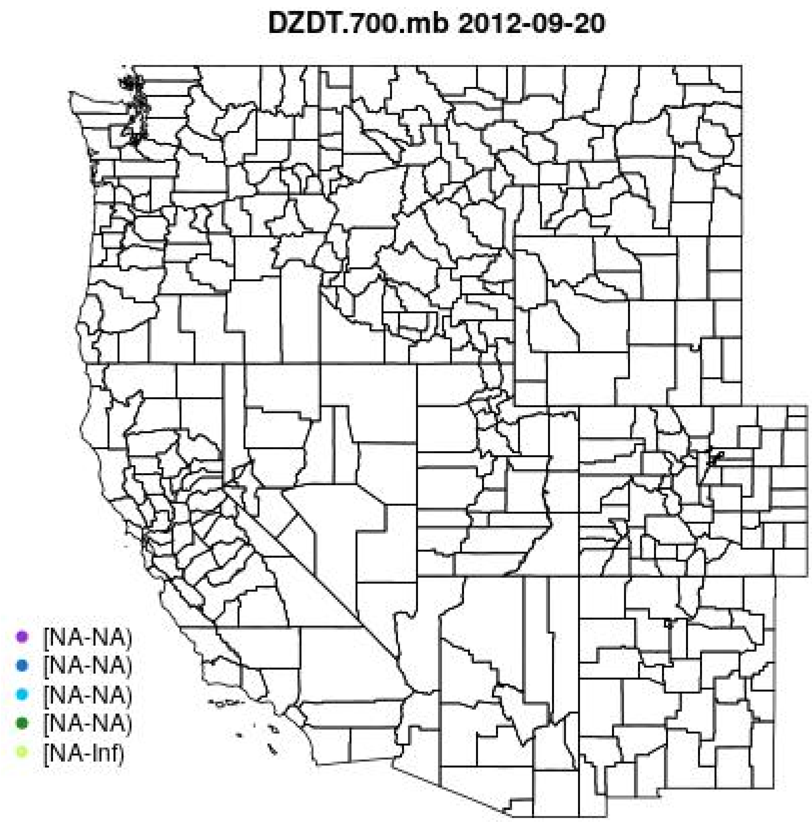

Figure 247: DZDT.700.mb 2012-09-20

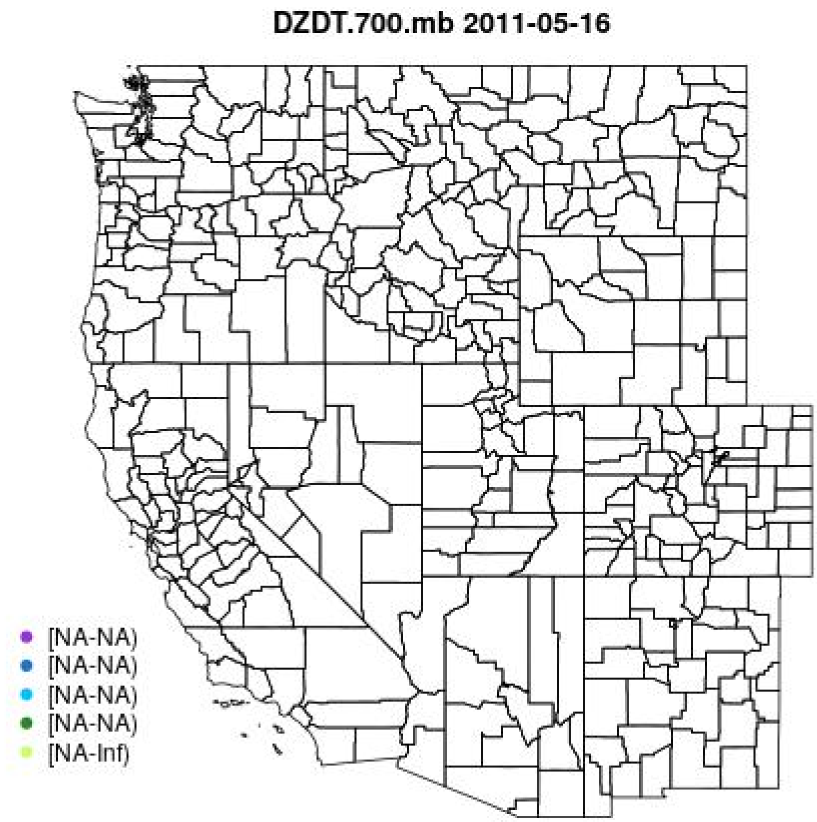

Figure 248: DZDT.700.mb 2011-05-16

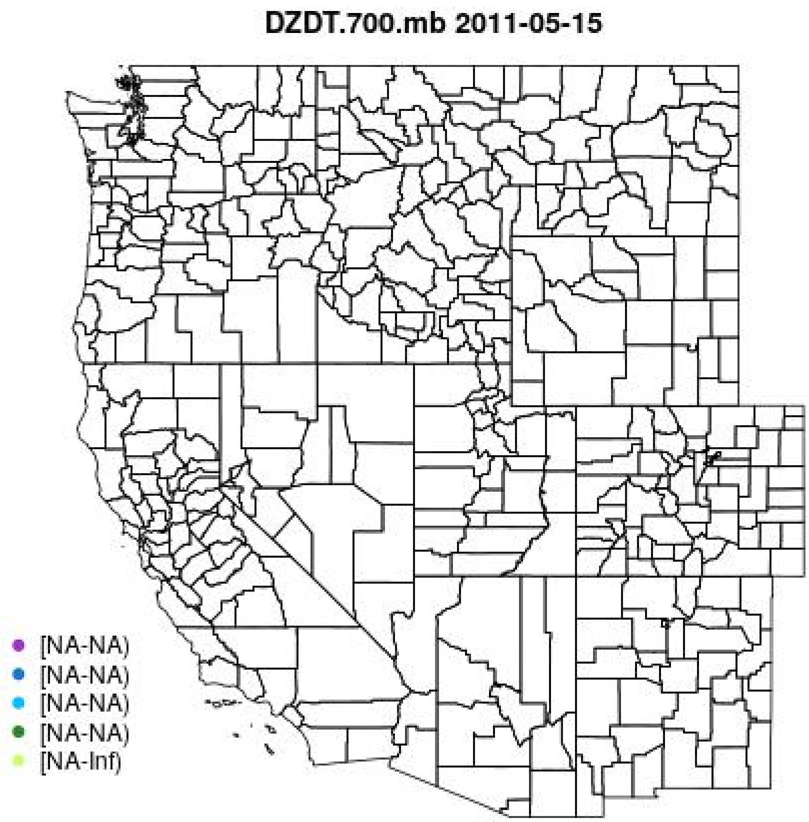

Figure 249: DZDT.700.mb 2011-05-15

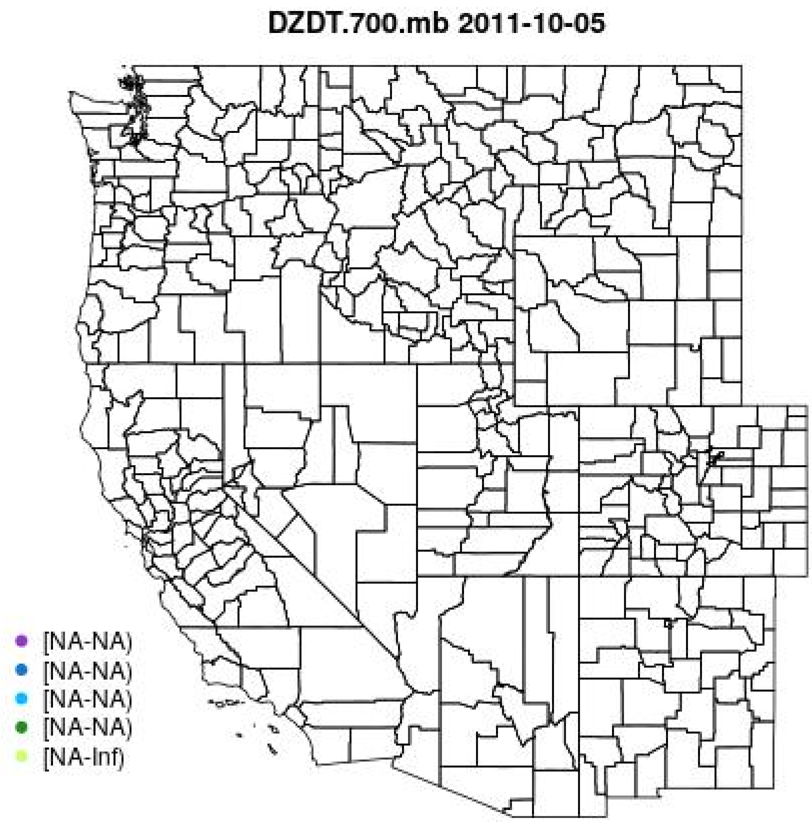

Figure 250: DZDT.700.mb 2011-10-05

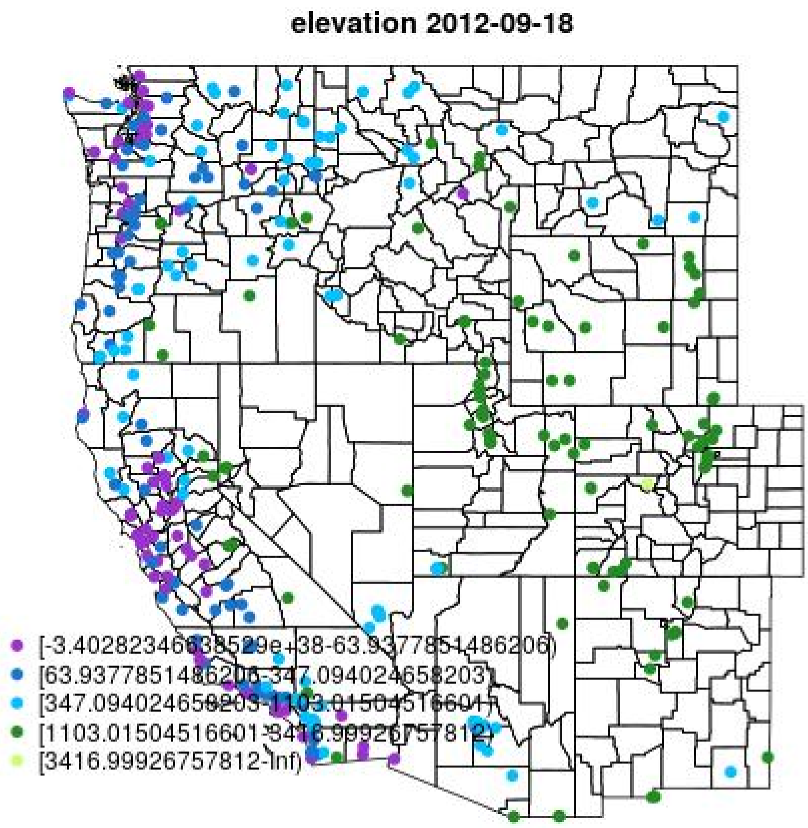

Figure 251: elevation 2012-09-18

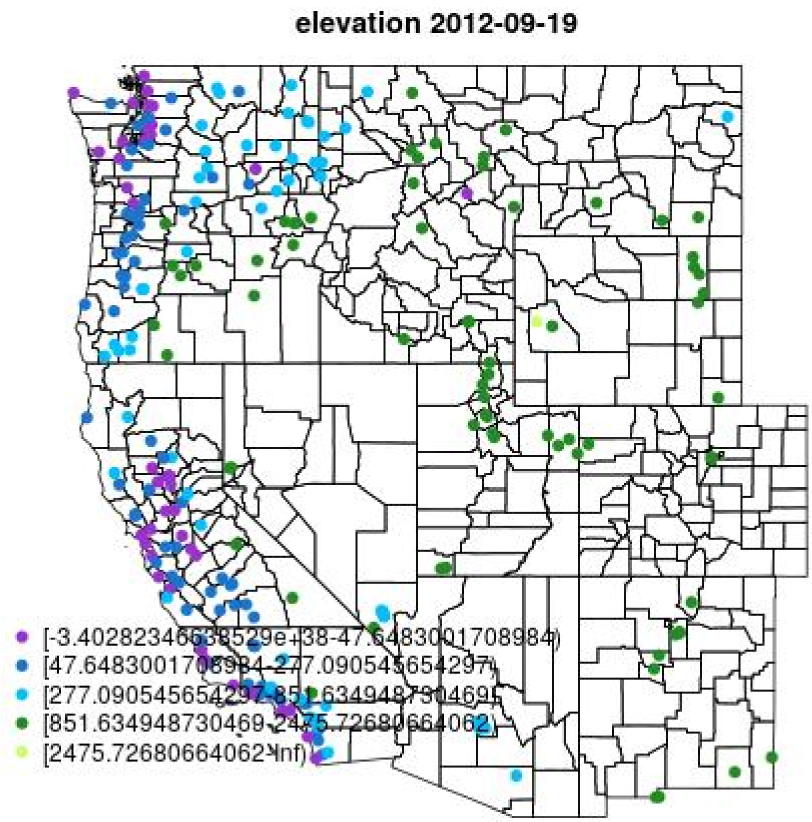

Figure 252: elevation 2012-09-19

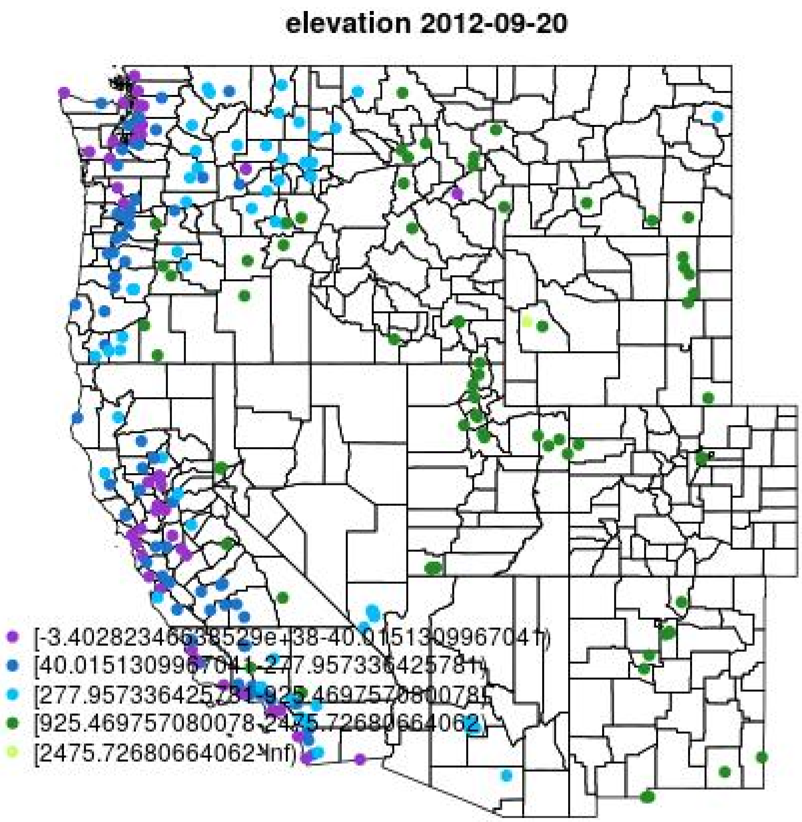

Figure 253: elevation 2012-09-20

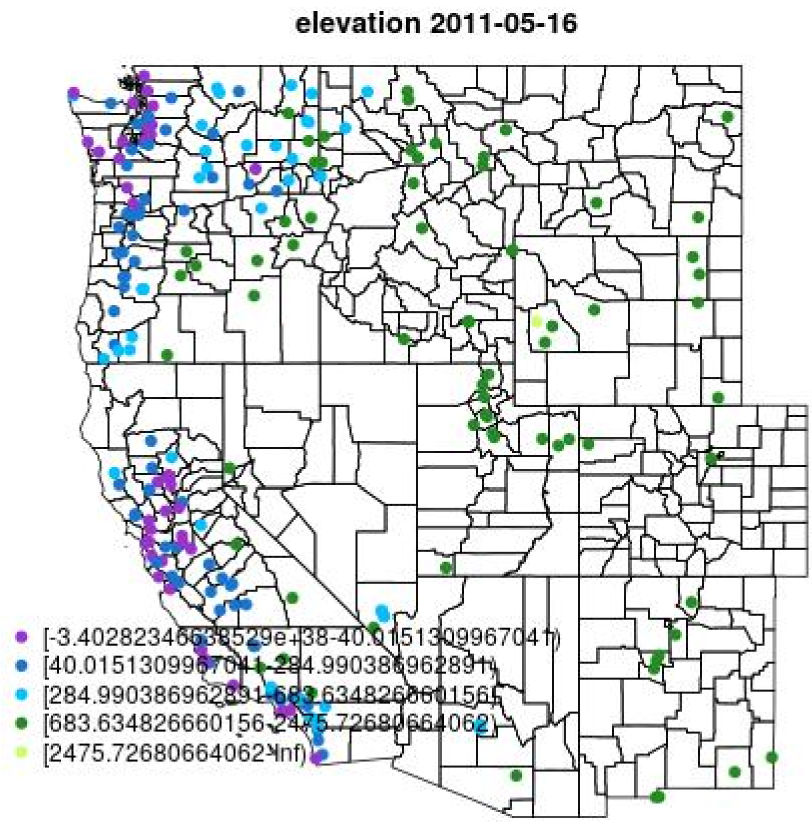

Figure 254: elevation 2011-05-16

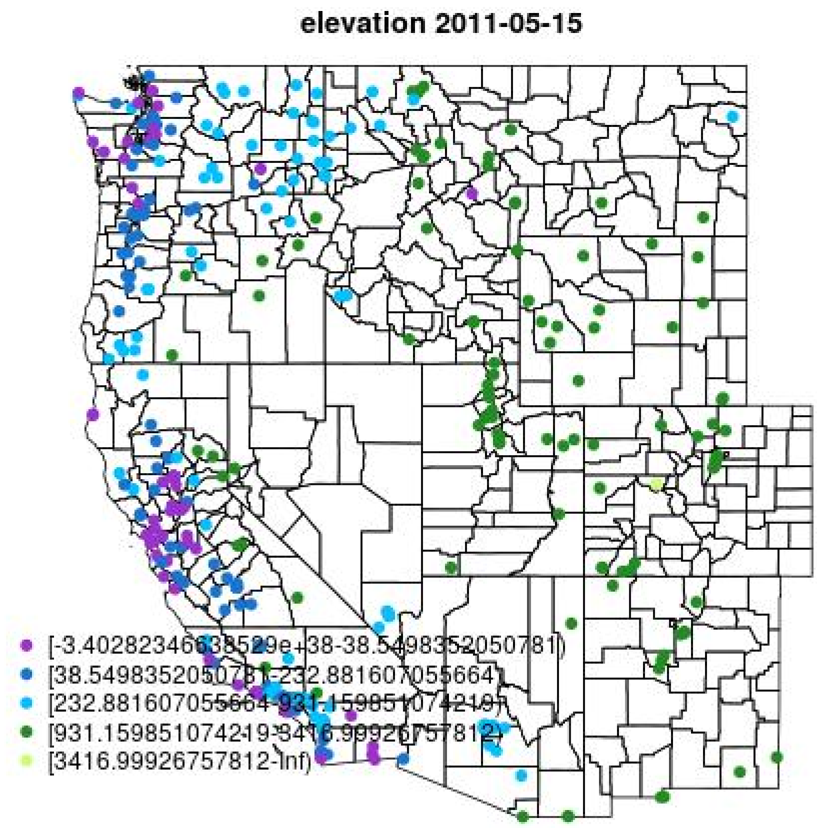

Figure 255: elevation 2011-05-15

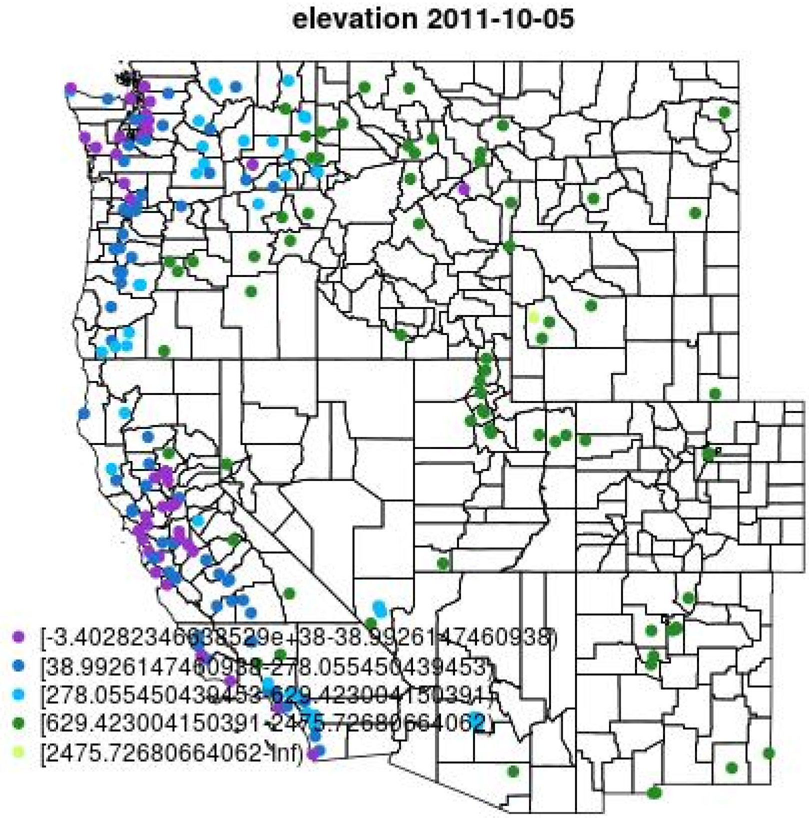

Figure 256: elevation 2011-10-05

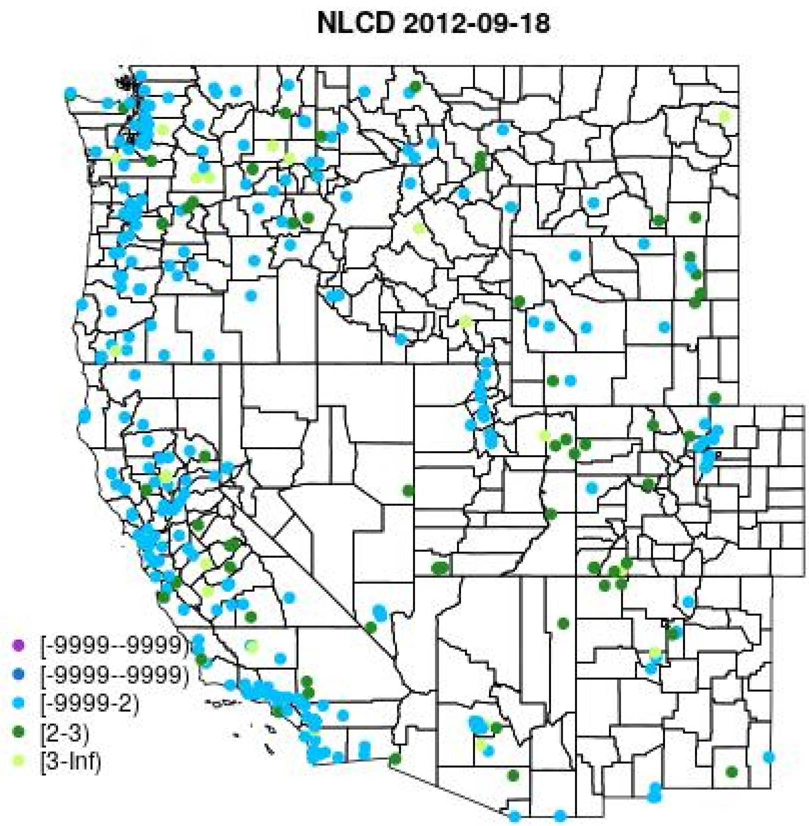

Figure 257: NLCD 2012-09-18

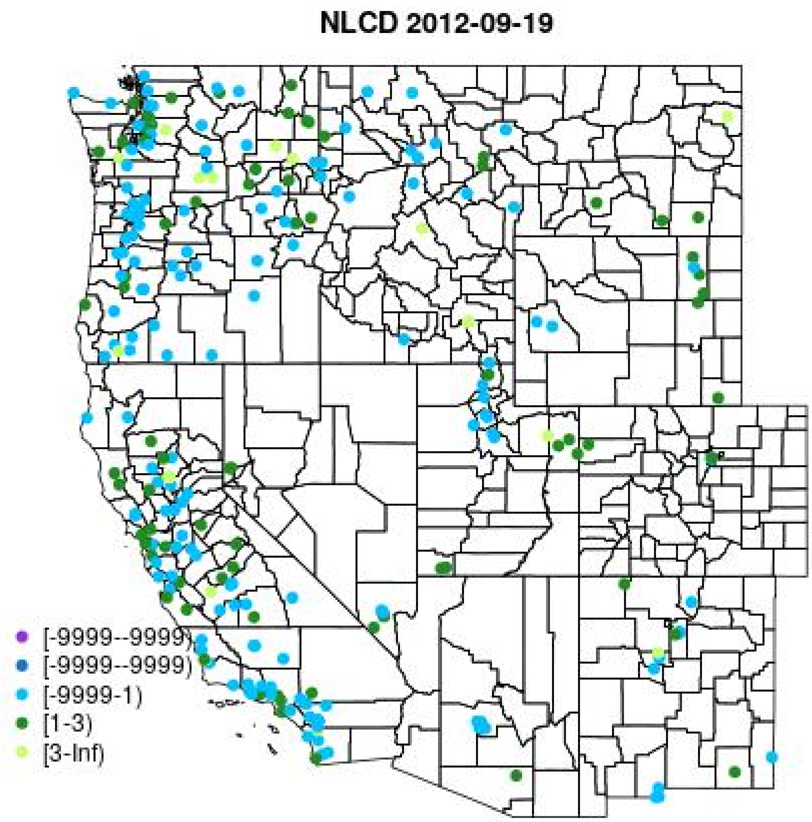

Figure 258: NLCD 2012-09-19

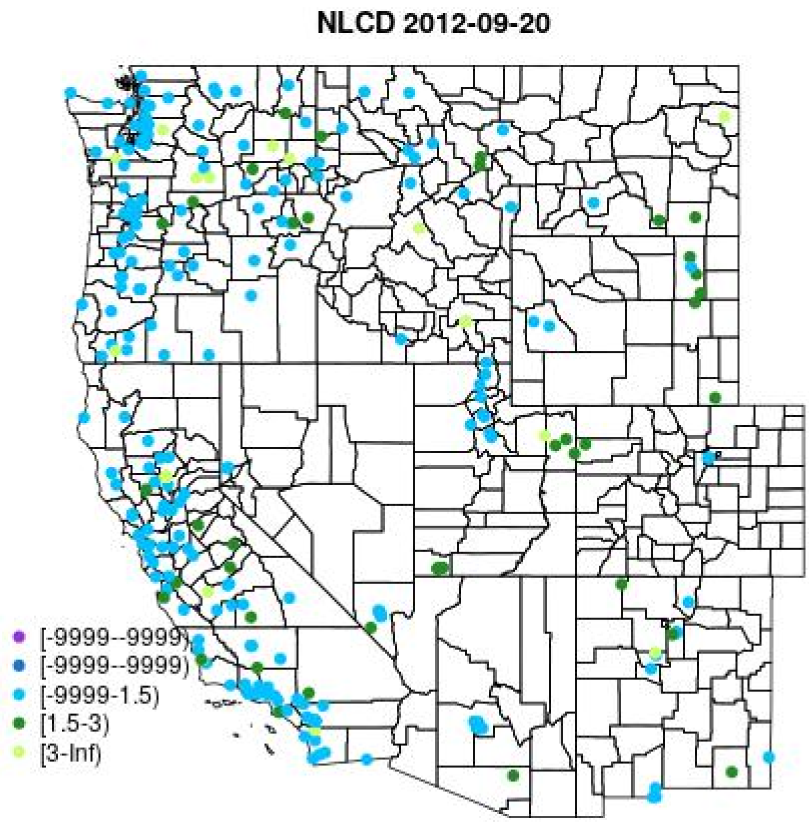

Figure 259: NLCD 2012-09-20

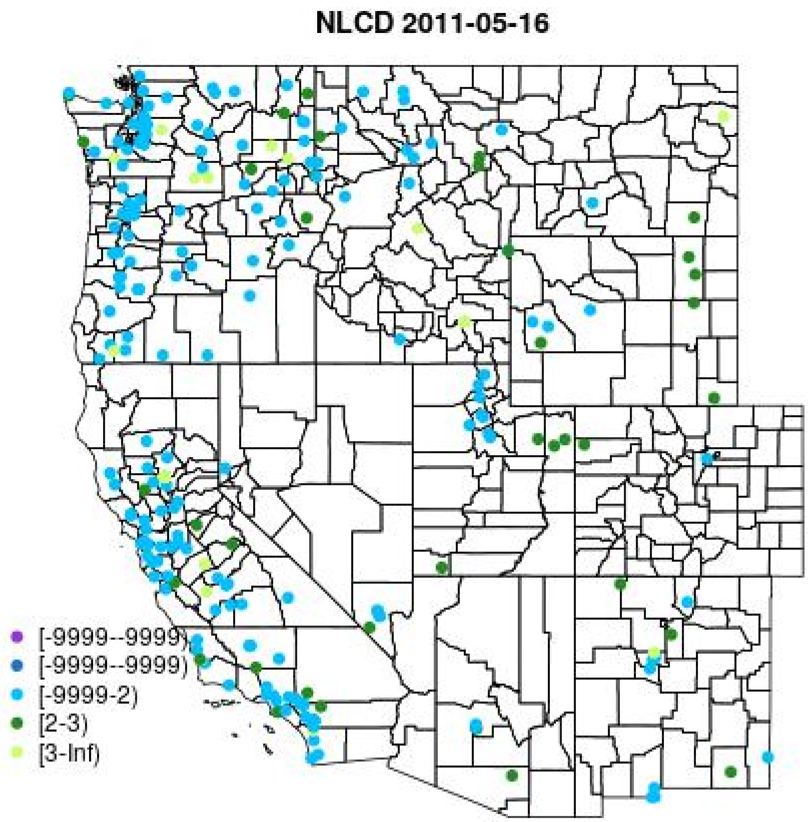

Figure 260: NLCD 2011-05-16

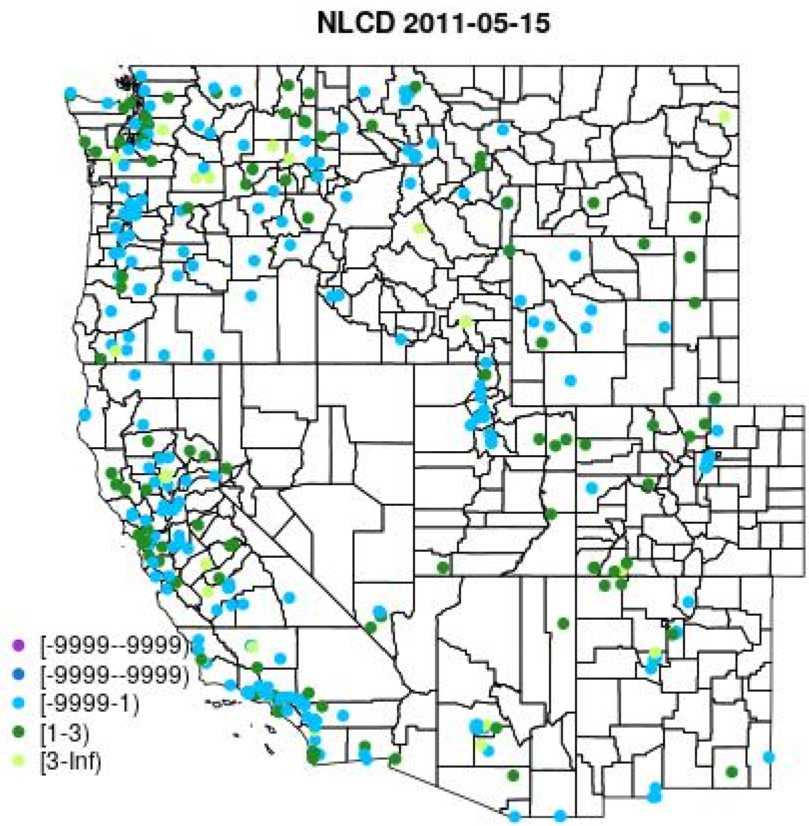

Figure 261: NLCD 2011-05-15

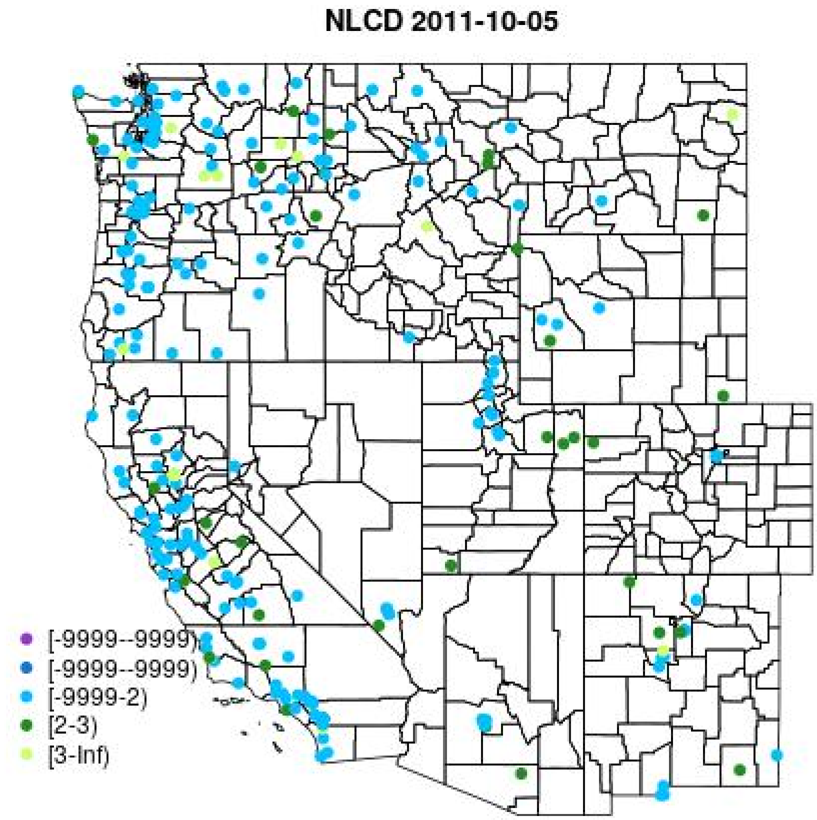

Figure 262: NLCD 2011-10-05

10.5 ML Inputs Map monthly medians Images

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