Model Evaluation Tools (MET) tutorial

4 - 5 February, 2009



• What?

- -Set of verification tools for evaluating forecasts via
 - standard statistics
 - object-based methods
 - scale decompositions



• Why?

- -Make verifiying easy.
- Encourage verification.
- Promote consistency across users.



• How?

- -A (unix like) package of software tools and scripts.
- Community contributed methods, graphics, etc.

Downloading MET

- MET distributed as a tarball to be downloaded and compiled locally.
 - METv1.1 released on July 11, 2008.
 - METv2.0 scheduled for February 2009.
 - METv2.0 BETA pre-installed on tutorial machines.
 - Register and download: www.dtcenter.org/met/users
- Language:
 - Written primarily in C++ with calls to a Fortran library
- Supported Platforms and Compilers:
 - Linux with GNU compilers
 - Linux with Portland Group (PGI) compilers
 - IBM machines with IBM compilers

www.dtcenter.org/met/users

DTC home Search RAL Model Evaluation Tools | DTC You are here: DTC . MET Users Page Model Evaluation Tools Home Joint Numerical Testbed Projects Terms of Use Developmental Testbed Center (DTC) Welcome Weather Research and Forecasting (WRF) Model Overview Welcome to the users page for the Model Evaluation Tools (MET) Support verification package. MET was developed by the National Center for Download Model Evaluation Tools (MET) Atmospheric Research (NCAR) Developmental Testbed Center (DTC) Documentation through the generous support of the U.S. Air Force Weather Agency Data Assimilation Testbed Center (DATC) (AFWA) and the National Oceanic and Atmospheric Administration User Support (NOAA). Joint Numerical Testbed Events Related Links Description WRF Tutorial (WRF, WRF-Var, MET) 01.26.2009 to 02.05.2009 Location: NCAR Foothills Laboratory, Boulder, CO MET is designed to be a highly-configurable, state-of-the-art suite of verification tools. It was developed using output from the Weather Research and Forecasting (WRF) modeling system but may be applied **MET Announcements** to the output of other modeling systems as well. Current release: METv1.1 (07.11.2008) Online Tutorial added for METv1.1. MET provides a variety of verification techniques, including: MET SPONSORS · Standard verification scores comparing gridded model data to U.S. Air Force Weather Agency (AFWA) point-based observations Standard verification scores comparing gridded model data to gridded observations · Object-based verification method comparing gridded model data to gridded observations National Oceanic and Atmospheric Administration (NOAA) ©2009, UCAR | Privacy Policy | Terms of Use Postal Address: P.O. Box 3000, Boulder, CO 80307-3000 • Shipping Address: 1850 Table Mesa Drive, Boulder, CO 80305 • Contact

Dependencies

REQUIRED:

- GNU Make Utility
- C++ and Fortran compilers (GNU, PGI, or IBM)
- NetCDF version 3 Library
- BUFRLIB Library
- GNU Scientific Library (GSL)
- F2C Library (f2c or g2c, for some compilers)

RECOMMENDED:

- WRF Post-Processor
- COPYGB (included with WRF-Post)
- R statistics and graphics package
- CWORDSH (not needed for v2.0)

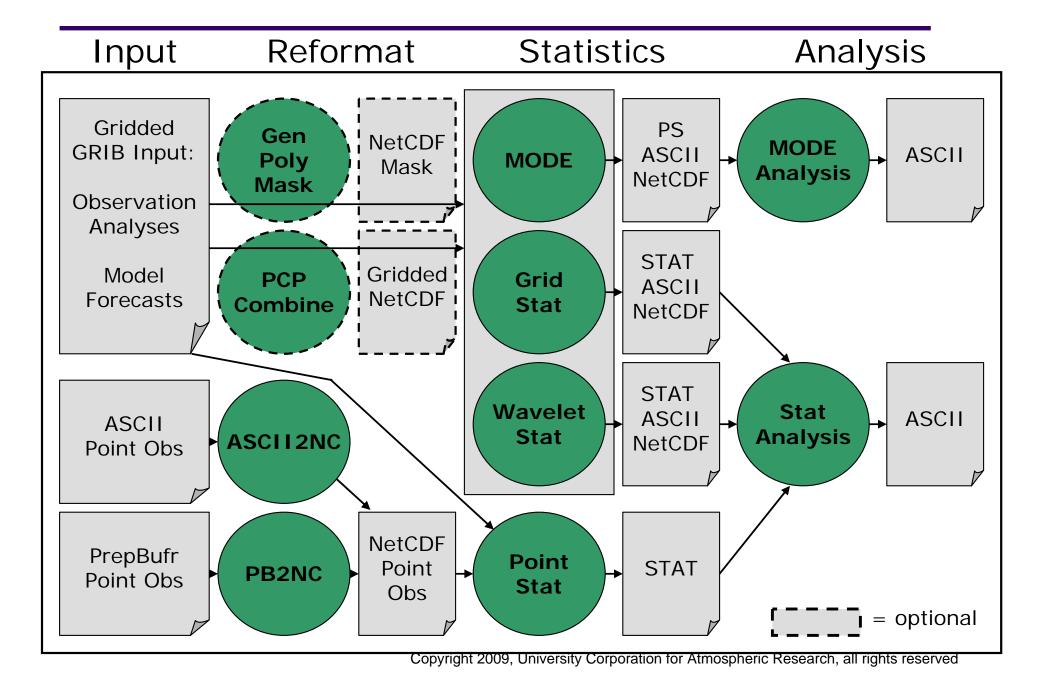
Directory Structure

File or Directory	Contents
README	Installation instructions and release notes.
Makefile_gnu (pgi, ibm)	Top-level Makefile to be configured prior to building MET.
lib/	Source code for internal MET libraries.
src/	Source code for the MET applications.
doc/	MET User's Guide.
bin/	Built MET executables.
scripts/	Test scripts to be run after building MET.
data/	Sample data used by the test scripts.
out/	Output generated by the test scripts.

Building MET

- Steps for building MET:
 - Build the required libraries with the same family of compilers to be used with MET.
 - 2. Select the appropriate Makefile.
 - GNU, PGI, or IBM
 - 3. Configure the Makefile.
 - C++ and Fortran compilers
 - Paths for NetCDF, BUFRLIB, GSL, and F2C libraries
 - 4. Execute the GNU Make utility to build all of the MET utilities.
 - 5. Run the test script and check for runtime errors.
 - Runs each of the MET tools at least once.
 - Uses sample data distributed with the tarball.

MET v2.0 Flowchart



Graphics

- Limited graphics incorporated into MET.
- Several options for plotting MET statistical output - including R, NCL, IDL, GNUPlot, and many others.
- Users are encouraged to submit their own plotting and/or analysis scripts using any tool for posting to the MET website.

R Statistics and Graphics

- The R Project for Statistical Computing (www.r-project.org)
 - Powerful statistical analysis and plotting tools.
 - Freely available and well supported for Linux/Windows/Mac.
- Sample R plotting and analysis scripts posted on the MET website.
- Recent praise for R in the technology section of the NY Times:
 - http://www.nytimes.com/2009/01/07/technology/ business-computing/07program.html?_r=2

Configuration Files

- Set of command line tools controlled using command line options and ASCII configuration files.
 - Well commented and documented in MET User's Guide.
 - Easy to modify.
 - Distributed with the tarball.
- Configuration files control things such as:
 - Fields/levels to be verified.
 - Thresholds to be applied.
 - Interpolation methods to be used.
 - Verification methods to be applied.
 - Regions over which to accumulate statistics.

Data Formats and Types

Tressa L. Fowler

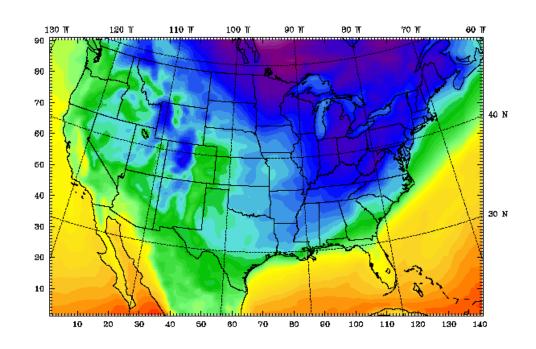
What you can do with MET depends on what type of data you have.

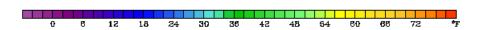
The **format** (grid, point) of your data determines your MET tool(s).

The **type** (continuous, binary) of your data determines the analyses to use within each tool.

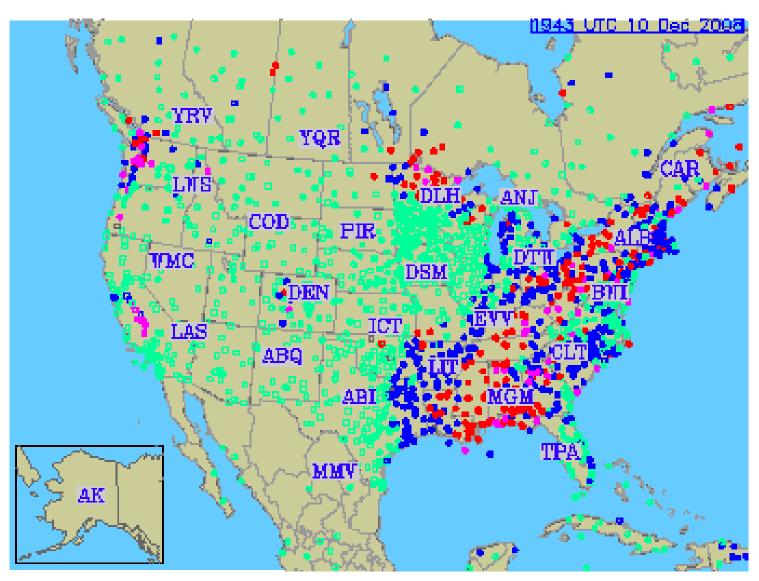
Data Formats

Gridded Forecasts (2D or 3D)





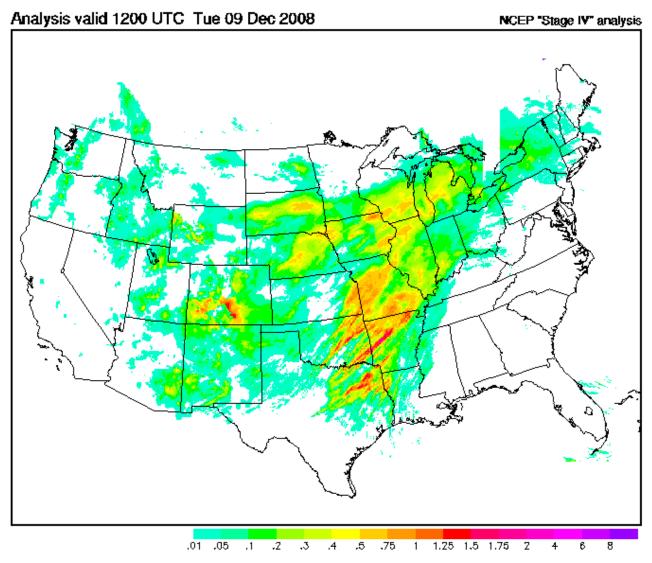
Point Observations (2D or 3D)



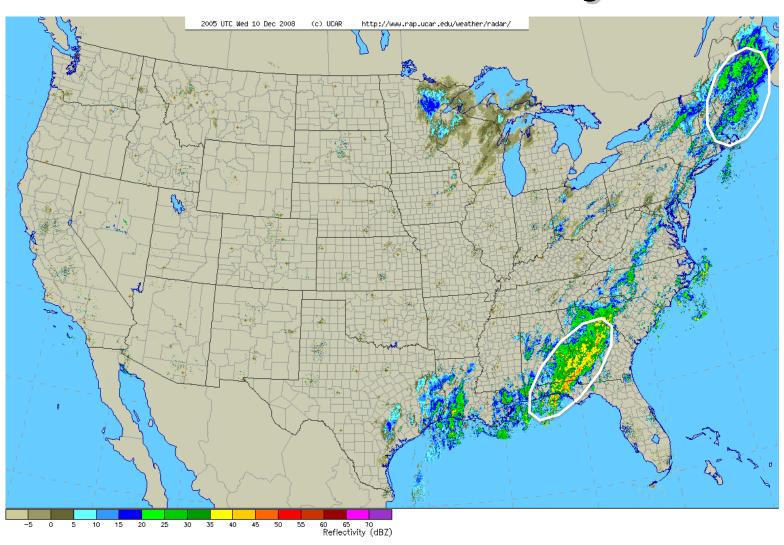
copyright 2009, UCAR, all rights reserved.

Gridded Observations (2D or 3D)

Past 24-hour accumulated precip. (water equiv inches)



Gridded data to transform into **Objects**



Data	MET Tool
Gridded Forecasts Gridded Observations	Grid stat Wavelet Stat MODE
Gridded Forecasts Point Observations	Point Stat

Data Types

Types of Forecasts

- Continuous
 - Wind speed
 - Temperature
- Categorical (includes Binary)
 - Rain / No Rain
 - Hurricane Category 1 5
- Probabilistic
 - Prob of freezing precip
- Ensembles

Types of Observations

- Continuous
 - Wind speed
 - Temperature
- Categorical (includes binary)
 - Rain / No Rain
 - No tornado, F1, F2, F3, F4, F5

Data type	Analyses
Continuous forecasts Continuous observations	Root Mean Squared Error (RMSE), Mean Absolute Error (MAE), Bias
Continuous forecasts, Categorical observations	Receiver Operating Characteristic (ROC) curve, Kolmogorov-Smirnov
Categorical forecasts, Categorical observations	Contingency table statistics and skill scores
Probabilistic forecasts, Categorical observations	Brier score, ranked probability score (RPS), reliability diagram

Introduction to Standard Verification

Tressa L. Fowler

Basics

 Match up forecasts and observations at points.

 Calculate differences, sums, or counts over all the points.

Summarize these things as statistics.

Matching Points to Grids

- Observation points are unlikely to fall exactly on forecast grid points.
- Match in horizontal space via choice of methods:
 - Closest
 - Interpolate
 - Function of surrounding points,
 e.g.
 - Min of closest 4
 - Median of closest 25
- Match in vertical by interpolating between level above and below.

X

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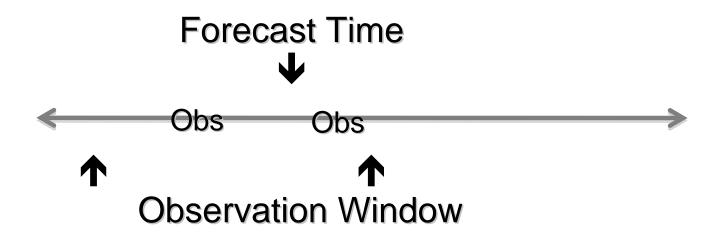
Matching Grids to Grids

 Must use some converter to put forecasts and observations on the same grid.

– Example: copygb

Time

 If your forecasts and observations are not at the same time, you may need to define a time window for your observations.



Now you have a bunch of forecast / observation pairs

How well do they match?

F	Ο	F	0
0.32	0.03	1	1
0. 51	0.0	0	0
0.42	0.48	0	1
0.08	0.14	1	0
0.20	0.23	0	1
0.75	0.33	1	1





From those pairs, you can calculate a variety of statistics . . . like these

http://www.bom.gov.au/bmrc/wefor/staff/eee/verif/verif_web_page.html

$$POD = \frac{hits}{hits + misses}$$

$$RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (F_i - O_i)^2}$$

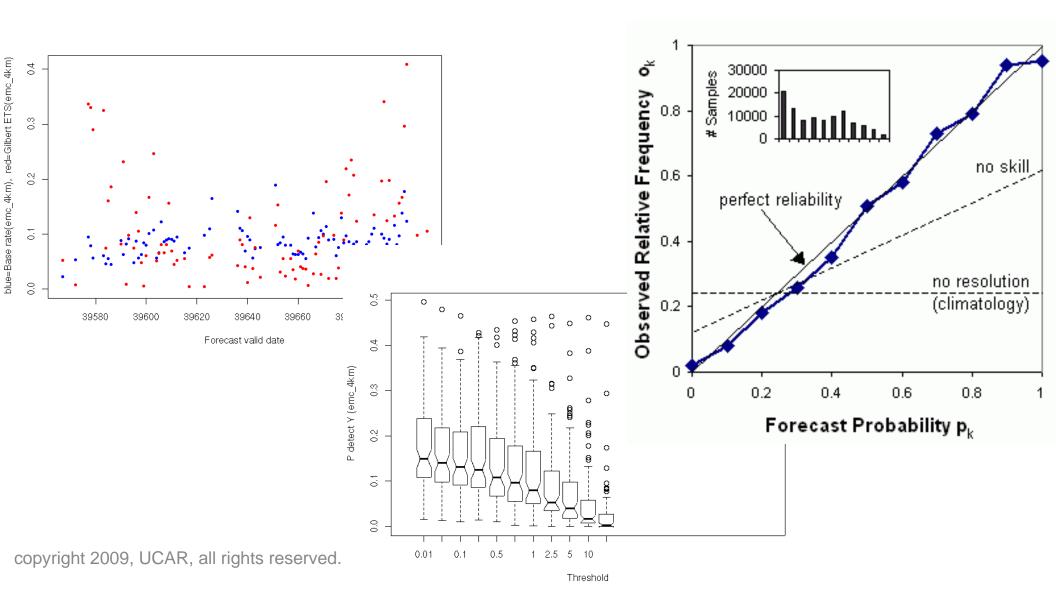
$$ETS = \frac{hits - hits_{random}}{hits + misses + false \ alarms - hits_{random}}$$

Correlation r

$$r = \frac{\sum (F - \overline{F})(\bigcirc - \overline{\bigcirc})}{\sqrt{\sum (F - \overline{F})^2} \sqrt{\sum (\bigcirc - \overline{\bigcirc})^2}}$$

	Observed Event	Observed Non-event
Forecast Event	Count = 532 (Hits)	Count = 219 (False Alarm)
Forecast Non-event	Count = 393 (Miss)	Count =1,627 (Correct No's)

Example Plots



References for information about verification

- http://www.bom.gov.au/bmrc/wefor/staff/eee/verif/verif_w eb_page.html
- Wilks, D.S., 2006. <u>Statistical Methods in the Atmospheric Sciences.</u> 2nd Ed, Academic Press.
- Jolliffe, Stephenson (eds.), 2003. <u>Forecast Verification: A</u> <u>Practitioner's Guide in Atmospheric Science.</u> Wiley.

File Formats and Pre-Processing

- Data Formats
- Pre-processing Tools
- Useful Links

Supported Data Formats

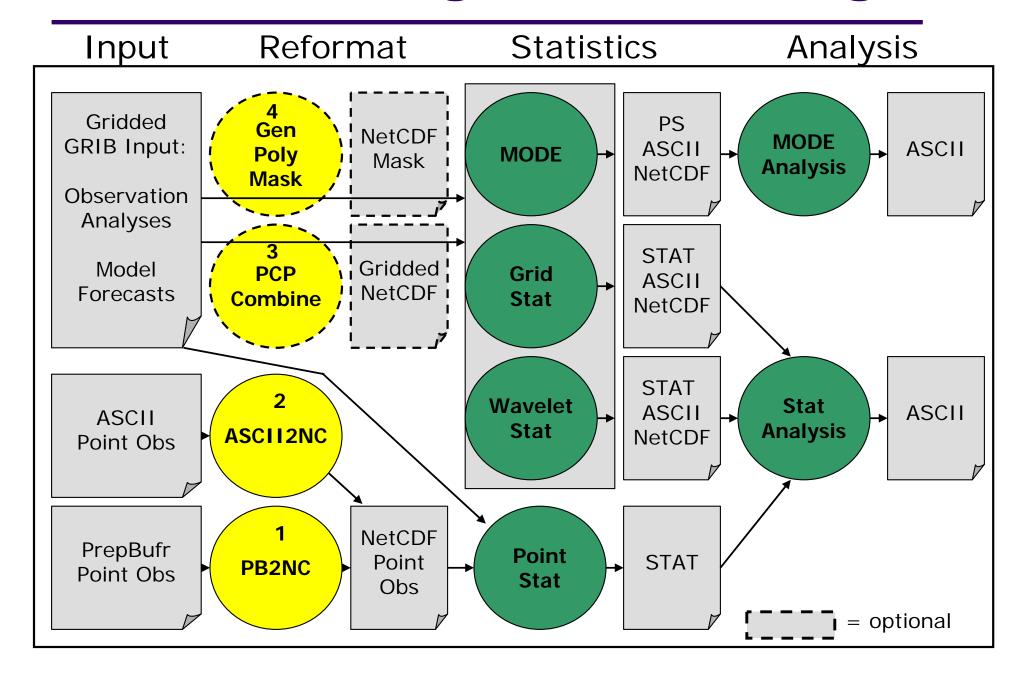
Forecasts

- GRIB GRIdded Binary data format (*version 1)
- NetCDF network Common Data Format

Observations

- PREPBUFR binary dataset prepared by NCEP from varied data sources.
- ASCII "Met_Point" format (10-cols x n-rows)
- gridded GRIB i.e. NEXRAD Level II or IV
- point NetCDF i.e. aircraft observations

Pre-Processing / Reformatting



Data Reformating Tools

PB2NC and ASCII2NC

 Arrange observational data into the NetCDF point format expected by Point-Stat.

PCP_Combine (optional)

- Sum precipitation values across two or more time periods.
- Subtract precipitation values to create values for finer subperiods.
- Produces gridded NetCDF file that can be used as input grid for any Statistics tool.

Gen_Poly_Mask (optional)

- Used when more complex masking is needed.
- Produces a NetCDF file of pre-defined mask.
- May be used for masking in any Statistics tools.

1. PB2NC Tool

Stands for "PREPBUFR to NetCDF"

Functionality:

- Filters and reformats PREPBUFR point observations into intermediate NetCDF format.
- Configuration file specifies:
 - Observation types, variables, locations, elevations, quality marks, and times to retain or derive for use in Point-Stat.

Data formats:

- Reads PREPBUFR using NCEP's BUFRLIB.
- Writes point NetCDF as input to Point-Stat.

Note: v2.0 no longer requires CWORDSH to pre-process PREPBUFR files.

PB2NC: Usage

Usage: pb2nc prepbufr_file netcdf file config_file [-pbfile prepbufr_file] [-valid_beg time] [-valid_end time] [-nmsg n] [-dump path] [-v level]

prepbufr_file	Input obs file in PrepBufr format
netcdf_file	Output name for NetCDF file
config_file	PB2NC configuration file
-pbfile	Additional input PrepBufr files
-valid_beg	Beginning of valid time window [YYYYMMDD_[HH[MMSS]]
-valid_end	End of valid time window [YYYYMMDD_[HH[MMSS]]
-nmsg	Number of PrepBufr messages to process
-dump	Dump entire contents o PrepBufr file to file in path
-V	Level of logging

PB2NC: Run

METv2.0/bin/pb2nc \
 ndas.t00z.prepbufr.tm12.20070401.nr \
 tutorial/out/pb2nc/tutorial_pb.nc \
 tutorial/config/PB2NCConfig_tutorial \
 -v 2

==> append : to filename to view the data source **BUFR** 230ADPUPA UPPER-AIR (RAOB, PIBAL, RECCO, DROPS) REPORTS 231AIRCAR MDCRS ACARS AIRCRAFT REPORTS 232AIRCFT AIREP/PIREP, AMDAR(ASDAR/ACARS), E-ADAS(AMDAR BUFR) ACF233SATWND SATELLITE-DERIVED WIND REPORTS 234PROFLR WIND PROFILER REPORTS 235VADWND VAD (NEXRAD) WIND 236SATEMP TOVS SATELLITE DATA (SOUNDINGS, **REPORTS** RETRIEVALS, RADIANCES) 237ADPSFC SURFACE LAND (SYNOPTIC, METAR) **REPORTS** 238SFCSHP SURFACE MARINE (SHIP, BUOY, C-MAN PLATFORM) REPORTS 239SFCBOG MEAN SEA-LEVEL PRESSURE BOGUS **REPORTS** 240SPSSMI SSM/I RETRIEVAL PRODUCTS (REPROCESSED WIND SPEED, TPW) 241SYNDAT SYNTHETIC TROPICAL CYCLONE BOGUS **REPORTS** 242ERS1DA ERS SCATTEROMETER DATA (REPROCESSED WIND SPEED) 243GOESND GOES SATELLITE DATA (SOUNDINGS, RETRIEVALS, RADIANCES) 244QKSWND QUIKSCAT SCATTEROMETER DATA (REPROCESSED WIND SPEED) 245MSONET MESONET SURFACE REPORTS (COOPERATIVE NETWORKS) 246GPSIPW GLOBAL POSITIONING SATELLITE-INTEGRATED PRECIP. WATER 247RASSDA RADIO ACOUSTIC SOUNDING SYSTEM (RASS) TEMP PROFILE RPTSM063000BYTCNT

Result of running linux "less" command:

>less \
ndas.t00z.prepbufr.tm12.2007
0401.nr

PREPBUFR

- BUFR is the World Meteorological Organization (WMO) standard binary code for the representation and exchange of observational data.
 - http://www.nco.ncep.noaa.gov/sib/decoders/BUFRLIB/
 - http://www.ecmwf.int/products/data/software/
- The PREPBUFR format is produced by NCEP for analyses and data assimilation. The system that produces this format:
 - Assembles observations dumped from a number of sources
 - Encodes
 - information about the observational error for each data type
 - background (first guess) interpolated to each data location
 - Performs both rudimentary multi-platform quality control and more complex platform-specific quality control
- MET currently only supports PREPBUFR. Let us know if you need to read in other BUFR format.

PREPBUFR - Known Issues

- CWORDSH blocking
 - In v1.0 and v1.1 it was needed to structure to data properly for reading by the MET fortran code.
 - Need eliminated in v2.0.
- Compilation 64-bit OS
 - Will work if you compile using 32-bit flags.
 - Investigation of this problem in ongoing.

2. ASCII2NC Tool

- Stands for "ASCII to NetCDF"
- Functionality:
 - Reformat ASCII point observations into intermediate NetCDF format.
 - One input ASCII format supported (10 columns):
 - No configuration file.

• Data formats:

- Reads Met_Point ASCII.
- Writes point NetCDF as input to Point-Stat.
- Future: support additional standard ASCII formats.

ASCII2NC: Usage

Usage: ascii2nc
 ascii_file
 netcdf_file
 [-format ascii_format]
 [-v level]

ascii_file	Input obs file in ASCII format
netcdf_file	Output name for NetCDF file
-format	Override default MET_Point format (Future Option)
-V	Level of logging

MET_Point Format

Msg	STID	ValidTime	Lat	Lon	Elev	GC L	vl	Hgt	Ob	_	
ADPUPA ADPUPA ADPUPA ADPUPA ADPUPA	72365 72365 72365 72365 72365	20070331_120000 20070331_120000 20070331_120000 20070331_120000 20070331_120000 20070331_120000 20070331_120000	35.03 - 35.03 - 35.03 - 35.03 - 35.03 -	-106.62 -106.62 -106.62 -106.62 -106.62	1618.0 1618.0 1618.0 1618.0	0 11 8 0 17 8 0 52 8 0 53 8 0 7 82	37.0 37.0 37.0 37.0) 1618) 1618) 1618) 1618 1724	3 273.05 3 271.85 3 92 3 0.00417 1724	* Use a value of "-9999" to indicate missing data	

Msg	Message type
STID	WMO Station ID
Lat	Latitude [North]
Lon	Longitude [East]
Elev	Elevation [m] (Note: currently not used by MET code so can be filled with -9999.)
GC	GRIB code for variable (i.e. AccPrecip = 61; MSLP = 2; Temp = 11, etc)
	http://www.cpc.ncep.noaa.gov/products/wesley/opn_gribtable.html
LvI	Pressure [mb] or Accumulation Interval [hr]
Hgt	Height above Mean Sea Level [m – MSL] (Note: currently not used by MET code so can be filled with -9999.)
Ob	Observed value

ASCII2NC: Run

 METv2.0/bin/ascii2nc sample_obs.txt sample_ascii.nc -v 2

```
netcdf sample_ascii {
                                                                   Result of
                                                                                                    obs_arr =
dimensions:
                                                                                                     1, 7, 837, 1618, 1618,
                                                                   ncdump -h
               mxstr = 15;
                                                                                                     1, 11, 837, 1618, 273.05,
               hdr_arr_len = 3;
                                                                                                    1, 17, 837, 1618, 271.85,
               obs_arr_len = 5;
                                                                                                     1, 52, 837, 1618, 92,
                                                                   Result of
               nhdr = 5:
                                                                                                     1, 53, 837, 1618, 0.00417,
                                                                   ncdump -v obs_arr
               nobs = UNLIMITED; // (2140 currently)
                                                                                                     1, 7, 826, 1724, 1724,
variables:
                                                                                                     1, 11, 826, 1724, 274.55,
               char hdr_typ(nhdr, mxstr) ;
                                                                                                    1, 17, 826, 1724, 272.15,
                              hdr_typ:long_name = "message type";
                                                                                                     1, 52, 826, 1724, 84,
               char hdr_sid(nhdr, mxstr) ;
                                                                                                     1, 53, 826, 1724, 0.00432,
                               hdr_sid:long_name = "station identification";
                                                                                                     1, 7, 815.3, 1829, 1829,
               char hdr_vld(nhdr, mxstr) ;
                                                                                                     1, 11, 815.3, 1829, 276.45,
                              hdr vld:long name = "valid time";
                                                                                                    1, 17, 815.3, 1829, 265.75,
                              hdr_vld:units = "YYYYMMDD_HHMMSS UTC";
                                                                                                    1, 52, 815.3, 1829, 45,
               float hdr_arr(nhdr, hdr_arr_len);
                                                                                                     1, 53, 815.3, 1829, 0.0027,
                               hdr arr:long name = "array of observation station header values";
                                                                                                    1, 7, 815, 1832, 1832,
                               hdr arr: fill value = -9999.f:
                                                                                                     1, 11, 815, 1832, 276.55,
                              hdr_arr:columns = "lat lon elv";
                                                                                                    1, 17, 815, 1832, 265.55,
                                                                                                    1, 52, 815, 1832, 44,
               float obs_arr(nobs, obs_arr_len);
                                                                                                    1, 53, 815, 1832, 0.00266,
                               obs_arr:long_name = "array of observation values";
                                                                                                    1, 7, 784.7, 2134, 2134,
                               obs_arr:_fill_value = -9999.f;
                                                                                                     1, 11, 784.7, 2134, 274.05,
                               obs_arr:columns = "hdr_id gc lvl hgt ob";
                                                                                                     1, 17, 784.7, 2134, 264.15,
                               obs_arr:hdr_id_long_name = "index of matching header data";
                                                                                                     1, 52, 784.7, 2134, 47,
```

3. PCP-Combine Tool

- Stands for "Precip-Combine"
- Functionality:
 - Mathematically combines precipitation fields across multiple files.
 - Add precipitation over 2 files
 - 2 NMM output files to go from 3-hr to 6-hr accumulation.
 - Sum precipitation over more than 2 files
 - 12 ARW output files to go from 1-hr to 12 hr accumulation.
 - Subtract precipitation in 2 files
 - 2 Level II radar files).
 - Specify field name on the command line.
 - No configuration file.

Data formats:

- Reads GRIB format.
- Writes gridded NetCDF as input to stats tools.

PCP-Combine: Usage

Usage: pcp_combine
[-sum sum_args]
or [-add add_args]
or [-subtract sub_args]
[-gc code]

[-v level]

-sum	Accumulates data over multiple files. Sum_args: (init_time, in_accum, valid_time, out_accum, out_file, -pcpdir path, -pcprx reg_exp)
-add	Accumulates data over two files. Add_args: (in_file1, Accum1, in_file2, Accum2, out_file).
-subtract	Subtracts data over two files. Sub_args: (in_file1, Accum1, in_file2, Accum2, out_file).
-gc	GRIB code for variable (i.e. AccPrecip = 61; MSLP = 2; Temp = 11, etc).
-V	Level of logging

PCP-Combine: Run

Two example command lines

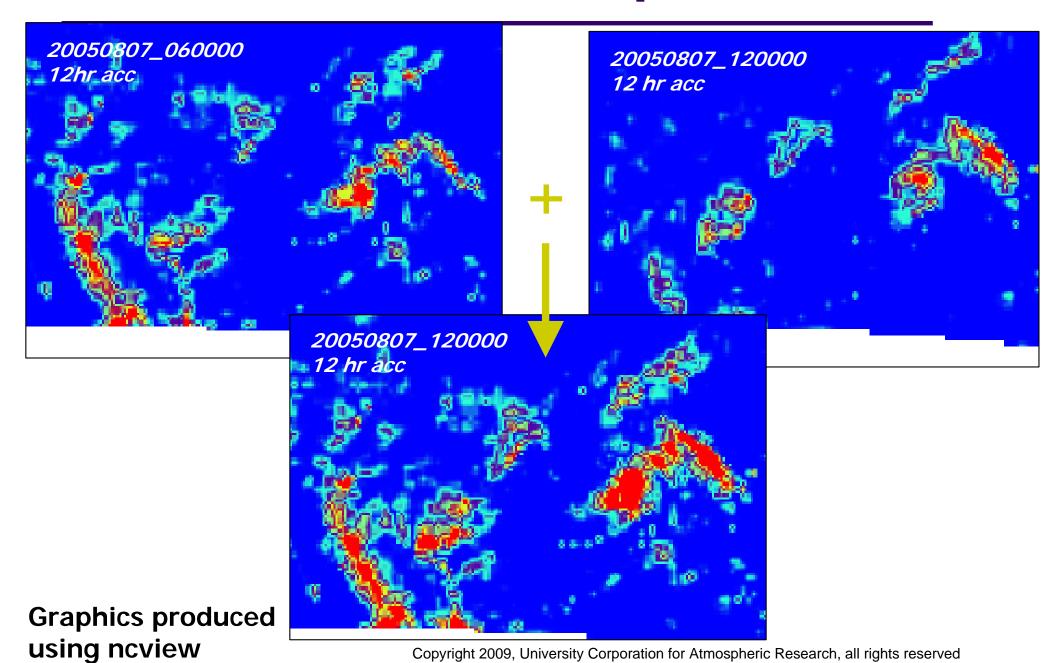
1) Summing two 6-hourly accumulation forecast files into a single 12-hour accumulation forecast.

```
METv2.0/bin/pcp_combine \
-add 20050807_060000 6 20050807_120000 6 \
sample_fcst.nc \
-pcpdir data/2005080700
```

2) Summing 12 1-hourly accumulation observation files into a single 12-hour accumulated observation.

```
METv2.0/bin/pcp_combine \
-sum 00000000_000000 1 20050807_120000 12 \
sample_obs.nc \
-pcpdir data/ST2ml
```

PCP-Combine: Example #1



GRIB vs GRIB2

- GRIB (or GRIB1): WRF postprocessor (WPP) produces GRIB1 format using copyGB
 - Performs horizontal interpolation and destaggering (in the case of WRF-NMM) onto a defined grid.
 - Useful for both cores in creating an output grid not fixed by the model integration domain.
 - http://www.dtcenter.org/wrf-nmm/users/downloads/
- GRIB2: NCEP and other WMO organizations have historical data in GRIB1 but now use GRIB 2 as the standard for gridded binary data.
- At least two GRIB2toGRIB1 converters available
 - NCEP cnvgrib
 - http://www.nco.ncep.noaa.gov/pmb/codes/GRIB2/
 - UCAR/CISL Grib Converter
 - http://dss.ucar.edu/libraries/grib/c.html

Data Inventory Tools

- wgrib dumps GRIB1 headers and data.
 - http://www.cpc.ncep.noaa.gov/products/wesley/wgrib.html
- wgrib2 dumps GRIB2 headers and data.
 - http://www.cpc.ncep.noaa.gov/products/wesley/wgrib2/
- ncdump dumps NetCDF headers and data.
 ncview plots gridded NetCDF data.
 - http://www.unidata.ucar.edu/software/netcdf/
- GrADS command line interface to produce plots.
 - http://www.iges.org/grads/downloads.html
- NCL command line interface to produce plots.
 - http://www.ncl.ucar.edu/
- IDV gui-driven visualization of many gridded and point datasets.
 - http://www.unidata.ucar.edu/software/idv/

4. Gen Poly Mask Tool

Stands for "General Polyline Mask"

Functionality:

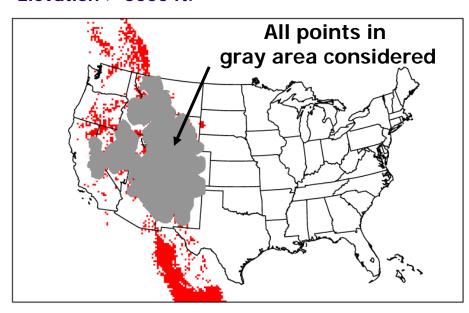
- Pre-computes a lat/lon polyline to generate a 0/1 mask field to be applied to your data.
- Applies this mask once prior to running Point-Stat or Grid-Stat
- No configuration file.

Data formats:

- Reads Ascii formatted polyline file.
- Reads GRIB file.
- Reads NetCDF files from PCP-Combine.
- Writes gridded NetCDF file of 0/1 mask.

Example polyline mask

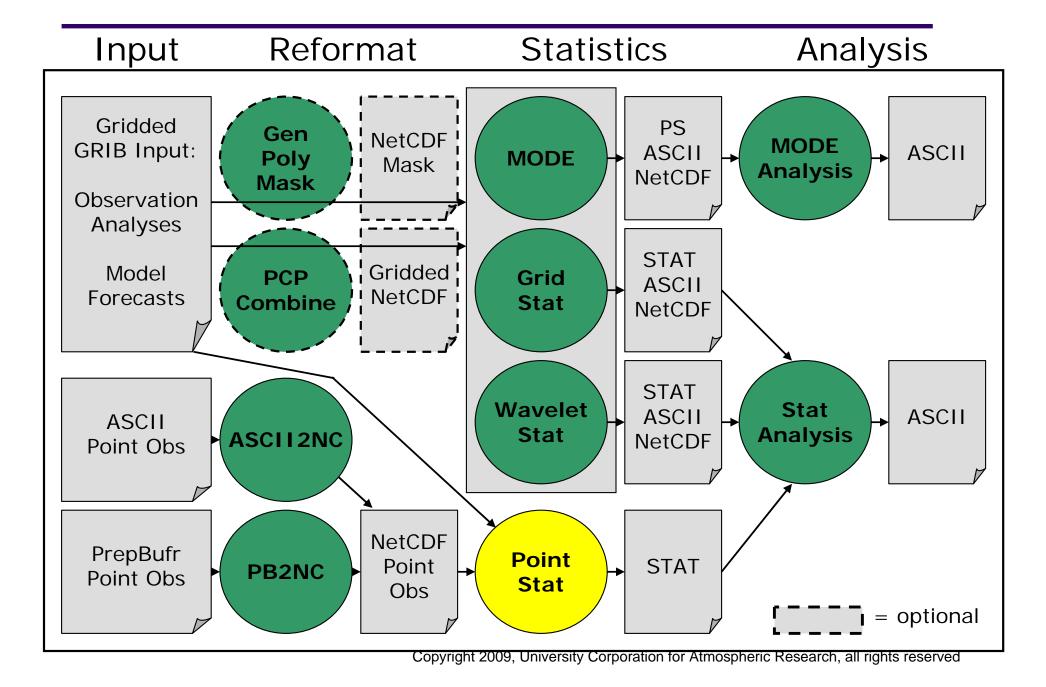
Coordinates generated from underlying points of Elevation > 5000 ft.



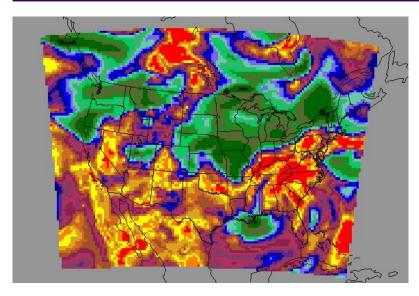
Example polyline file: data/poly/CONUS.poly

CONUS
35.81352 -122.33439
36.01565 -121.51420
35.35092 -121.27343
34.68935 -121.03940
34.03100 -120.81184
34.21258 -120.01634
33.55392 -119.80508
...

Point-Stat Tool



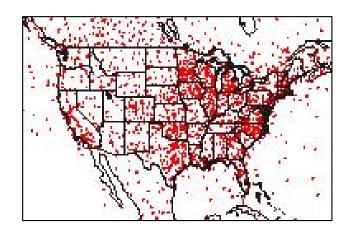
Point-Stat: Overview



- Compare gridded forecasts to point observations.
- Accumulate matched pairs over a defined area at a single point in time.
- Verify one or more variables/levels.
- Analysis tool provided to aggregate through time.

Verification methods:

- Continuous statistics for raw fields.
- Categorical counts and statistics for thresholded fields.
- Parametric and non-parametric confidence intervals for statistics.
- Compute partial sums for raw fields and/or the raw matched pair values.
- Probabilistic methods.



Point-Stat: Input/Output

- Input Files
 - Gridded forecast file
 - GRIB output of WRF Post-Processor (or other)
 - NetCDF output of PCP-Combine
 - Point observation file
 - NetCDF output of PB2NC
 - NetCDF output of ASCII2NC
 - ASCII configuration file
- Output Files
 - ASCII statistics file with all output lines (end with ".stat")
 - Optional ASCII files sorted by line type with a header row (ends with "_TYPE.txt")

Point-Stat: Usage

Usage: point_stat fcst_file obs_file config_file [-climo climo_file] [-ncfile netcdf_file] [-valid_beg time] [-valid_end time] [-outdir path] [-v level]

fcst_file	Forecast file in GRIB or NetCDF
obs_file	Point observation file in NetCDF (PB2NC or ASCII2NC)
config_file	ASCII configuration file
-climo	Climatological file for computing anomaly partial sums
-ncfile	Additional point observation files in NetCDF
-valid_beg	Beginning of valid time window for matching
-valid_end	End of valid time window for matching
-outdir	Output directory to be used
-V	Level of logging

Point-Stat: Configuration

- 25 configurable parameters only set a few:
 - Temperature at the surface (2-meter).
 - fcst_field[] = ["TMP/Z2"];
 - Temperature below freezing.
 - fcst_thresh[] = ["le273"];
 - Match to observations at the surface.
 - message_type[] = ["ADPSFC"];
 - Look at all the points in my domain.
 - mask_grid[] = ["FULL"];
 - Match observation to the nearest forecast value.
 - interp_wdth[] = [1];
 - Generate all possible statistic types.
 - output_flag[] = [2, 2, 2, 2, 2, 2, 2, 2, 2];

Point-Stat: Run

METv2.0/bin/point_stat \
 sample_fcst.grb sample_pb.nc \
 PointStatConfig_TMPZ2 -outdir out -v 2

```
Forecast File: sample fcst.grb
Climatology File: none
Configuration File: PointStatConfig TMPZ2
Observation File: sample pb.nc
Reading records for TMP/Z2.
For TMP/Z2 found 1 forecast levels and 0 climatology levels.
Searching 179772 observations from 44076 PrepBufr messages.
Processing TMP/Z2 versus TMP/Z2, for observation type ADPSFC, over region FULL, for interpolation method
UW MEAN(1), using 11370 pairs.
Computing Categorical Statistics.
Computing Continuous Statistics.
Computing Scalar Partial Sums.
Output file: out/point stat 360000L 20070331 120000V.stat
Output file: out/point stat 360000L 20070331 120000V fho.txt
Output file: out/point stat 360000L 20070331 120000V ctc.txt
Output file: out/point_stat_360000L_20070331_120000V_cts.txt
Output file: out/point stat 360000L 20070331 120000V cnt.txt
Output file: out/point_stat_360000L_20070331_120000V_s1112.txt
Output file: out/point stat 360000L 20070331 120000V sal112.txt
Output file: out/point_stat_360000L_20070331_120000V_v1112.txt
Output file: out/point stat 360000L 20070331 120000V val112.txt
Output file: out/point stat 360000L 20070331 120000V mpr.txt
```

Point-Stat: ASCII Output Types

- Statistics line types: 9 possible
 - Categorical apply threshold
 - Contingency table counts (FHO, CTC)
 - Contingency table statistics (CTS)
 - Continuous raw fields
 - Continuous statistics (CNT)
 - Partial Sums (SL1L2, SAL1L2, VL1L2, VAL1L2)
 - Matched Pairs
 - Raw matched pairs a lot of data! (MPR)
 - Additional line type(s) for probabilistic methods
- 21 header columns common to all line types
- Remaining columns specific to each line type

Point-Stat: Sample Output

- STAT file output for sample run:
 - 1 line each for FHO, CTC, CTS, CNT, and SL1L2
 - 11,370 lines for MPR!
- 6 additional TXT files with lines sorted by type

Point-Stat: CTC Output Line

VERSION	V2.0
MODEL	WRF
FCST_LEAD	360000
FCST_VALID_BEG	20070331_120000
FCST_VALID_END	20070331_120000
OBS_LEAD	000000
OBS_VALID_BEG	20070331_103000
OBS_VALID_END	20070331_133000
FCST_VAR	TMP
FCST_LEV	Z2
OBS_VAR	TMP
OBS_LEV	Z2
OBTYPE	ADPSFC

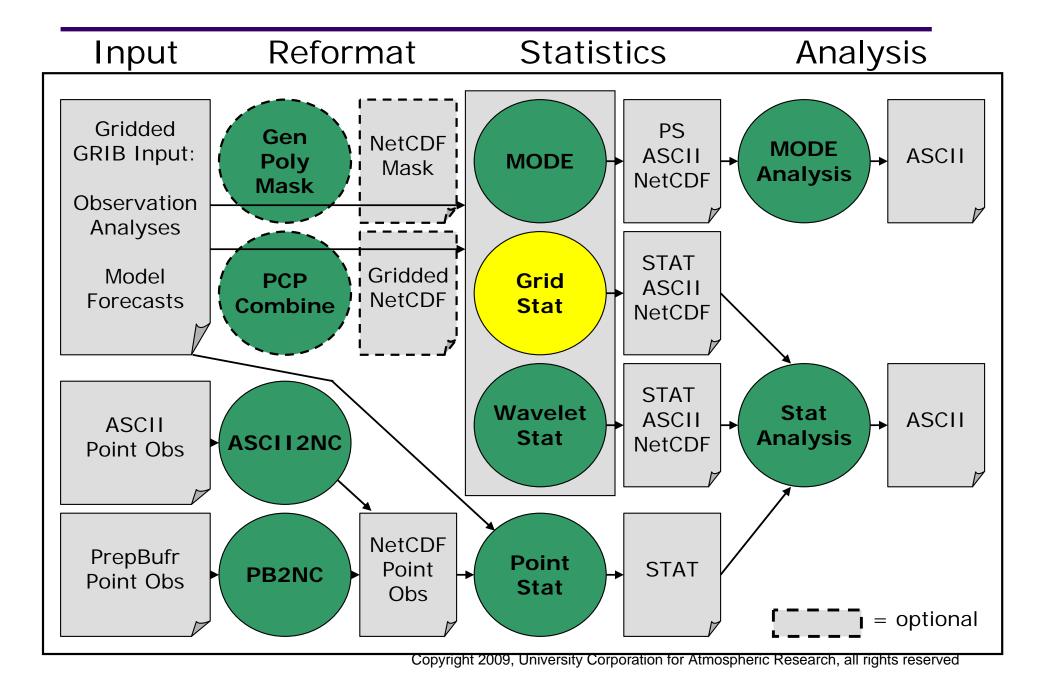
VX_MASK	FULL
INTERP_MTHD	UW_MEAN
INTERP_PNTS	1
FCST_THRESH	<273.000
OBS_THRESH	<273.000
COV_THRESH	NA
ALPHA	NA
LINE_TYPE	CTC
TOTAL	11370
FY_OY	1635
FY_ON	380
FN_OY	438
FN_ON	8917

Point-Stat: Matched Pairs

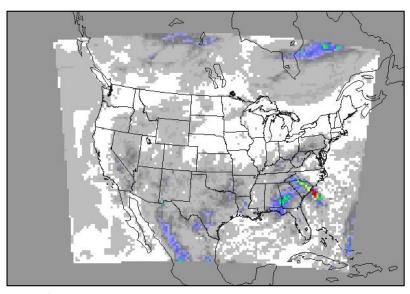
- Matched Pair (MPR) line type contains raw matched pairs.
 - Data overload!

TOTAL	INDEX	OBS_LAT	OBS_LON	OBS_LVL	OBS_ELV	FCST	OBS	CLIMO
11370	1	43.93000	-60.01000	1010.79999	4.01053	271.87788	271.54999	NA
11370	2	46.43000	-71.93000	1016.09998	102.04903	268.50255	269.45001	NA
11370	3	44.23000	-78.36000	1004.50000	191.44466	272.94013	272.35001	NA
11370	4	51.67000	-124.40000	916.50000	872.82202	263.69020	264.95001	NA
11370	5	58.61000	-117.16000	973.90002	337.50449	272.37757	270.95001	NA
11370	6	52.18000	-122.04000	906.50000	938.08594	271.37738	264.35001	NA
11370	7	50.68000	-127.36000	1020.20001	22.03931	275.44020	275.04999	NA
11370	8	50.45000	-100.59000	949.09998	562.38477	272.18978	271.75000	NA
11370	9	57.13000	-61.47000	899.70001	834.87476	258.06464	254.64999	NA
11370	10	47.56000	-59.16000	1000.90002	40.06803	272.06486	269.54999	NA
11370	11	47.41000	-72.79000	1006.90002	169.37592	266.37724	265.95001	NA
11370	12	45.76000	-62.68000	1014.00000	1.99518	268.94018	268.64999	NA
11370	13	49.24000	-65.33000	1014.90002	28.96468	264.25276	267.25000	NA
11370	14	43.29000	-79.79000	1017.79999	77.03765	273.56474	275.85001	NA
11370	15	48.78000	-123.04000	1015.70001	23.93772	278.12724	280.25000	NA

Grid-Stat Tool



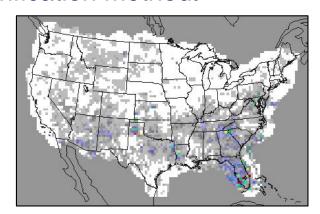
Grid-Stat: Overview



- Compare gridded forecasts to gridded observations on the same grid.
- Accumulate matched pairs over a defined area at a single point in time.
- Verify one or more variables/levels.
- Analysis tool provided to aggregate through time.

Verification methods:

- Continuous statistics for raw fields.
- Categorical counts and statistics for thresholded fields.
- Parametric and non-parametric confidence intervals for statistics.
- Compute partial sums for raw fields.
- Probabilistic methods.
- Continuous statistics and categorical counts/statistics using neighborhood verification method.



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Grid-Stat: Input/Output

Input Files

- Gridded forecast and observation files
 - GRIB output of WRF Post-Processor (or other)
 - NetCDF output of PCP-Combine
- ASCII configuration file
- Output Files
 - ASCII statistics file with all output lines (end with ".stat")
 - Optional ASCII files sorted by line type with a header row (ends with "_TYPE.txt")
 - Optional NetCDF matched pairs file

Grid-Stat: Usage

Usage: point_stat

fcst_file
obs_file
config_file
[-outdir path]
[-v level]

fcst_file	Forecast file in GRIB or NetCDF
obs_file	Observation file in GRIB or NetCDF
config_file	ASCII configuration file
-outdir	Output directory to be used
-V	Level of logging

Grid-Stat: Configuration

- 24 configurable parameters only set a few:
 - Precipitation accumulated over 24 hours.
 - fcst_field[] = ["APCP/A24"];
 - Any rain and moderate rain.
 - fcst_thresh[] = ["gt0.00, gt20.00"]; (mm)
 - Look at all the points and only the eastern United States.
 - mask_grid[] = ["FULL"];
 - mask_poly[] = ["EAST.poly"];
 - Compute neighborhood statistics with two sizes.
 nbr_width[] = [3, 5];
 - Generate all possible statistic types.
 - output_flag[] = [2, 2, 2, 2, 2, 2, 2, 2, 2];

Grid-Stat: Run

METv2.0/bin/grid_stat \
 sample_fcst.grb sample_obs.nc \
 GridStatConfig_APCP24 -outdir out -v 2

```
Forecast File: sample fcst.grb
Observation File: sample obs.nc
Configuration File: GridStatConfig APCP24
Processing APCP/A24 versus APCP/A24, for interpolation method UW_MEAN(1), over region FULL, using 6412 pairs.
Computing Categorical Statistics.
Computing Continuous Statistics.
Processing APCP/A24 versus APCP/A24, for interpolation method UW MEAN(1), over region EAST, using 2586 pairs.
Computing Categorical Statistics.
Computing Continuous Statistics.
Processing APCP/A24 versus APCP/A24, for interpolation method NBRHD(9), raw thresholds of >0.000 and >0.000,
over region EAST, using 5829 pairs.
... MORE NEIGHBORHOOD COMPUTATIONS ...
Output file: out/grid stat 240000L 20050808 000000V.stat
Output file: out/grid stat 240000L 20050808 000000V fho.txt
Output file: out/grid_stat_240000L_20050808_000000V_ctc.txt
Output file: out/grid stat_240000L_20050808_000000V_cts.txt
Output file: out/grid stat 240000L 20050808 000000V cnt.txt
Output file: out/grid_stat_240000L_20050808_000000V_sl112.txt
Output file: out/grid stat 240000L 20050808 000000V nbrctc.txt
Output file: out/grid stat 240000L 20050808 000000V nbrcts.txt
Output file: out/grid stat 240000L 20050808 000000V nbrcnt.txt
Output file: out/grid_stat_240000L_20050808_000000V_pairs.nc
```

Grid-Stat: ASCII Output Types

- Statistics line types: 8 possible
 - Same as Point-Stat
 - FHO, CTC, CTS, CNT, and SL1L2
 - Omitted for Grid-Stat
 - SAL1L2, VL1L2, or VAL1L2
 - Neighborhood define neighborhood, apply threshold
 - Neighborhood continuous statistics (NBRCNT)
 - Neighborhood contingency table counts (NBRCTC)
 - Neighborhood contingency table statistics (NBRCTS)
 - Additional line type(s) for probabilistic methods
- 21 header columns common to all line types
- Remaining columns specific to each line type

Grid-Stat: Sample Output

- STAT file output for sample run:
 - 2 lines each for CNT and SL1L2
 - = 2 verification regions (FULL and EAST)
 - 4 lines each for FHO, CTC, and CTS
 - = 2 regions * 2 thresholds
 - 8 lines each for NBRCNT, NBRCTC, NBRCTS
 - = 2 regions * 2 thresholds * 2 neighborhood sizes
- 8 additional TXT files with lines sorted by type
- NetCDF file containing matched pairs

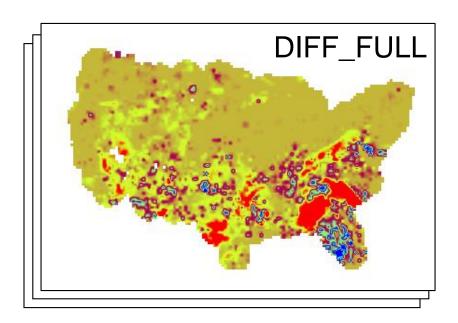
Grid-Stat: CTC Output Line

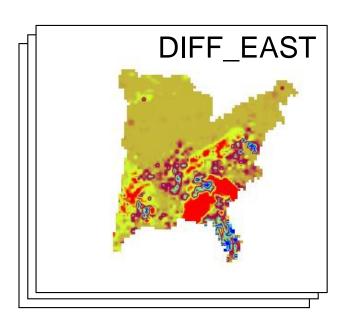
VERSION	V2.0
MODEL	WRF
FCST_LEAD	240000
FCST_VALID_BEG	20050808_000000
FCST_VALID_END	20050808_000000
OBS_LEAD	000000
OBS_VALID_BEG	20050808_000000
OBS_VALID_END	20050808_000000
FCST_VAR	APCP
FCST_LEV	A24
OBS_VAR	APCP
OBS_LEV	A24
OBTYPE	MC_PCP

VX_MASK	EAST
INTERP_MTHD	UW_MEAN
INTERP_PNTS	1
FCST_THRESH	>=20.000
OBS_THRESH	>=20.000
COV_THRESH	NA
ALPHA	NA
LINE_TYPE	CTC
TOTAL	2586
FY_OY	5
FY_ON	104
FN_OY	70
FN_ON	2407

Grid-Stat: NetCDF Matched Pairs

- Forecast, observation, and difference fields for each combination of...
 - Variable, level, masking region, and interpolation method (smoothing)
- Sample output contains 6 variables:
 - FCST, OBS, and DIFF for FULL and EAST





Variations

- Grid-Stat and Point-Stat may be used to compare two different variables.
 - Not strictly verification.
 - Leave observation blank to use forecast setting.
 - Selecting variable/levels:
 - fcst_field[] = ["61/A24"];
 - obs_field[] = [];
 - Selecting thresholds:
 - fcst_thresh[] = ["gt0.0 ge20.0"];
 - obs_thresh[] = [];

Stat Analysis Tool

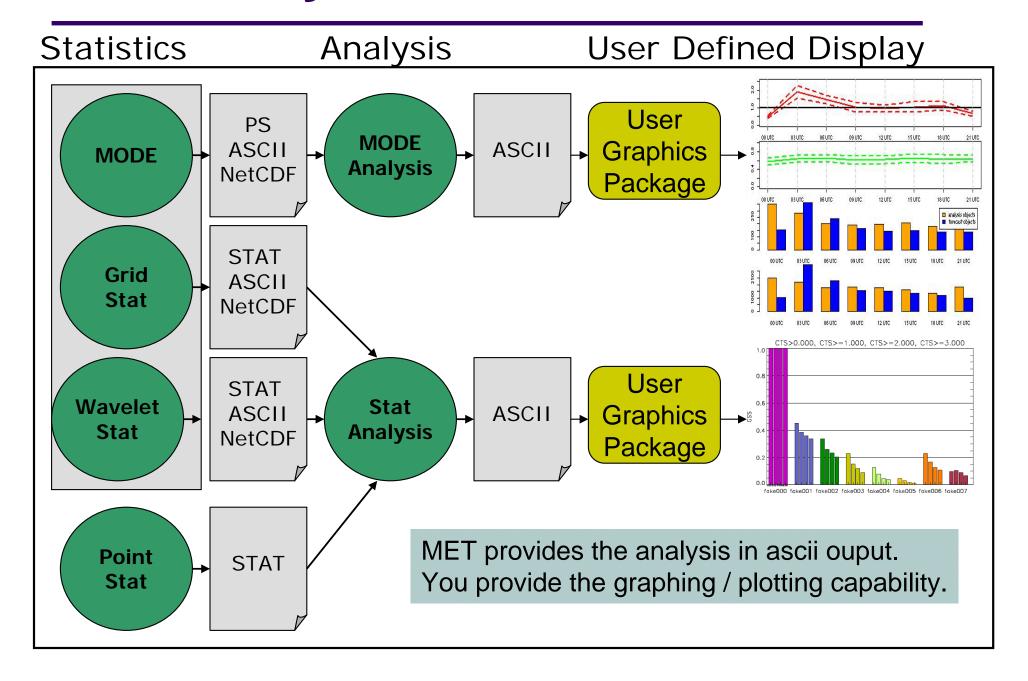
- Filtering
- Summarizing
- Aggregating

of Grid Stat, Point Stat, & Wavelet output

What can Stat Analysis do for you?

- ✓ Q: If I wanted the overall statistics for all the gridded observations compared to the forecasts for hours 0 through 24 together, can MET do this?
 - A: Yes using Stat Analysis Tool on Grid-Stat output
- ✓ Q: Can I compute long-term statistics at individual sites (eg, mean absolute error or RMS error for daily forecasts for a month).?
 - A: Yes using Stat Analysis Tool on Point-Stat output
- ✓ Q: Can I aggregate my contingency table statistics over multiple runs?
 - A: Yes using Stat Analysis Tool on any output
- Q: Can I aggregate statistics for a large number of individual stations in one simultaneous run?
 - A: No but you could get the aggregate statistics for each individual station and run Stat Analysis Tool n-times.

Stat Analysis Tool



Stat Analysis Jobs

Filtering

 stat_job_filter - filters out lines from one or more stat files based on user-specified filtering options.

Summarizing

- stat_job_summary produces summary information from a single column of data including:
 - mean, standard deviation, min, max, and the 10th, 25th, 50th, 75th, and 90th percentiles.

Customized tool for AFWA

 stat_job_go_index - computes the GO Index, a performance Statric used primarily by the United States Air Force.

Stat Analysis Jobs

Aggregation

- stat_job_aggr aggregates stat data across multiple time steps or masking regions to sum contingency table data or partial sums across multiple lines of data.
- stat_job_aggr_cts aggregates contingency table to produce aggregated contingency table statistics.
- stat_job_aggr_cnt computes as many continuous statistics from aggregated partial sums as possible.
- stat_job_aggr_mpr aggregates the matched pair output of Point-Stat and recomputes the requested stat line type.

PB2NC: Usage

Usage: stat_analysis

-lookin path

-out filename

[-v level]

[-config config_file]

or command line

options with associated

<u>arguments</u>

[stat_job_filter]

[stat_job_summary]

[stat_job_aggr]

[stat_job_cts]

[stat_job_aggr_cnt]

[stat_job_aggr_mpr]

[stat_job_go_index]

-lookin	Directory where *.stat files reside
-out	Output name for NetCDF file
-V	Level of logging
-config	StatAnalysisConfig file
stat_job_filter	See previous 2 slides
stat_job_summary	See previous 2 slides
stat_job_aggr	See previous 2 slides
stat_job_cts	See previous 2 slides
stat_job_aggr_cnt	See previous 2 slides
stat_job_aggr_mpr	See previous 2 slides
stat_job_go_index	See previous 2 slides

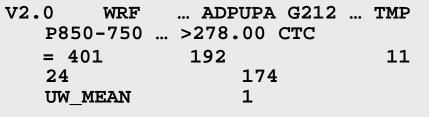
Point-Stat: Configuration

- 22 configurable parameters only set a few:
 - Apply NAM G212 mask
 - vx_mask[] = ["G212"];
 - Using only the Temperature variable
 - var[] = ["TMP"];
 - Filter on CTC lines that have been threshholded var[]>278
 - line_type[] = ["CTC"];
 - fcst_thres[] = [">278"];
 - Dump the filtered stat data to a file
 AND sum contingency table count (CTC) lines of data for pressure levels between 850 and 750
 - joblist[] = ["-job stat_job_filter -dump_row out/filter_job.stat",
 "-job stat_job_aggr -dump_row out/aggr_ctc_job.stat -level P850-750"];

Stat Analysis Tool: Run stat_job_aggr

"-job stat_job_aggr -dump_row out/aggr_ctc_job.stat -level P850-750"

Point Stat Output (i.e. point_stat_out.stat)



	OBS				
F		Y	N		
F C S T		192	11	203	
TN	N	24	174	198	
		216	185	401	

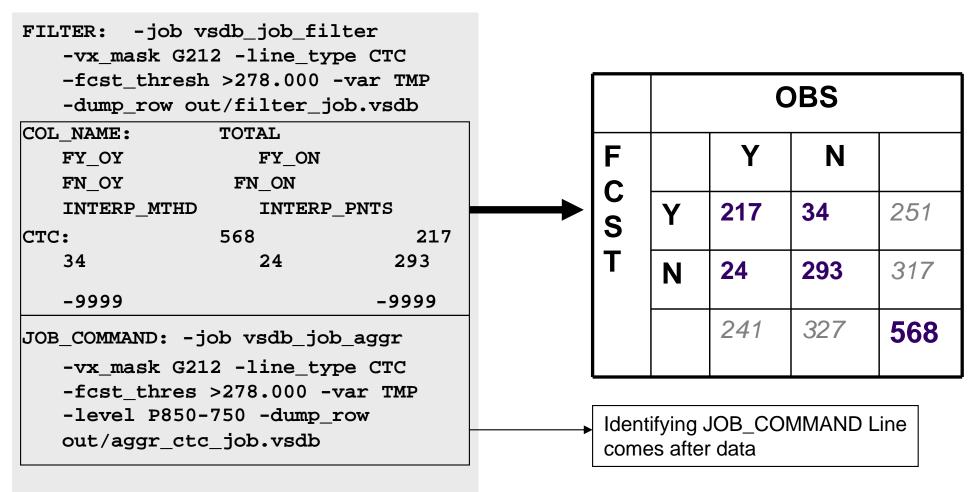
V2.0			ADPSFC G212	•••	TMP
I	?850 - 750	•••	>278.00 CTC		
=	167		25		23
()		119		
τ	JW_MEAN		1		

(NOTE: header modified to show only pertinent info)

23			OBS			
		F		Υ	N	
, only		C S	Υ	25	23	48
only		T	Z	0	119	119
				25	142	167
Copyright 2009,	University Corporation	for Atm	ospheric	Research,	all rights res	

Stat Analysis Tool: Run stat_job_aggr

Stat Analysis Output (i.e. stat_analysis.out)



Stat Analysis Tool: Run stat_job_summary

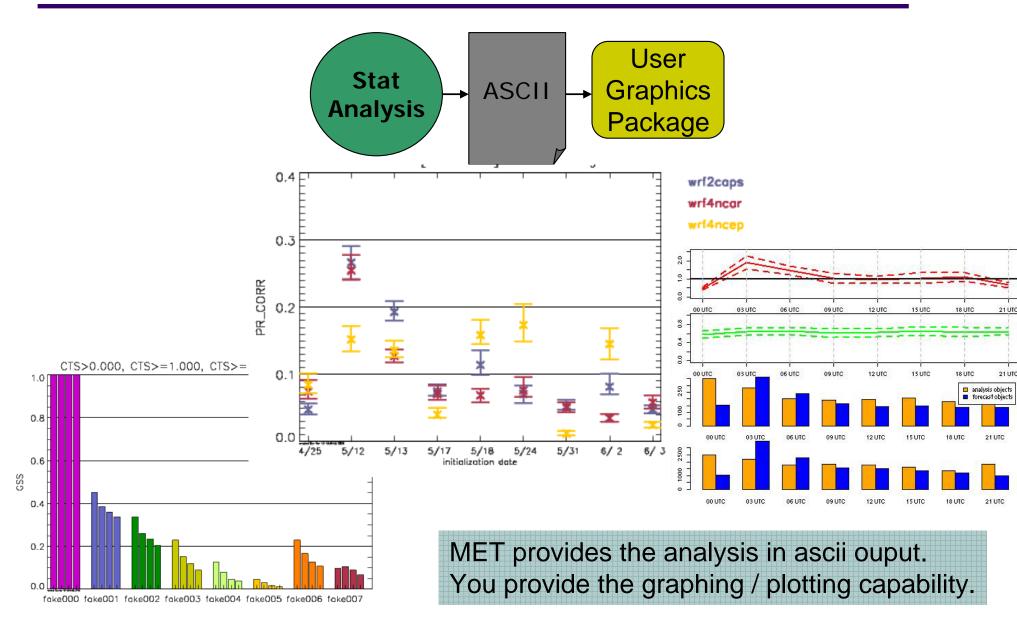
"-job stat_job_summary -line_type CNT -alpha = 0.50 -var TMP \
-dump_row ../out/job_summary_RMSE.vsdb -column RMSE"

Column Number	Description
1	Summary (job type)
2	Total
3-7	Mean* Includes normal and bootstrap upper and lower confidence limits
8-10	Standard deviation** Includes bootstrap upper and lower confidence limits
11	Minimum value
12	10 th percentile
13	25 th percentile
14	Median (50 th percentile)
15	75 th percentile
16	90 th percentile
17	Maximum value

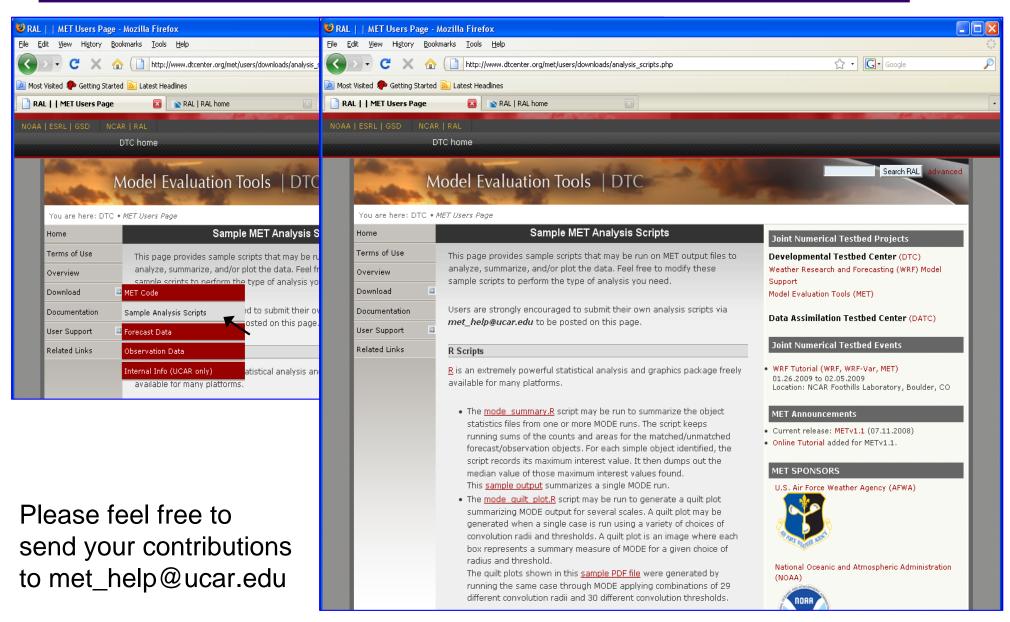
(stat_analysis.out cont.) COL NAME: TOTAL MEAN MEAN NCL MEAN NCU MEAN BCL MEAN BCU STDEV STDEV BCL STDEV BCU MIN P10 P25 P50 P75 P90 MAX SUMMARY: 4 1.98438 1.33219 2.63656 1.58837 2.29289 0.40986 0.04574 0.55950 1.41291 1.59671 1.87241 2.07130 2.18328 2.18328 2.30251 JOB COMMAND: -job vsdb job summary -line type CNT Identifying JOB_COMMAND Line

comes after data

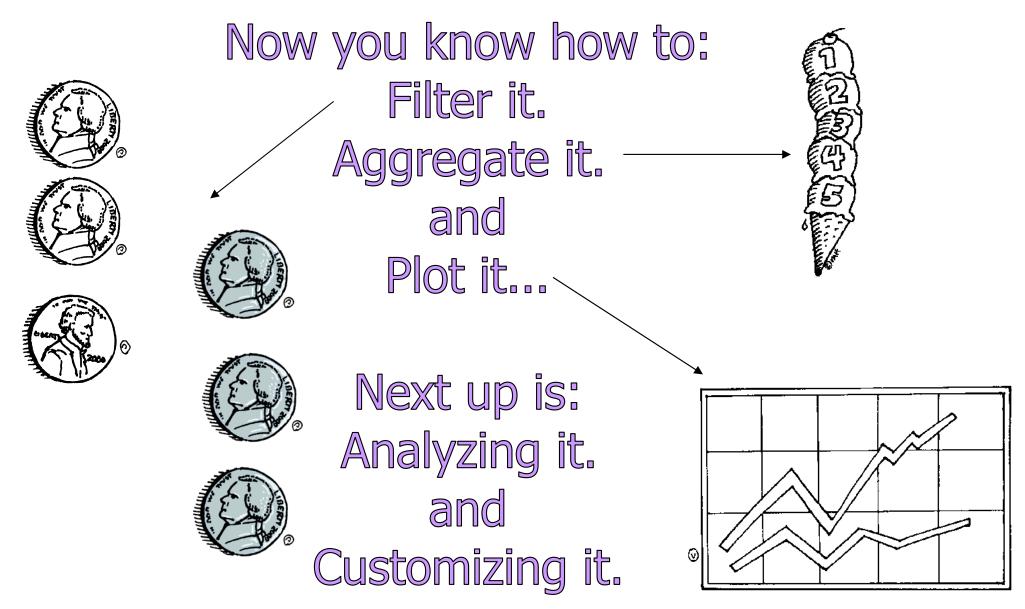
Use your favorite plotting software



User Contributed Plotting Scripts



Thanks - Any Questions?



Interpretation of Output – Grid and Point Stat

Tressa L. Fowler

Point and Grid Stat Output Lines

- CTC Contingency Table Counts
- CTS Contingency Table Statistics
- CNT Continuous Statistics
- FHO Forecast, Hit, Observation Rates
- SL1L2 Scalar L1L2 Partial Sums
- SAL1L2 Scalar Anomaly L1L2 Partial Sums when climatological data is supplied
- VL1L2 Vector L1L2 Partial Sums
- VAL1L2 Vector Anomaly L1L2 Partial Sums when climatological data is supplied
- MPR Matched Pair data

Available in the newest release MET 2.0, line types for probability forecasts.

Point and Grid stat output example

Header Line:

Data Line:

CTS 1.00000 0.56552 1.00000 NA NA 0.40000 0.11762 0.76928 5 NA NA 0.40000 0.11762 0.76928 NA NA 0.40000 NA NA 0.40000 NA NA NA NA NA NA 0.11762 0.76928 NA NA NA NA NA NA NA 0.00000 0.00000 0.43448 NA 0.40000 0.11762 0.76928 NA NA 0.00000 NA NA NA NA NA NA NA 0.00000 NA NA NA NA NA NA NA

See Appendix C of MET documentation for equations and details

- TOTAL Count of total pairs included in the stats for this line
- BASER base rate (e.g. sample climatological rate of event)
- FMEAN mean of forecast values
- ACC accuracy
- FBIAS frequency bias
- PODY probability of detection (events)
- PODN probability of detection (non-events)
- POFD probability of false detection
- FAR False Alarm Ratio
- CSI Critical success index
- GSS Gilbert Skill Score
- HK Hanssen-Kuipers Discriminant
- HSS Heidke Skill Score
- ODDS Odds Ratio
- Etc., etc., etc. . .

```
Confidence Intervals for some measures are included with the following suffixes:
```

```
_NCL = Normal Confidence Lower
```

_NCU = Normal Confidence Upper

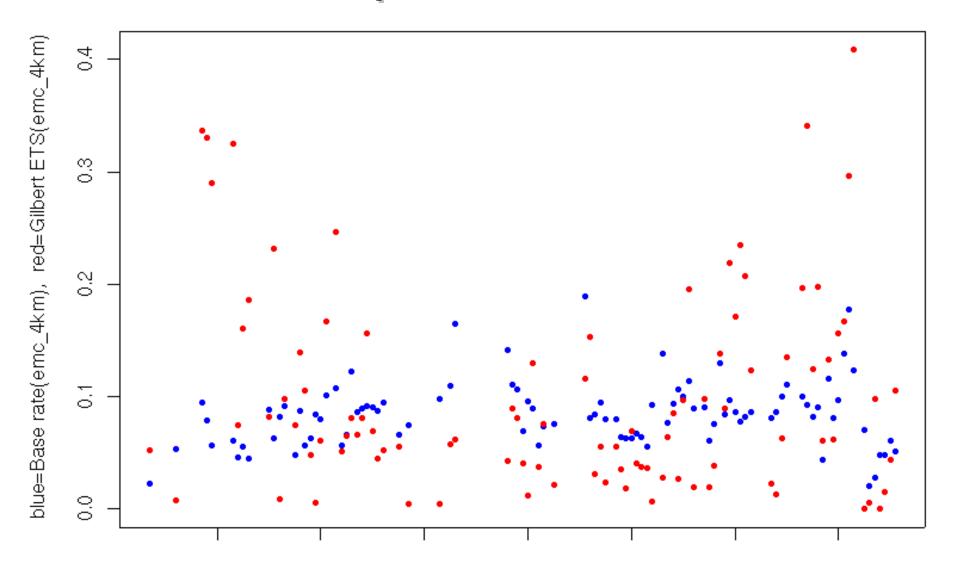
_BCL = Bootstrap Confidence Lower

_BCU = Bootstrap Confidence Upper

Example of Appendix C info

- Accuracy for a 2x2 contingency table is defined as $n_{11} + n_{00}$
- That is, it is the proportion of forecasts that were either hits or correct rejections – the fraction that were correct.
- Accuracy ranges from 0 to 1; a perfect forecast would have an accuracy value of 1.
- Accuracy should be used with caution, especially for rare events, because it can be strongly influenced by large values of n_{00} .

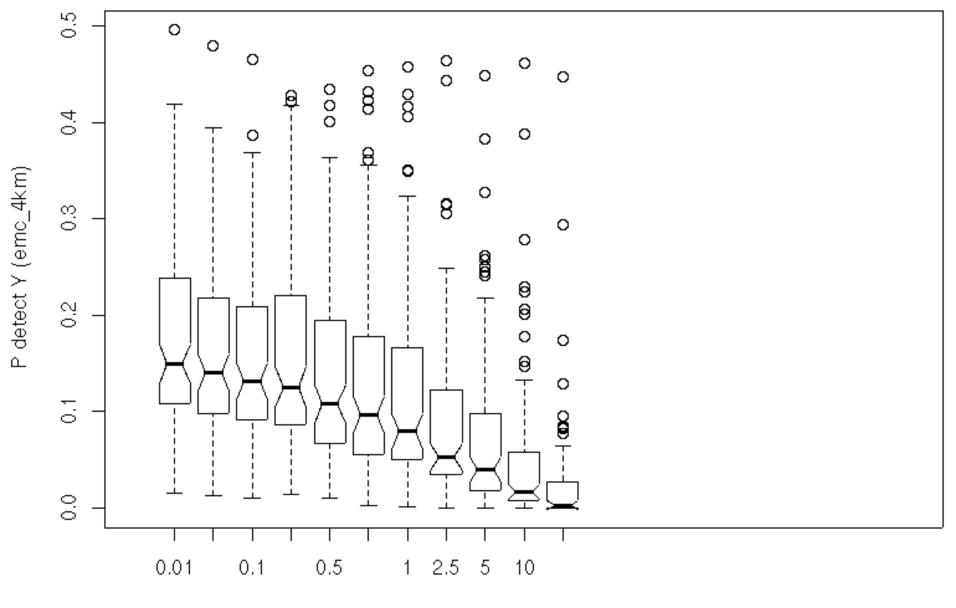
Example Scatter Plot



Forecast valid date

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Example Box Plot



Threshold

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Customizing Point-Stat and Grid-Stat Output

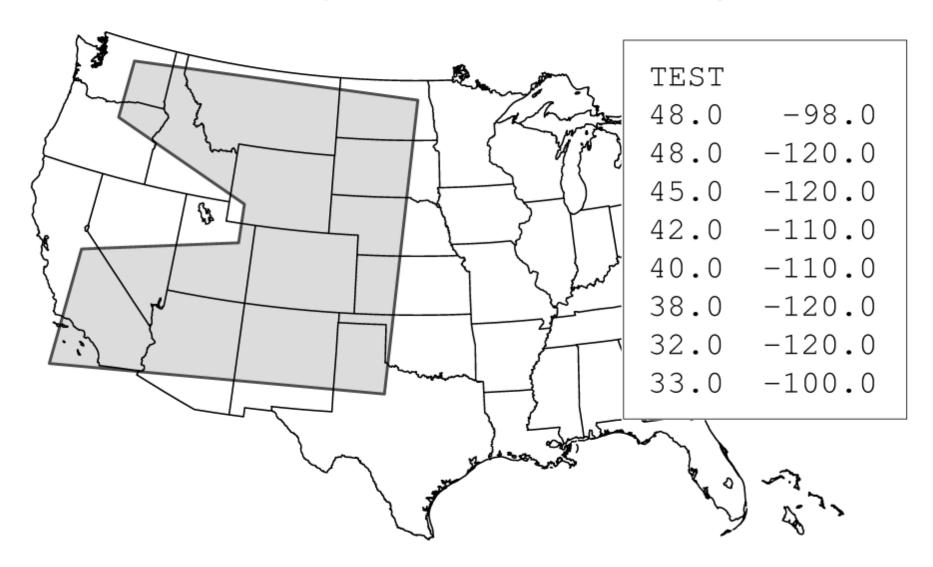
We'll restrict this discussion to an explanation of masking and interpolation.

Masking Methods

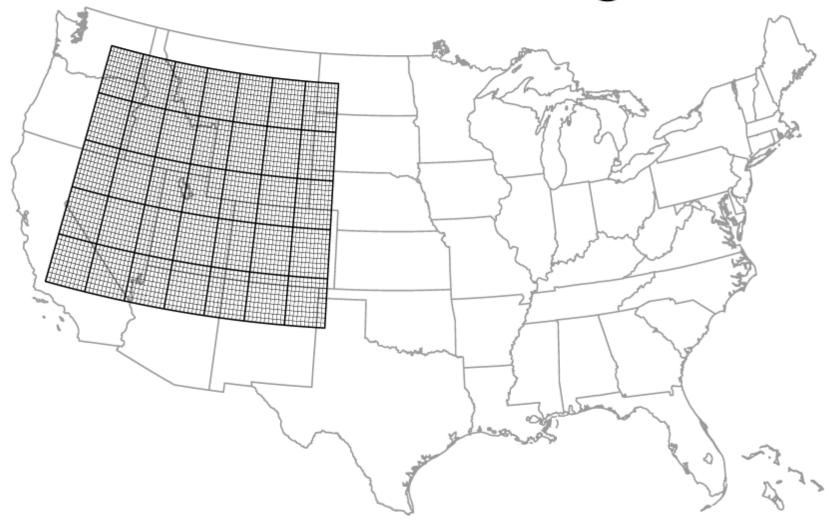
	Polyline	Grid	Stations	On/Off Bit Map
Point Stat				
Grid Stat			N/A	

Masking by stations doesn't apply to Grid Stat.

Polyline Masking



Grid Masking

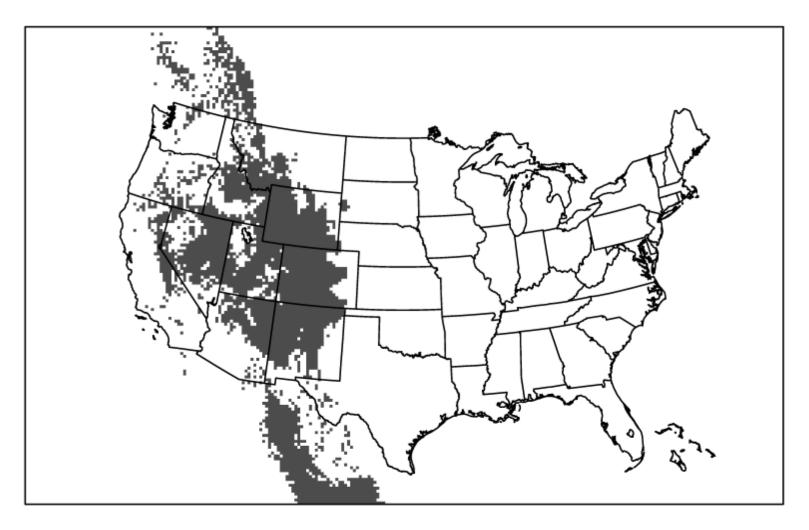


NCEP grids link: www.nco.ncep.noaa.gov/pmb/docs/on388/tableb.html

Station Masking



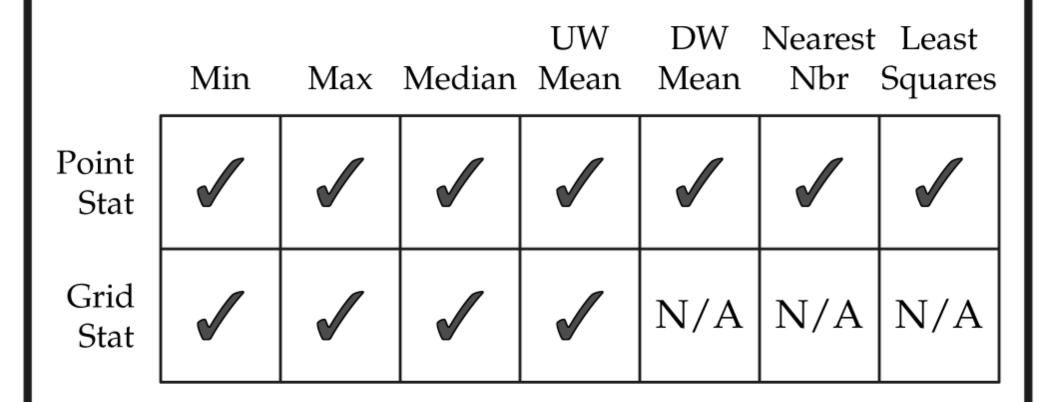
Data Threshold Mask



Topography > 5000 feet

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Interpolation Methods



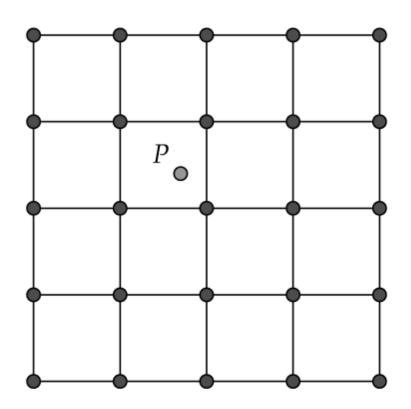
For Grid Stat, these are smoothing methods.

Interpolation

Need to Choose:

(1) Method

(2) Width



Min, Max, Median

Takes minimum, maximum or median of values in interpolation square.

Median separates the upper half of data values from the lower half. This is different from the mean, which is an average.

Nearest Neighbor

Essentially, no interpolation is performed.

Value at interpolation point is simply the data value at the closest grid point.

Unweighted Mean Distance-Weighted Mean

Unweighted Mean is the average.

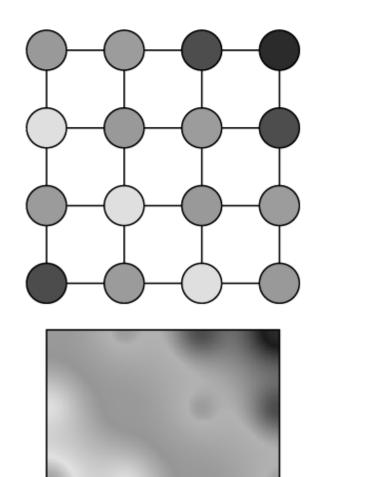
Distance-Weighted Mean is an average weighted according to distance from nearby grid points.

Least Squares

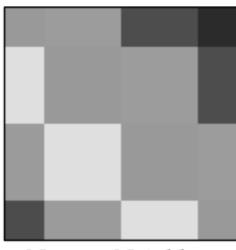
Performs a local Least-Squares linear fit in interpolation square.

$$z = Ax + By + C$$

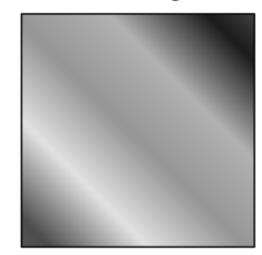
Interpolation Examples



DW Mean



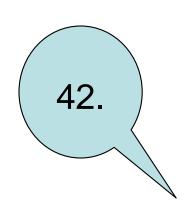
Nearest Neighbor



Least Squares

Giving meaning to your forecast verification results

What is the answer to life, the universe and everything?



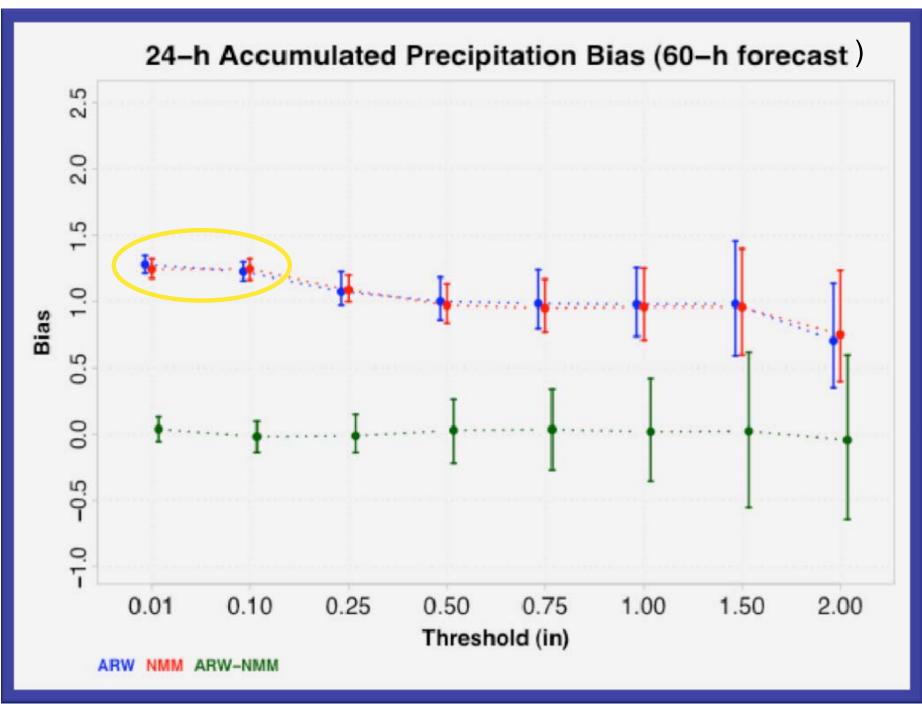
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Boulder, Colorado

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Accounting for Uncertainty

- Observational
- Model
 - Model parameters
 - Physics
 - Verification scores
- Sampling
 - Verification statistic is a realization of a random process.
 - What if the experiment were re-run under identical conditions?

Hypothesis Testing and Confidence Intervals

Hypothesis testing

- Given a null hypothesis (e.g., no bias), is there enough evidence to reject it?
- One- or two-sided, but test is against a single null hypothesis.

Confidence intervals

- Related to hypothesis tests, but more useful.
- How confident are we that the true value of the statistic (e.g., bias) is different from a particular value?
- Interpretation for most frequentist intervals is a bit awkward.

Hypothesis Testing and Confidence Intervals

"If we re-run the experiment 100 times, and create 100 $(1-\alpha)100\%$ Cl's, then we expect the true value of the parameter to fall inside $(1-\alpha)100$ of the intervals."



Hypothesis Testing and Confidence Intervals

Example: The difference in bias between two models is 0.01.

Hypothesis test: Is this different from zero?

Confidence interval: Does zero fall within the interval? Does 0.5 fall within the interval?

Confidence Intervals (CI's)

Parametric

- Assume the observed sample is a realization from a known population distribution with possibly unknown parameters (e.g., normal).
- Normal approximation Cl's are most common.
- Quick and easy.

Nonparametric

- Assume the distribution of the observed sample is representative of the *population* distribution.
- Bootstrap Cl's are most common.
- Can be computationally intensive, but easy enough.

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$$\hat{\theta} \pm z_{\alpha/2} se(\theta)$$

Is a $(1-\alpha)100\%$ Normal CI for Θ , where Θ is the statistic of interest (e.g., the forecast mean), se(Θ) is the standard error for the statistic, and z_v is the v-th quantile of the standard normal distribution.

Example: Let $X_1, ..., X_n$ be independent and identically distributed (iid) sample from a normal distribution with variance σ_X^2 .

Then, $\bar{X} = \frac{1}{n} \sum_{i=1}^{n} X_i$ is an estimate of the mean

of the sample. And a (1-α)100% CI is given by

$$\overline{X} \pm z_{\alpha/2} \frac{\sigma_X}{\sqrt{n}}$$

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- Numerous verification statistics can take this approximation in some form or another. In other cases (e.g., forecast/observation variance, linear correlation), different parametric Cl's can be used that still rely on the underlying sample's being iid normal.
- Contingency table verification scores such as probability of detection (POD) and false alarm ratio (FAR) also have normal approximation Cl's (for large enough sample sizes).

- Check the validity of the independence assumption.
- •Check the validity of the normal distribution (e.g., qq-plots, automatic tests exist too).

(cf. Gilleland, 2008)

IID Bootstrap Algorithm

- 1. Resample with replacement from the sample, X_1, \ldots, X_n .
- 2. Calculate the verification statistic(s) of interest from the resample in step 1.
- 3. Repeat steps 1 and 2 many times, say B times, to obtain a sample of the verification statistic(s).
- 4. Estimate (1-α)100% Cl's from the sample in step 3.

IID Bootstrap Algorithm: Types of Cl's

- 1. Percentile Method Cl's*
- 2. Bias-corrected and adjusted (BCa)*
- 3. ABC
- 4. Basic bootstrap Cl's
- 5. Normal approximation
- 6. Bootstrap-t

*1 and 2 are available In MET

Block Bootstrap Algorithm

- Block bootstrapping is one way to obtain bootstrap Cl's for dependent samples.
- Same as IID Bootstrap, but resample blocks of contiguous data points.
- Use percentile method for Cl's.
- Block sizes should be substantially larger than the correlation length, but substantially smaller than the sample size. Usually, one takes the greatest integer below the square root of the sample size.

Block Bootstrap Algorithm: Blocks

Non-overlapping (NBB)

$$Y_1 = \{X_1, \dots, X_l\}, \dots, Y_k = \{X_{(k-1)l+1}, \dots, X_{kl}\}, \dots, Y_b = \{X_{(b-1)l+1}, \dots, X_n\}$$

Moving (MBB)

$$Y_i = \{X_i, ..., X_{i+l-1}\}$$

Circular (CBB)

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Parametric Bootstrap Algorithm:

- Same as IID Bootstrap algorithm, but first model the dependence in the data, then in step 1, take random samples from the model instead of the data.
- Generally preferable to block bootstrap.
- Stronger assumptions about the sample than block bootstrap.
- Use percentile method for Cl's.

Configurable parameters for the bootstrap in MET. For example, usually it is appropriate to take resamples of size n, where *n* is the size of the original sample. However, there are cases where it is better to take samples of size, m < n (e.g., if the population distribution is heavy-tailed; see Gilleland, 2008).

Practical Considerations

- Point-stat is quicker than Grid-stat, so bootstrap is quicker with Point-stat.
- May be prohibitively computationally inefficient to bootstrap over an entire field (i.e., over several thousand points), but can also bootstrap the statistics for each field over time. Measures the (betweenfield) uncertainty of the estimates over time, rather than the within field uncertainty.
- Normal approximation intervals are quick, and generally accurate. Check the normality assumption!
- Check whether the samples (for either type of interval) are independent or not.

Thank you. Questions?

For more information, see:

Developmental Testbed Center, 2008. Model Evaluation Tools Version 1.1 (METv1.1) User's Guide. Available at: http://www.dtcenter.org/met/

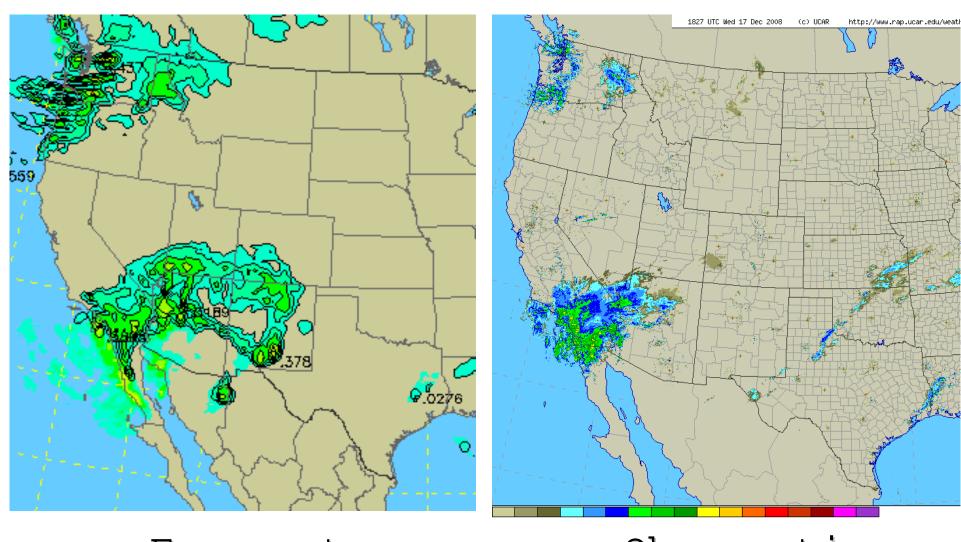
Gilleland E, 2008. Confidence intervals for forecast verification. *Submitted* as an NCAR Technical Note. Available at:

http://www.ral.ucar.edu/~ericg/Gilleland2008.pdf

Why Use Objects?

Tressa L. Fowler

Typical situation



Forecast

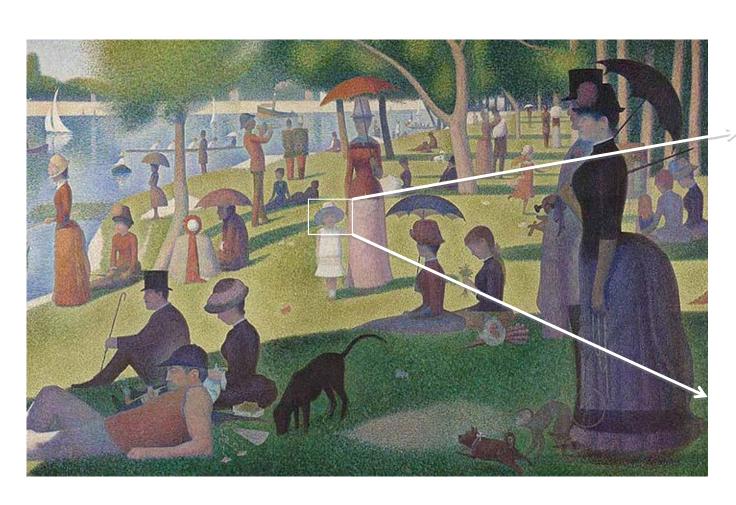
Observation

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Traditional verification matches up points, then sums them up.

Many forecasts are more than the sums of their parts.

Pixels or Pictures?





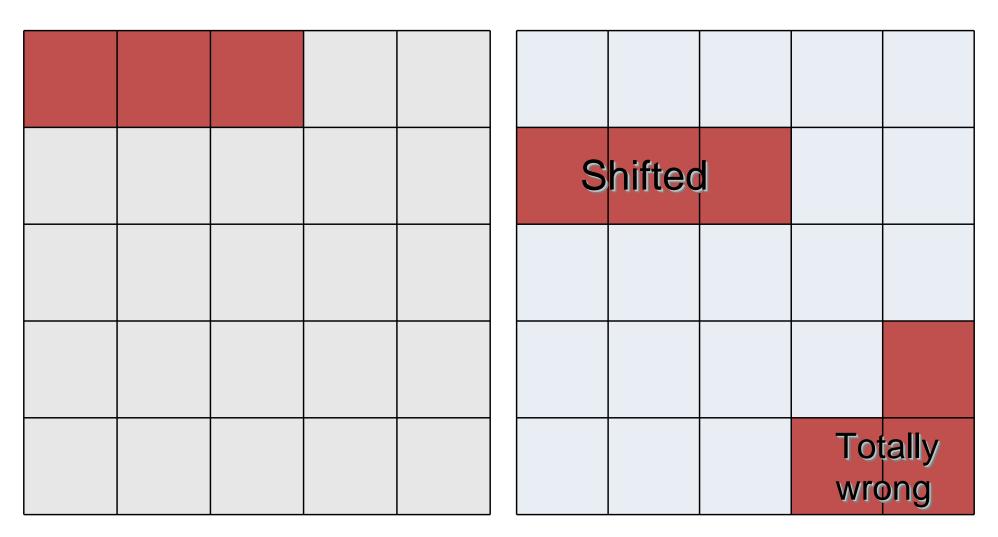
Object verification is more like what humans do.

Objects recognize the spatial relationship between points.

Simple example

Observed

Forecast



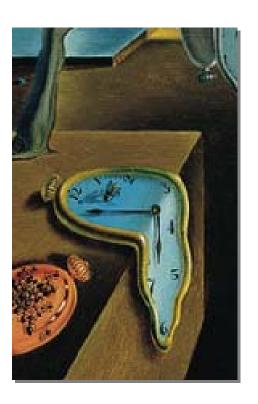
REAL - observed



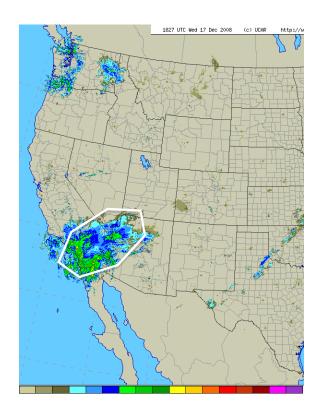
Forecast 1 — Distorted view of reality



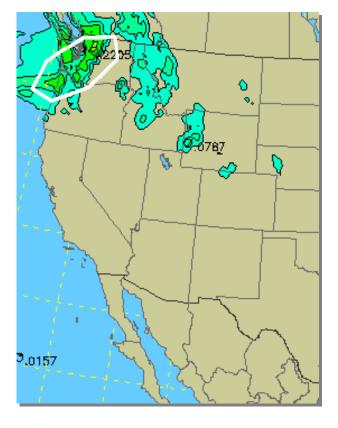
Forecast 2 – Another distorted view of reality



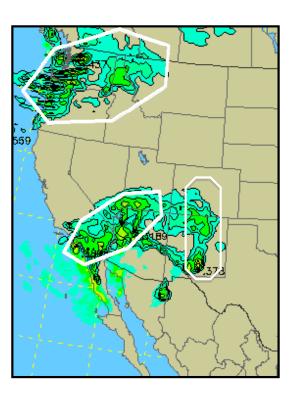
REAL - observed



Forecast 1 — Distorted view of reality



Forecast 2 – Another distorted view of reality



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This is not really a new idea . . .

Analytic cubists "analyzed" natural forms and reduced the forms into basic geometric parts on the two-dimensional picture plane.

Analytic cubism was developed between 1908 and 1912...

Comparing objects can tell you things about your forecast like . . .

This: Instead of this:

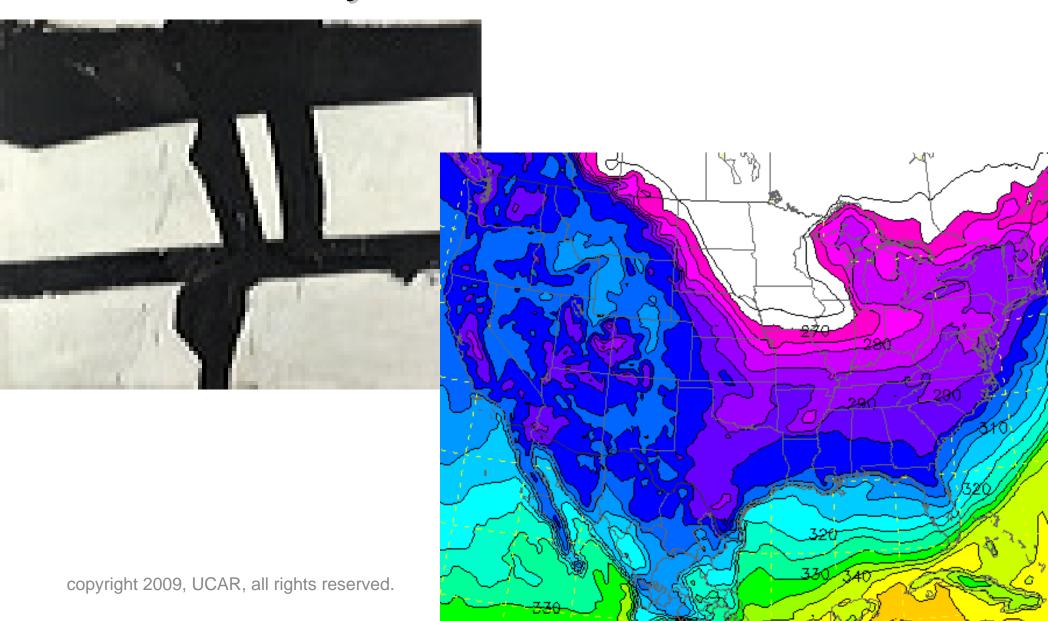
30% Too Big POD = 0.35

Shifted west 1 km FAR = 0.7235

Rotated 15° CSI = 0.1587

Peak Rain 1/2" too much

Verifying with objects doesn't always make sense...

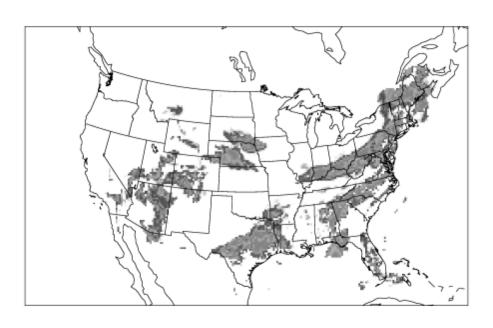


 In MET, object based verification is done using the MODE (Method for Object-Based Diagnostic Evaluation) tool.

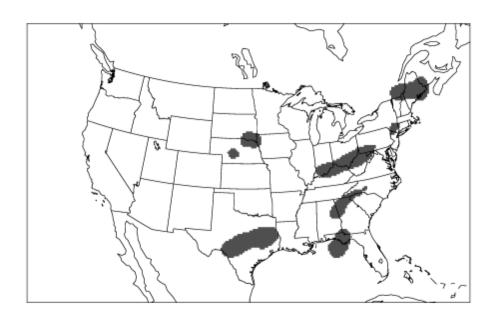
- Define objects
- Compute attributes (e.g. area, centroid, axis angle, intensity)
- Merge objects (e.g. thunderstorm cells merge into line)
- Match forecast and observed objects
- Compare attributes between matches
- Output summary statistics



What are Objects?

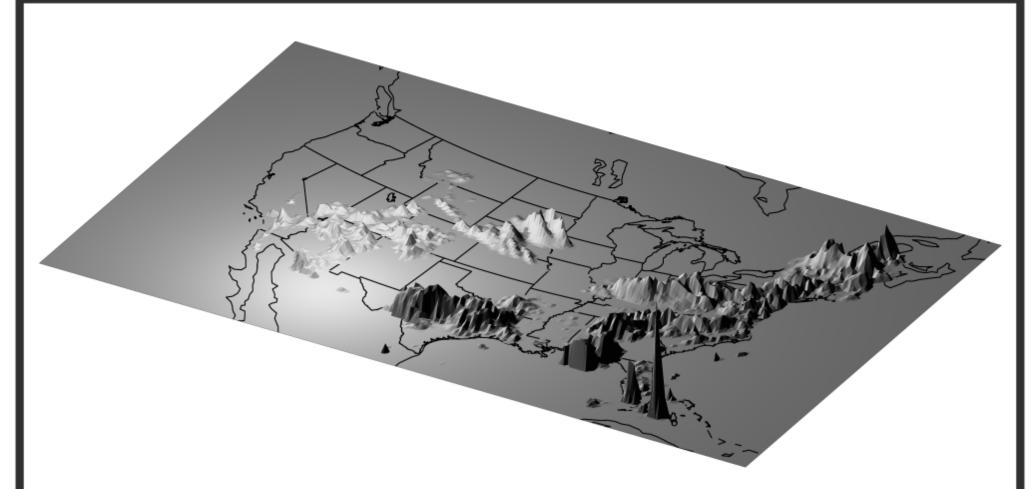


Raw Field



Object Field

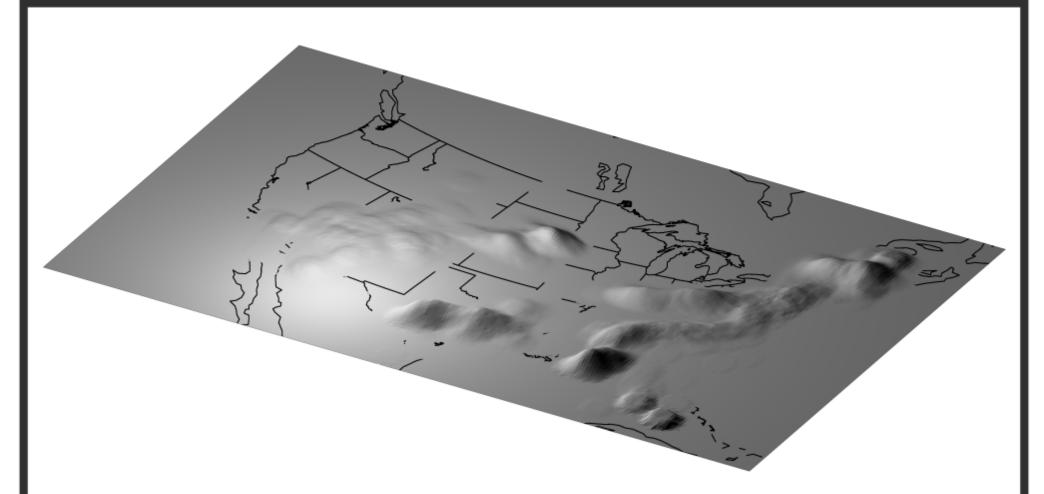
Objects are Regions of Interest



Step #1

Start with the raw data field.

In this case, a precipitation field.

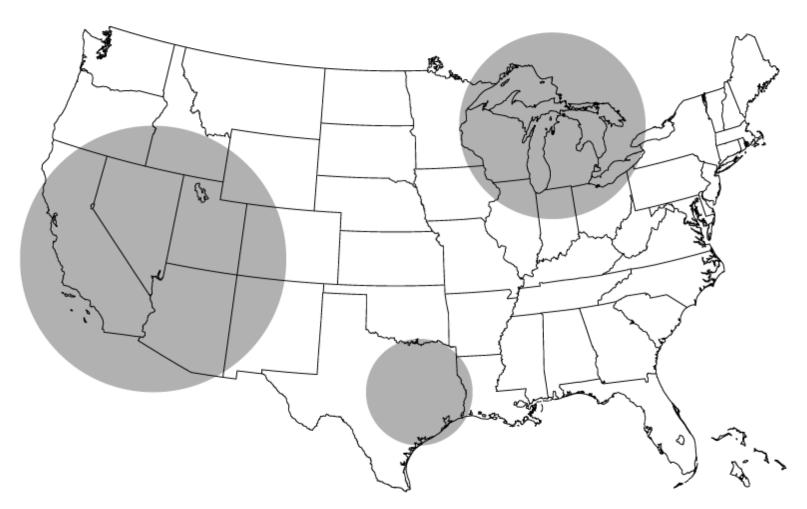


Step #2

Apply convolution operator.

This is basically a smoothing operation.

Convolution Radius



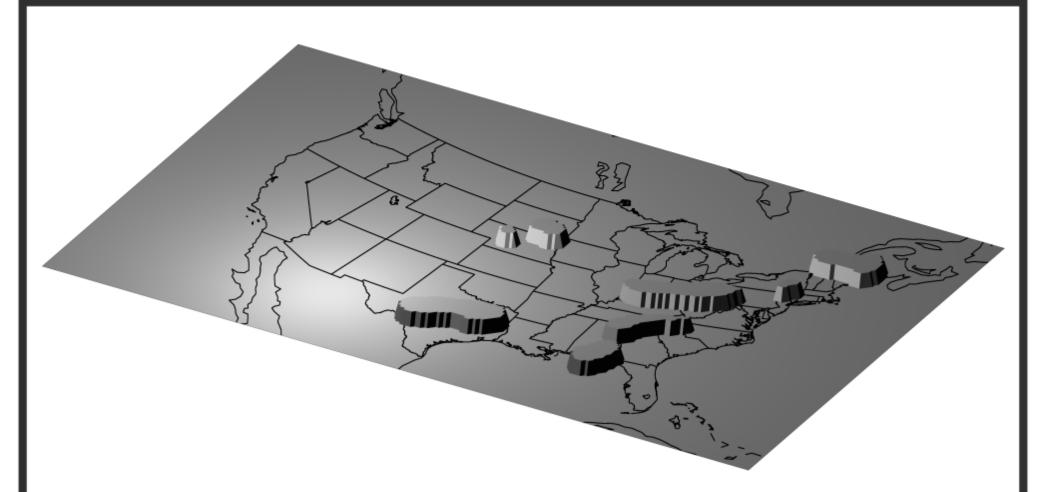
Radius of Influence

Convolution

Requires raw field f(x,y) and

filter function $\phi(x,y)$

$$C(x,y) = \sum_{(\hat{x},\hat{y})\in G} \phi(\hat{x},\hat{y}) f(x - \hat{x},y - \hat{y})$$



Step #3

Threshold the smoothed field.

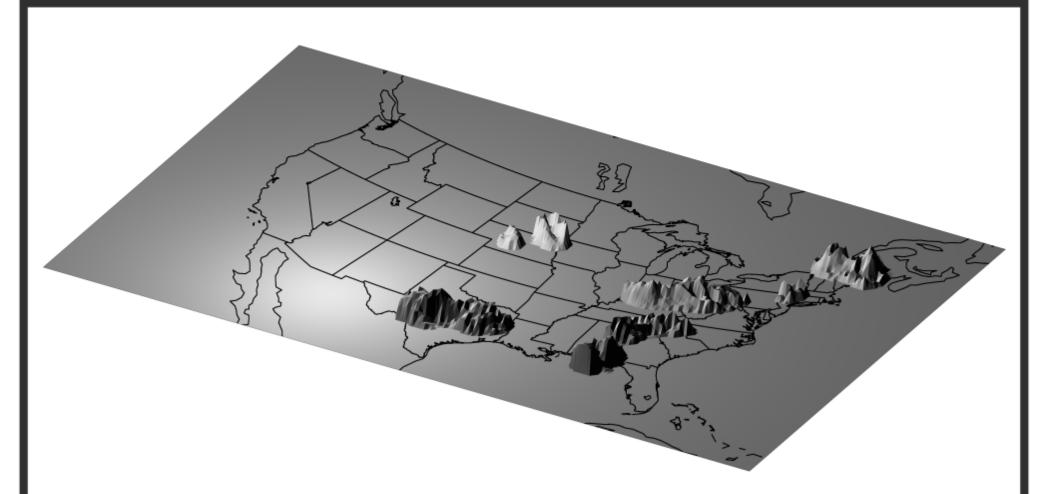
This produces an on/off mask field.

Masking

Uses convolved field C(x,y)

and threshold T

$$M(x,y) = \begin{cases} 1 & \text{if } C(x,y) \ge T \\ 0 & \text{else} \end{cases}$$

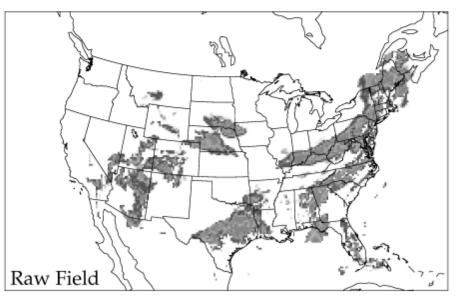


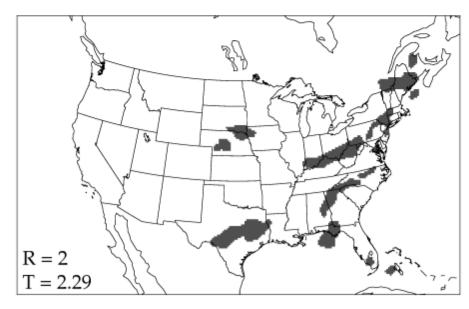
Step #4

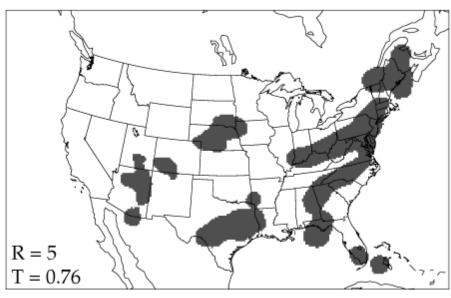
Restore original data to object interiors.

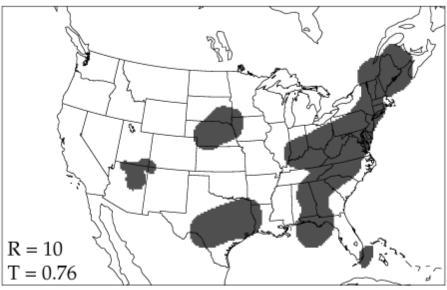
This gives us our objects.

Changing Object-Definition Parameters









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Matching & Merging

Merging:

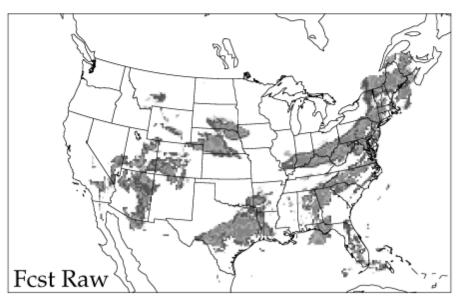
Associating objects in the same field.

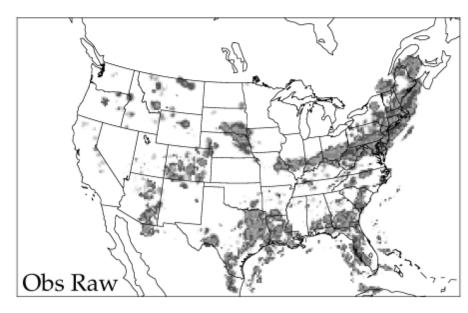
Matching:

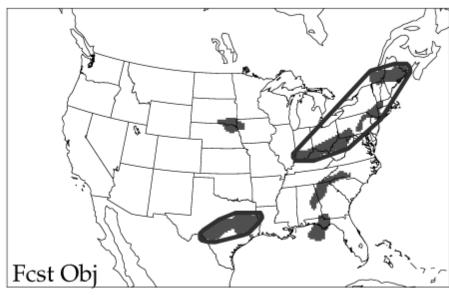
Associating objects in different fields.

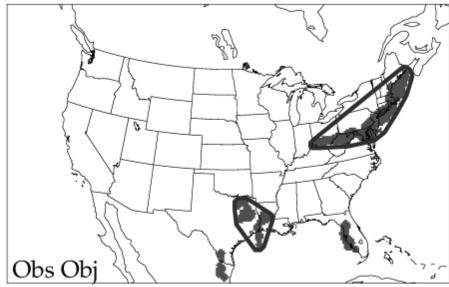
MODE does this using a Fuzzy-Logic engine.

Example of Matching & Merging









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Object Attributes

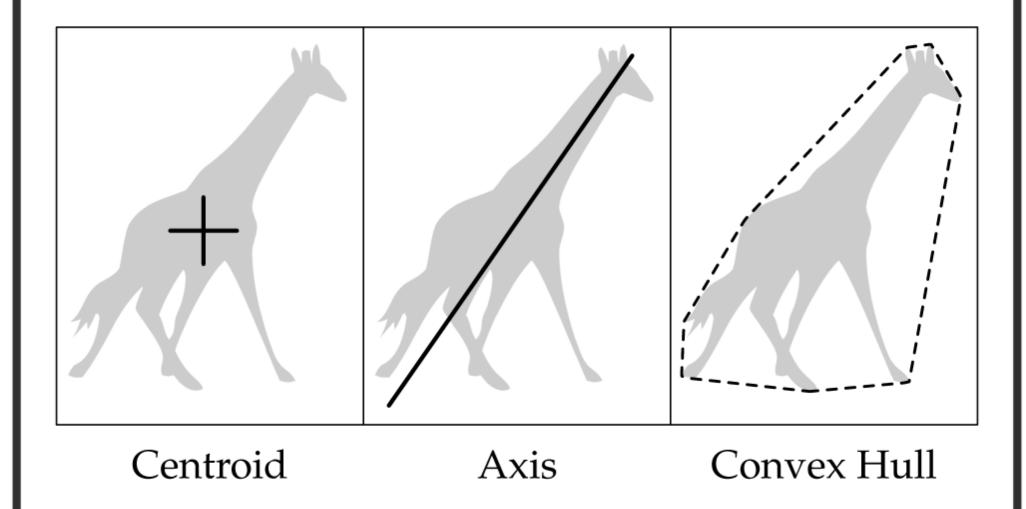
Single:

- Area
- Centroid
- Axis Angle
- Median Intensity
- Complexity
- Aspect Ratio
- Curvature

Pair:

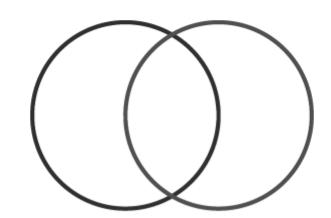
- Centroid Distance
- Angle Difference
- Median Intensity Ratio
- Intersection Area
- Convex Hull Distance
- Boundary Distance
- Area Ratio

Example Single Attributes

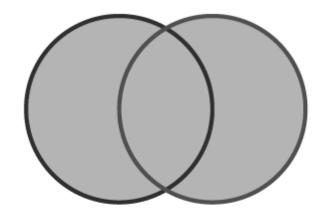


Example Pair Attributes

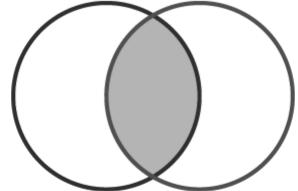
Forecast Object



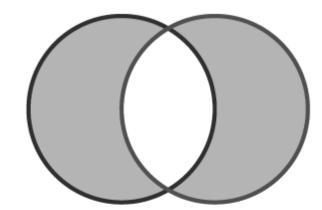
Observed Object



Union



Intersection



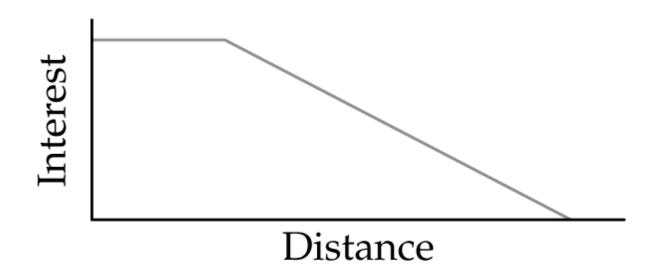
Symmetric Difference

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Interest Maps

Map attributes to interest values.

Example: Centroid Distance



All interest maps can be changed in the config file.

Weights

Express relative importance of different attributes in matching and merging.

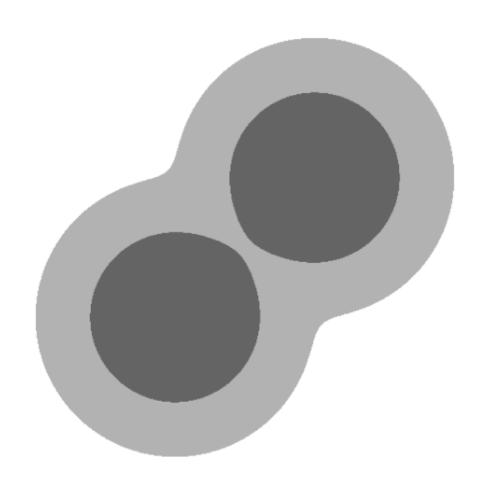
All weights can be changed in the config file.

Total Interest

Calculated from weights, attributes, and interest maps.

$$T(\alpha) = \frac{\sum_{i} w_{i} C_{i}(\alpha) I_{i}(\alpha)}{\sum_{i} w_{i} C_{i}(\alpha)}$$

Alternative Merging Method



Double Thresholding

Verifying WRF with Objects

Running MODE and MODE analysis tool

Running MODE - Input

- 2 gridded fields, same grid, possibly different variables
 - forecast (fcst_file) and observation (obs_file)



- configuration settings (config_file)
- syntax
 - >mode fcst_file obs_file config_file

Running MODE - Output

PostScript

- object pictures
- parameter summary
- total interest for each object pair

MODE: APCP_01 at SFC vs APCP_01 at SFC

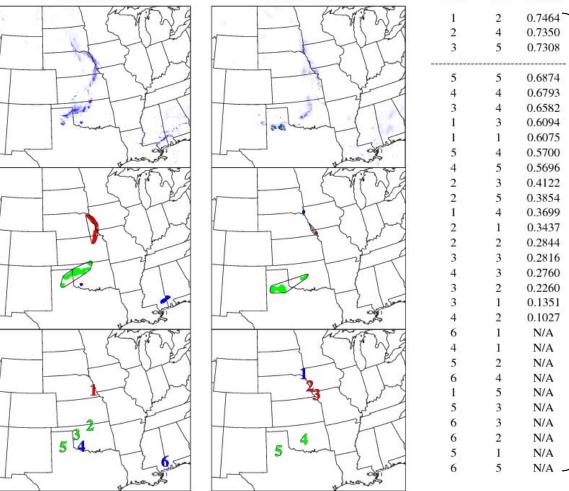
Observation

Fcst

Obs

Interest

Forecast



objectpairs total interest

model info name, times

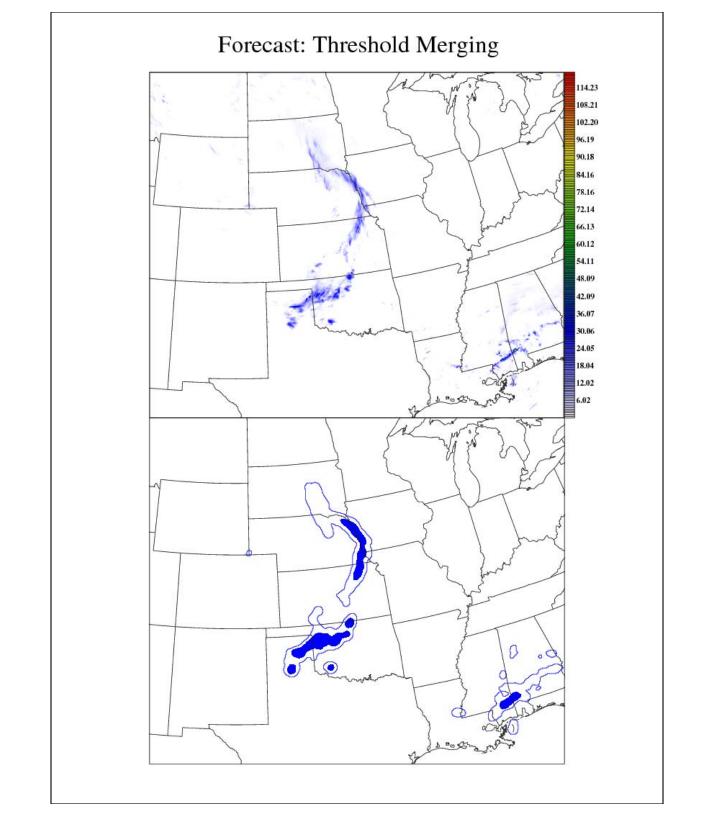
object pictures<

object-attribute weights

Model Name: wrf2caps Forecast Observation May 31, 2005 00:00:00 Mask Missing: off off Init Time: Jun 1, 2005 00:00:00 off/off Valid Time: Mask Gr/Ply: off/off Lead Time: 24:00:00 Raw Thresh: >=0.000 kg/m^2=0.000 kg/m^2 Accum Time: 01:00:00 Conv Radius: 7 gs Centroid Distance: 4.00 Conv Thresh: >=7.000 kg/m^2=7.000 kg/m^2 >=0.000 gsBoundary Distance: 6.00 Area Thresh: >=0.000 gs Convex Hull Distance: 0.00 Inten Perc: 100 th 100 th Angle Difference: 2.00 Inten Thresh: $>=0.000 \text{ kg/m}^2=0.000 \text{ kg/m}^2$ Merge Thresh: >=1.750 kg/m^2=1.750 kg/m^2 Area Ratio: 3.00 Intersection/Area: 2.00 Merging: thresh thresh 0.00 Matching: Complexity Ratio: match/merge 0.00 Simple(M/U): 6 (4/2) 5 (4/1) Intensity Ratio: 2 Total Interest Thresh: 0.70 Composites: 2

parameter
summary
threshold
smoothing radius
match/merging

Forecast Objects with Observation Outlines Observation Objects with Forecast Outlines



Running MODE - Output

PostScript

- object pictures
- parameter summary
- total interest for each object pair

ASCII

- object sizes, shapes, positions
- stats for simple, paired objects and clusters
- standard contingency table stats based on objects

netCDF

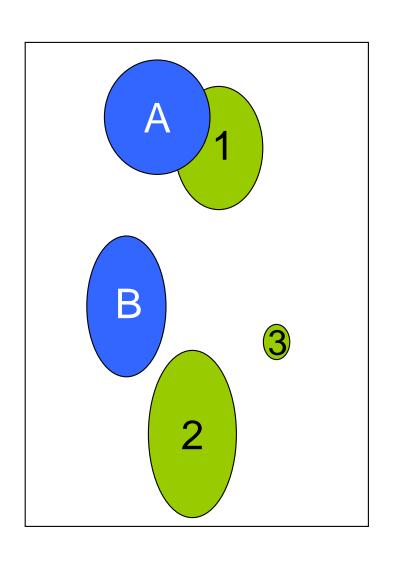
gridded object fields

Interpretation of MODE output

- How well do objects match?
 - median of the maximum interest (MMI)*
- What % of objects are matched?
- Displacement of matched objects
- mode_analysis tool

^{*} Davis et al., 2009: The Method for Object-based Diagnostic Evaluation (MODE) Applied to WRF Forecasts from the 2005 SPC Spring Program. Submitted to Weather and Forecasting

Median of the Max. Interest (MMI)

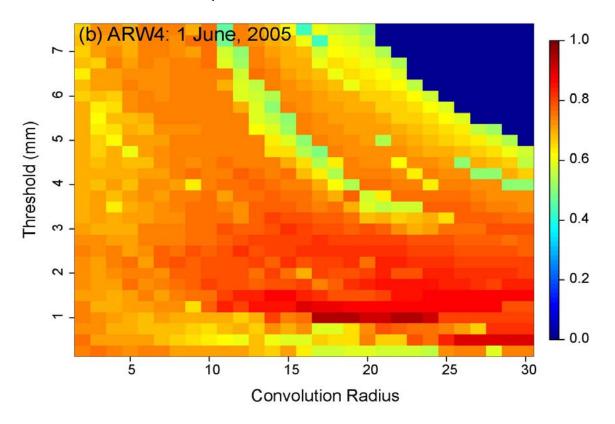


Interest Matrix observed

		A	В
forecast	1	0.90	0.65
	2	0.50	0.80
	3	0.40	0.55

 $MMI = median \{ 0.55, 0.80, 0.80, 0.90, 0.90 \} = 0.80$

Median of the Max. Interest (MMI) Quilt Plot



MMI as a function of convolution radius (grid squares) and threshold (mm) for 24-h forecast of 1-h rainfall

What % of Objects are Matched?

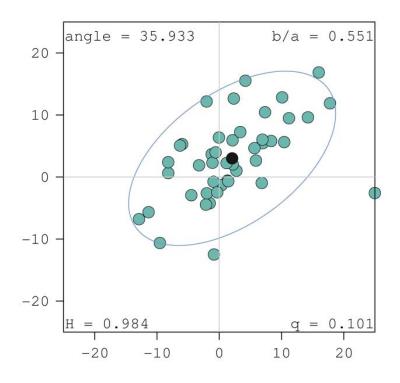
- Hit rate, Threat Score, Heidke Skill Score
- Other standard contingency table scores

	Forecasted object	Unforecasted object
Observed object	Matched	Missed
Unobserved object	False alarm	Correct null

MODE Analysis

 displacement of forecast clusters from matched observed clusters

> Composite Object Centroid Differences over the Appalachian Mountains



MODE Analysis Tool SUMMARY Example

Command Line

```
mode_analysis -summary
-mask_file ttt -config config/mode_test_config \
-dump_lines out -lookin /d1/score/mode_files \
-fcst -composite -area_min 3000 \
-centroid_x_min 600 \
-centroid_x_max 1100 \
-column CENTroid_x \
-column centroid_y \
-column centroid_lat \
-column centroid_lon \
-column area \
-column axis_ang \
-column length
```

Output

Total mode lines read = 73,330 Total mode lines kept = 539

Field	Min	Max	Mean	StdDev	P10	P25	P50	P75	P90
centroid_x	600.23	914.61	779.36	97.98	626.36	687.96	804.30	866.75	894.24
centroid_y	55.22	560.08	335.55	113.08	189.48	240.58	333.51	421.06	496.82
centroid_lat	22.00	40.14	32.08	4.06	26.83	28.66	32.00	35.15	37.87
centroid_lon	-107.03	-95.01	-100.18	3.75	-106.03	-103.67	-99.23	-96.84	-95.79
area	3210.00	85486.00	12680.96	9931.67	4935.00	6256.00	9445.00	16106.00	23678.00
axis_ang	-88.84	89.90	13.54	44.82	-55.80	-16.36	17.28	48.23	71.27
length	100.57	494.54	200.08	82.11	112.44	133.62	179.96	249.06	315.33

MODE Analysis Tool By Case Example

Command Line

```
mode_analysis -bycase
-dump_lines out
-mask_file ttt -config config/mode_test_config
-fcst_valid_min 20070702 -fcst_valid_max 20070702_12
-area_min 3000
-centroid_x_min 600 -centroid_x_max 1100
/d1/score/mode_files/ncwf2_vs_ncwdp/* /d1/score/mode_files/rcpf_vs_ncwdp/*
```

Output

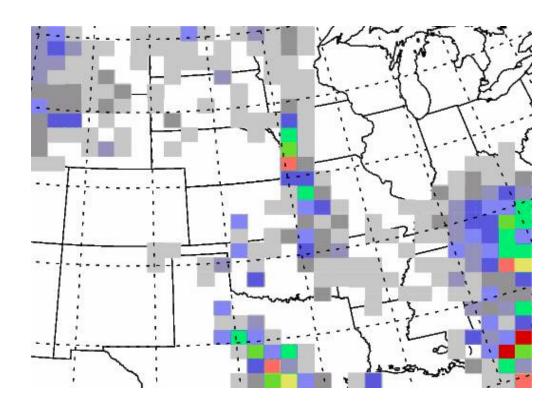
		Fcst V	alid Time	Area Matched	Area Unmatched	# Fcst Matched	# Fcst Unmatched	# Obs Matched	# Obs Unmatched
Jul	2.	2007	00:00:00	12392	20786	0	1	1	1
Jul		2007	01:00:00	6706	11038	0	0	1	2
Jul	2,	2007	02:00:00	7507	18696	0	0	1	3
Jul	2,	2007	03:00:00	19401	32268	2	3	1	2
Jul	2,	2007	04:00:00	0	16551	0	2	0	1
Jul	2,	2007	05:00:00	15311	29730	1	2	1	2
Jul	2,	2007	06:00:00	4730	8182	0	0	1	2
Jul	2,	2007	07:00:00	3733	13285	0	1	1	2
Jul	2,	2007	08:00:00	6994	6994	0	0	1	1
Jul	2,	2007	09:00:00	15981	15981	0	0	2	2
Jul	2,	2007	10:00:00	51501	53427	2	2	4	4
Jul	2,	2007	11:00:00	15779	21089	1	1	1	2
Jul	2,	2007	12:00:00	31339	40665	1	2	2	2

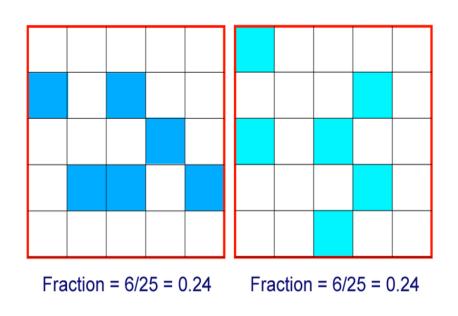
Thurs AM Lecture Wrap-Up

- Confidence intervals
- Verifying WRF with objects
 - matching and merging
- Running MODE
 - MODE analysis
- Questions?

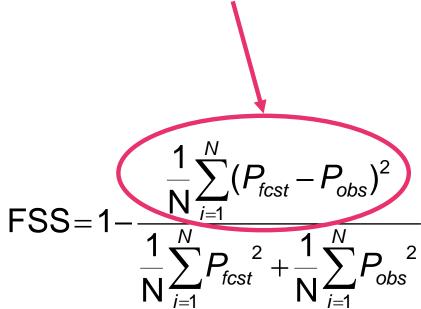


- Smoothing (e.g., FSS, Upscaling)
 - Neighborhood or "fuzzy" (Ebert 2008)
 - At what resolution does the model have skill?
 - Credit a forecast for being partially correct.
- Band-pass (e.g., Fourier, Wavelets)
 - How well does a model reproduce the spatial structure of the verification field?
 - Examine performance at different scales separately.





Numerator is Fractions Brier Score (FBS)



Roberts and Lean (2008)

Fractions Skill Score (FSS)

Slide from E Ebert. See Ebert (2008) for full references.

Fuzzy method	Matching strategy*	Decision model for useful forecast
Upscaling (Zepeda-Arce et al. 2000; Weygandt et al. 2004)	NO-NF	Resembles obs when averaged to coarser scales
Minimum coverage (Damrath 2004)	NO-NF	Predicts event over minimum fraction of region
Fuzzy logic (Damrath 2004), joint probability (Ebert 2002)	NO-NF	More correct than incorrect
Fractions skill score (Roberts and Lean 2007)	NO-NF	Similar frequency of forecast and observed events
Area-related RMSE (Rezacova et al. 2006)	NO-NF	Similar intensity distribution as observed
Practically perfect hindcast (Brooks et al. 1998)	NO-NF	Resembles a forecast based on perfect knowledge of observations
Pragmatic (Theis et al. 2005)	SO-NF	Can distinguish events and non-events
CSRR (Germann and Zawadzki 2004)	SO-NF	High probability of matching observed value
Multi-event contingency table (Atger 2001)	SO-NF	Predicts at least one event close to observed event

*NO-NF = neighborhood observation-neighborhood forecast, SO-NF = single observation-neighborhood forecast

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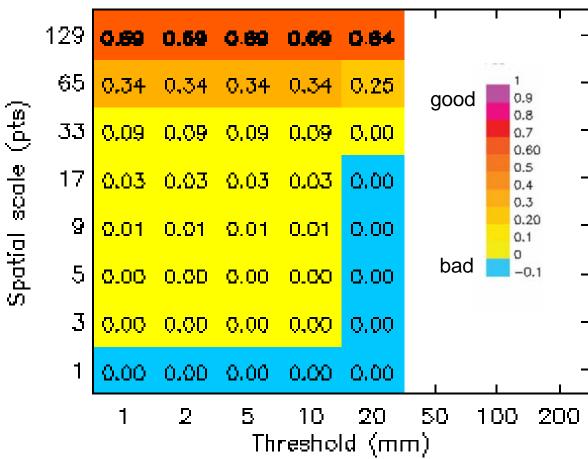
12.7 mm < 25.4 mm <

forecast

Filter Methods

Fig. courtesy of E Ebert.

Fractions skill score — FSS



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- In MET, Neighborhood methods are in grid-Stat tool. See MET documentation for specific filters.
- Can be cumbersome (many numbers for one filter).
- Choose filter that addresses the verification question of interest.
- See Ebert (2008) for a good summary and comparison of these techniques (and references).

Fourier

$$Z(\mathbf{x}) = \sum_{k=1}^{n} Z_k \exp(i\omega_k^T \mathbf{x})$$

Wavelets

$$Z(\mathbf{x}) = \sum_{j,k} c_{j,k} \varphi_k \left(2^{j-1} \mathbf{x} \right) + \sum_{j,k} d_{j,k} \psi_k \left(2^{j-1} \mathbf{x} \right)$$

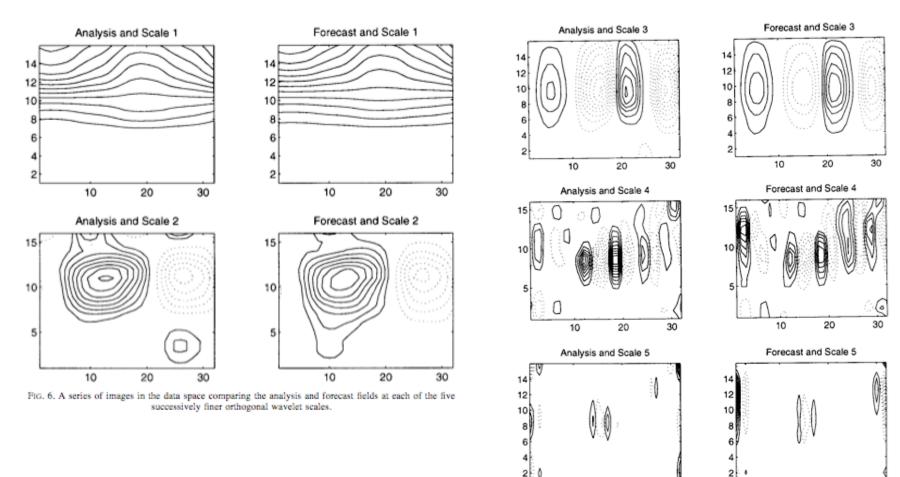
Fourier

- Skamarock (2004), MWR **132**:3019-3032
- Harris et al. (2001), J Hydrometeorol. 2:406-418
- Tustison et al. (2001), JGR 106(D11):11775-11784
- and many more...

Wavelet

- Briggs and Levine (1997), MWR 125:1329-1341
- Casati et al. (2004). [In MET Wavelet tool]

Wavelets



Briggs and Levine (1997)

10

20

30

Fig. 6. (Continued)

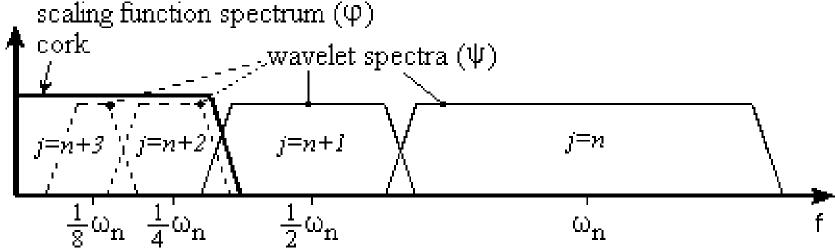
20

10

30

Wavelets

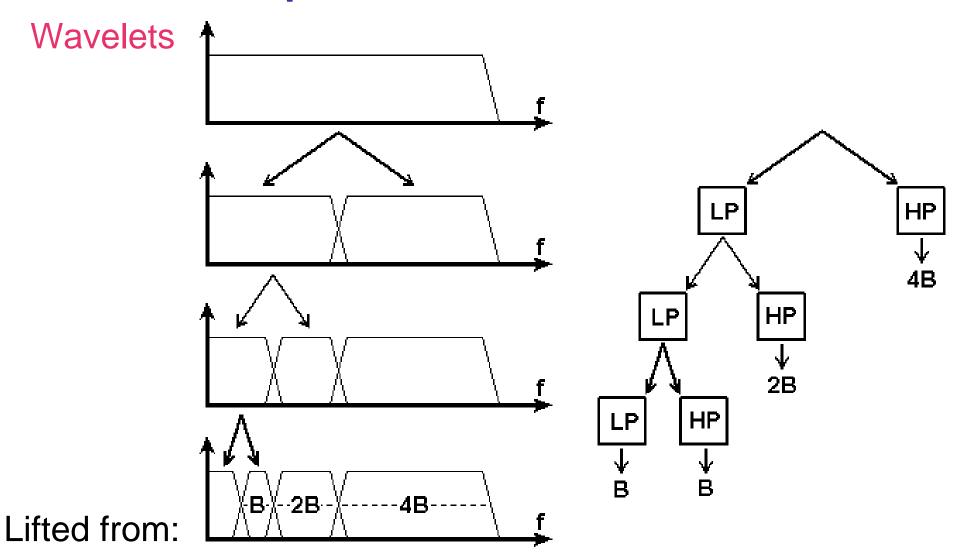
$$Z(\mathbf{x}) = \sum_{j,k} c_{j,k} \varphi_k \left(2^{j-1} \mathbf{x} \right) + \sum_{j,k} d_{j,k} \psi_k \left(2^{j-1} \mathbf{x} \right)$$



Lifted from:

http://pagesperso-orange.fr/polyvalens/clemens/wavelets/

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http://pagesperso-orange.fr/polyvalens/clemens/wavelets/

Wavelets: Practical considerations

- Applicable only on square grids of size, 2ⁿ.
- If grid does not meet this, then must alter the grid in some way. For example,
 - Pad with zeros,
 - Cut the domain down.
- Several choices for wavelets. Haar Wavelets are a good choice for precipitation (because of their discrete nature), but other choices might be better for other fields.

Quick Contingency Table Review

	obs > threshold	obs < threshold	
forecast > threshold	a hits	b false alarms	a+b
forecast > threshold	c misses	d correct rejections	c+d
	. , ,	b+d UCAR, all rights rved.	n=a+b+c

Intensity Scale (IS) (Casati et al., 2004)

- Create binary fields for a threshold, and calculate MSE (denote, MSE(t)).
- 2. Apply wavelet decomposition to binary fields.
- Calculate MSE for each threshold (t) and scale (j) (denote, MSE(t,j)), and find MSE%(t,j)=MSE(t,j)/MSE(t).
- 4. Find MSE(t)_{random} = FBI*Br*(1-Br) + Br*(1-FBI*Br), where FBI=(a+b)/(a+c) is the frequency bias index and Br=(a+c)/n is the sample climatology from the contingency table.
- 5. Calculate $SS(t,j)=1-MSE(t,j)*(n+1)/MSE(t)_{random}$.

In addition to MSE and SS, the energy squared (denote En2) is calculated. That is,

$$En2(X) = \sum_{i=1}^{n} X_i^2$$

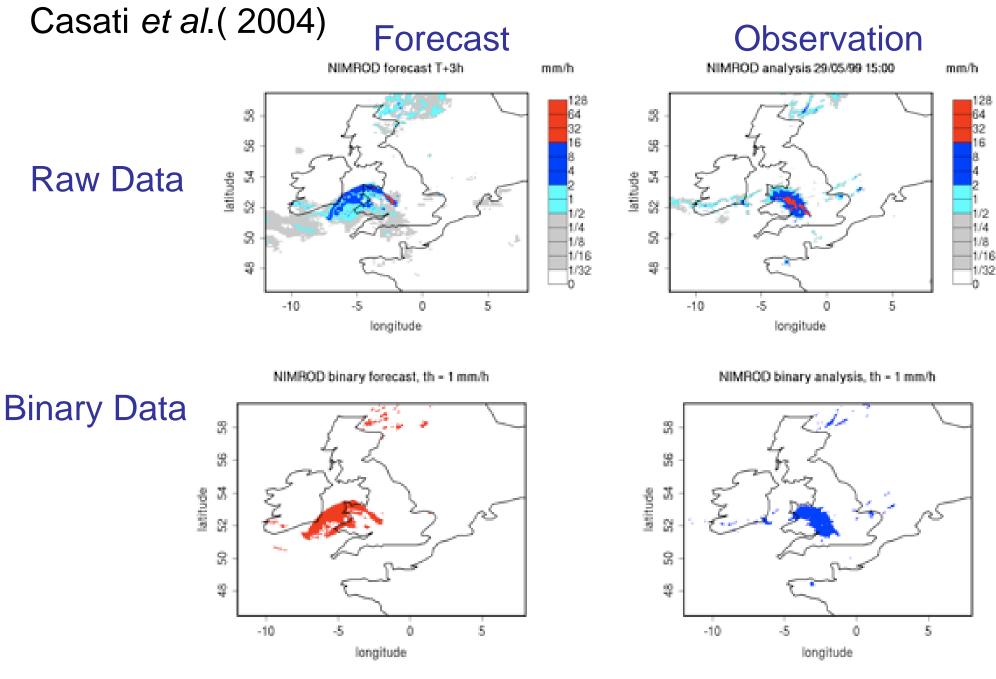
Also find the relative difference in En2 between the forecast and verification fields. That is,

[En2(F)-En2(O)]/[En2(F)+En2(O)]

Similar to MSE, let En2(X,t) be the energy for the raw binary field X (where X is either the verification or forecast field), and En2(X,t,j) is the same, but taken only at scale j. Then

En2%(X,t,j) = En2(X,t,j)/En2(X,t).

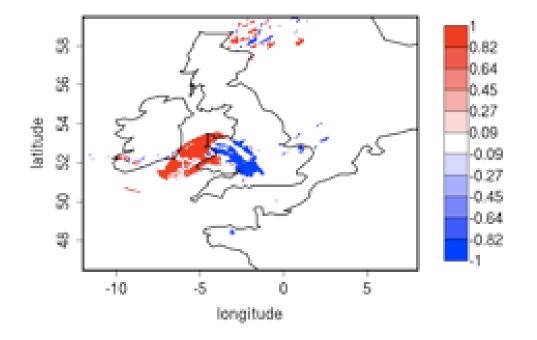
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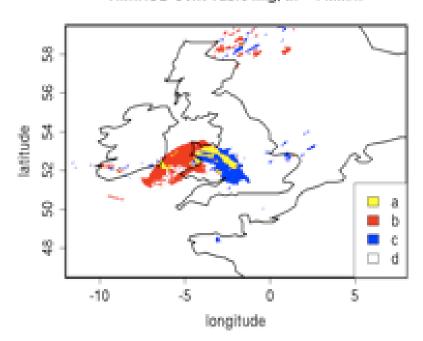
Binary Difference Field

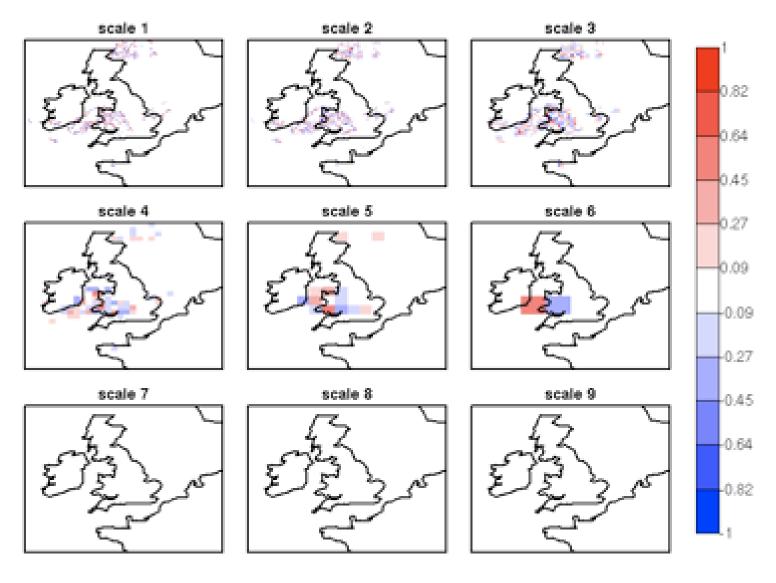
NIMROD binary field difference, th - 1 mm/h



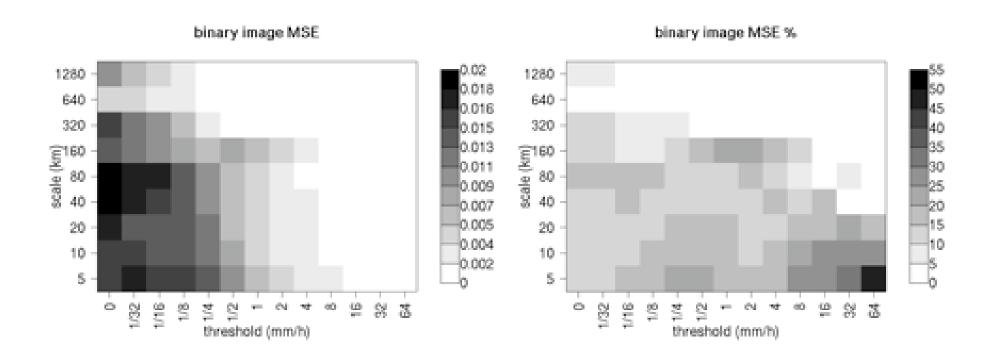
Contingency Table Contributions

NIMROD Cont Table Img, th - 1 mm/h



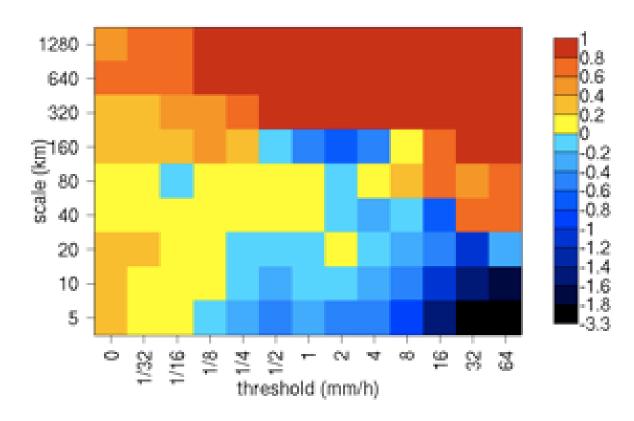


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Intensity Skill Score (ISS) is the Heidke Skill Score (HSS), but calculated at each threshold and wave number.

Intensity-Scale skill score



HSS is an accuracy score corrected for random chance.

Thank you...Questions?

References

Casati B, G Ross, and DB Stephenson, 2004. A new intensity-scale approach for the verification of spatial precipitation forecasts. *Meteorol. Appl.* **11**:141--154.

Ebert EE, 2008. Fuzzy verification of high resolution gridded forecasts: A review and proposed framework. *Meteorol. Appl.*, **15**:51--64. DOI: 10.1002/met.25.

See also, http://www.ral.ucar.edu/projects/icp

Customizing and Understanding Wavelet Stats

Tressa L. Fowler

Options for Handling Missing data

Points with valid forecasts.



Points with valid observations.



- Points with both valid forecasts and observations.
- All points. $F \cup C$

Thresholds

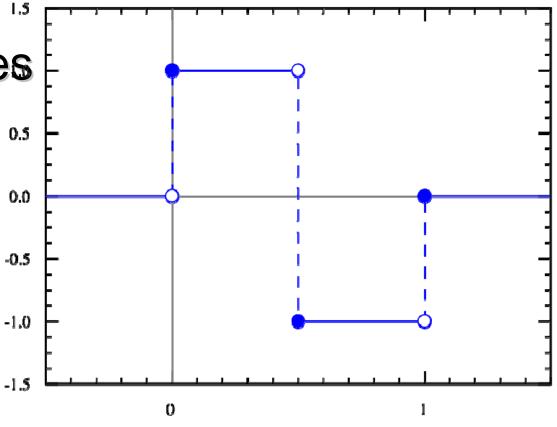
Forecast		Threshold	Event
0.05	<	0.2	0
0.17	<	0.2	0
0.45	>	0.2	1
2.15	>	0.2	1
0.05	<	1	0
0.17	<	1	0
0.45	<	1	0
2.15	>	1	1

Tiles – Grid must be 2ⁿ x 2ⁿ

- Cut down user selected subset (square)
- Tiles automated selection of subset(s)
- Pad with zeros not recommended unless adds very small number of points.

Wavelets

- Haar
- Centered Haar
- Daubechies
- Centered Daubechies
- B spline
- Centered B spline



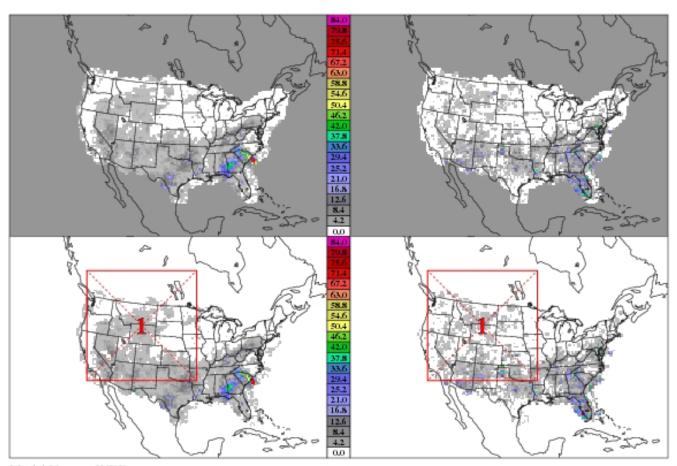
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Output

- Text files
 - Options
 - Statistics
 - Can be read into stat analysis tool.
- Postscript files
 - Graphics
 - Options
 - Statistics

Wavelet-Stat: APCP at A24

Forecast Observation



Model Name: WRF

Init Time: Aug 7, 2005 00:00:00 Valid Time: Aug 8, 2005 00:00:00

Lead Time: 24:00:00 Accum Time: 24:00:00

Tile Method: User-Defined

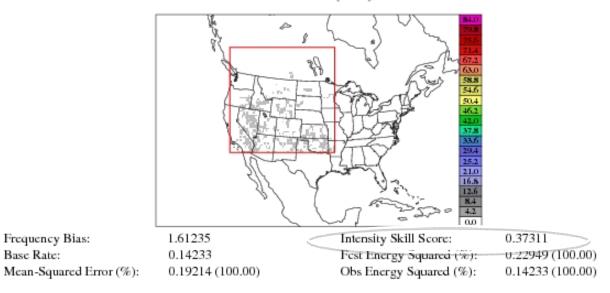
Tile Count: 1
Tile Dim: 64 x 64
Tile Corner: (45, 45)

Mask Missing: Fcst/Obs Wavelet(k): Haar (2)

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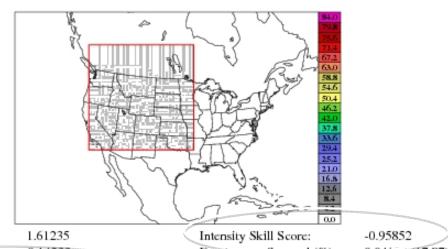
Wavelet-Stat: APCP at A24, Tile 1, >1.000, Binary

Difference (F-0)



Wavelet-Stat: APCP at A24, Tile 1, >1.000, Scale 1

Difference (F-0)



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Frequency Bias: 1.61235

Backate: 0.14233

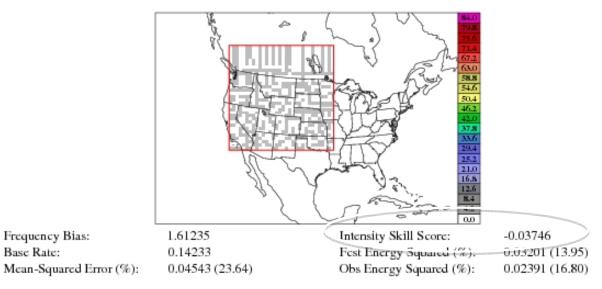
Mean-Squared Error (%): 0.08575 (44.63)

Intensity Skill Score: -0.958
Fest Energy Squared (%): 0.0410
Obs Energy Squared (%): 0.0517

0.04102 (17.87) 0.05170 (36.32)

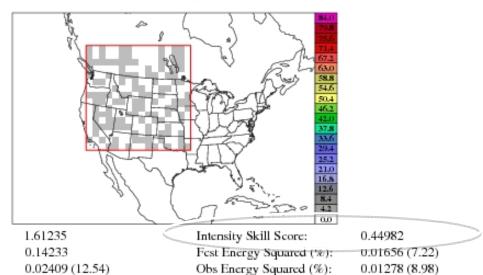
Wavelet-Stat: APCP at A24, Tile 1, >1.000, Scale 2

Difference (F-0)



Wavelet-Stat: APCP at A24, Tile 1, >1.000, Scale 3

Difference (F-0)



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Frequency Bias:

Mean-Squared Error (%):

Base Rate:

Wavelet-Stat: APCP at A24, Tile 1, >1.000, Scale 4

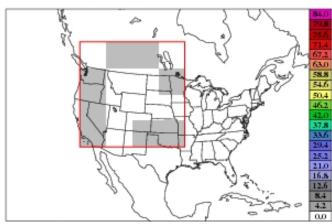
Difference (F-0)



Frequency Bias: Base Rate: Mean-Squared Error (%): 1.61235 0.14233 0.01313 (6.83) Intensity Skill Score: Fest Energy Squared (%): Obs Energy Squared (%): 0.70019 0.02332 (10.16) 0.00954 (6.70)

Wavelet-Stat: APCP at A24, Tile 1, >1.000, Scale 5

Difference (F-0)



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Frequency Bias: Base Rate: Mean-Squared Error (%): 1.61235 0.14233 0.00553 (2.88) Intensity Skill Score: Fest Energy Squared (%): Obs Energy Squared (%): 0.87362 0.02745 (11.96) 0.01600 (11.24)

Wavelet-Stat: APCP at A24, Tile 1, >1.000, Scale 6

Difference (F-0)



Frequency Bias: Base Rate: Mean-Squared Error (%): 1.61235 0.14233 0.01061 (5.52) Intensity Skill Score: Fest Energy Squared (%): Obs Energy Squared (%): 0.75764 0.03646 (15.89) 0.00815 (5.72)

Wavelet-Stat: APCP at A24, Tile 1, >1.000, Scale 7

Difference (F-0)



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Frequency Bias: Base Rate: Mean-Squared Error (%): 1.61235 0.14233 0.00760 (3.95) Intensity Skill Score: Fest Energy Squared (%): Obs Energy Squared (%): 0.82650 0.05267 (22.95) 0.02026 (14.23)

Summary

- Wavelet tool provides a flexible method for decomposing spatial fields into different scales.
- Once decomposed, verification measures at each *physical* scale can be examined and compared.