

Building Python-based operational systems for prediction of atmospheric processes

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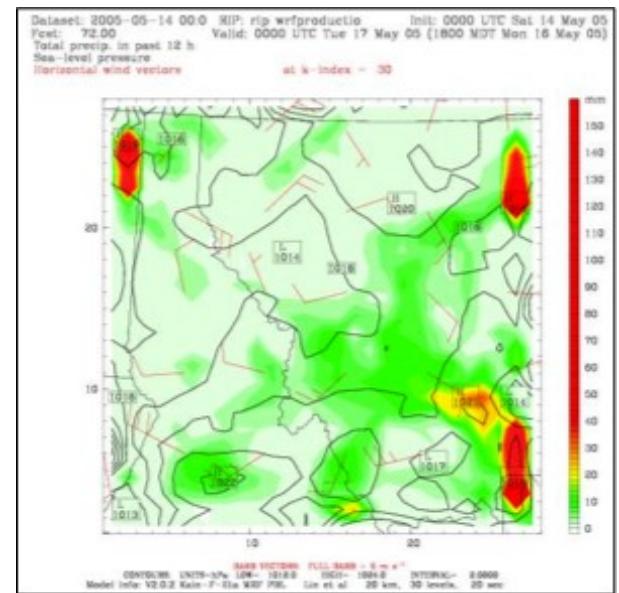
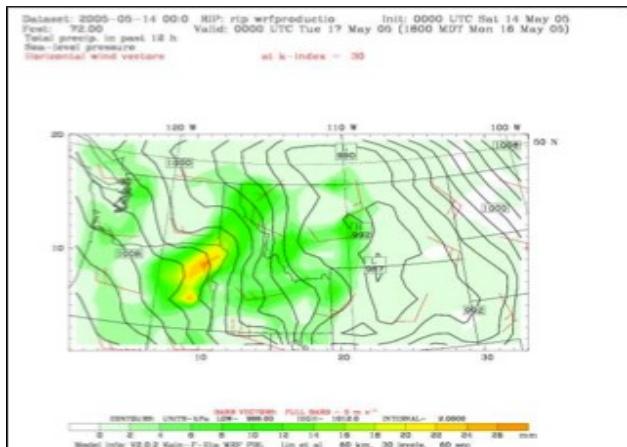
Outline

- Evolution and overview of operational activities
- Python – the glue that has held it together and enabled deployment of complex, evolving systems
- Some examples
- Ambitions



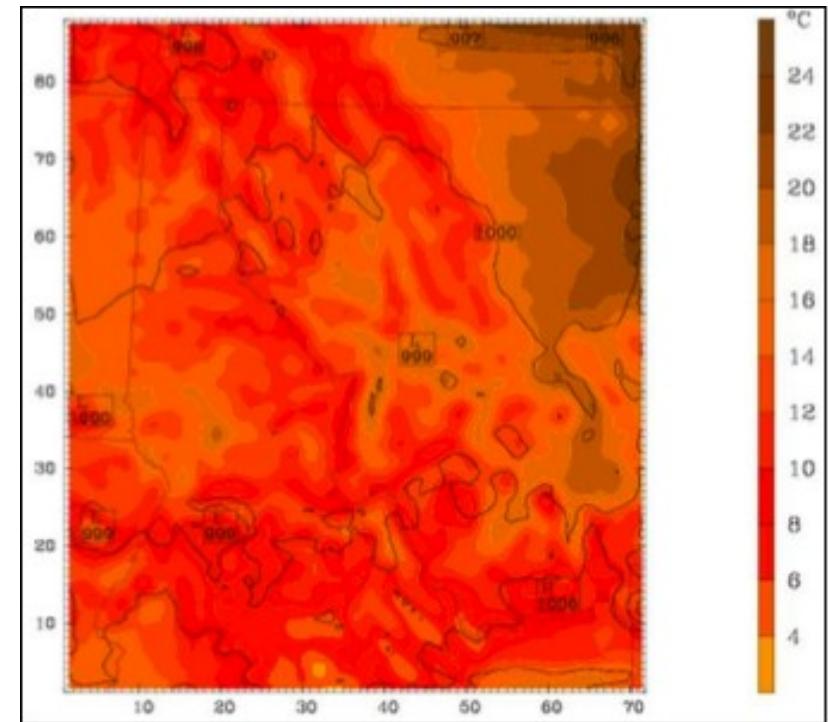
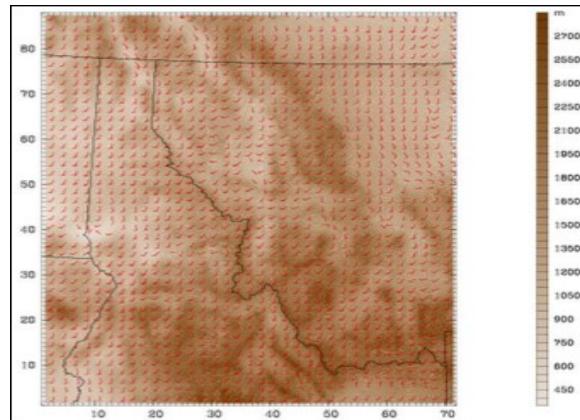
Beginnings of operational NWP

- Early 2004 – MM5 prototyping
- After first WRF Tutorial, June 2004
- Daily 60km with 20km nest
- Performed on home computer – Linux VM with guest PGI compiler running on Windows XP 1 GHz, *bash*-driven
- 72 hour forecast took 7.5 hours
- Moved to spare Linux machine at U. Montana



2004 – Missoula WFO forecasts

- Single Linux machine
- Evolved to Gene Petrescu Memorial Supercomputer – 2-machine, 4-cpu Linux cluster
- *bash*, local web posting, one forecast per day
- Daily 7.5km (73x86x75 grid)
- 72-hour forecast in 12 hours

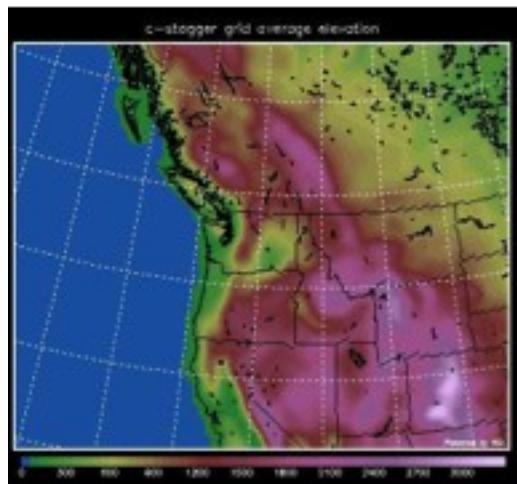
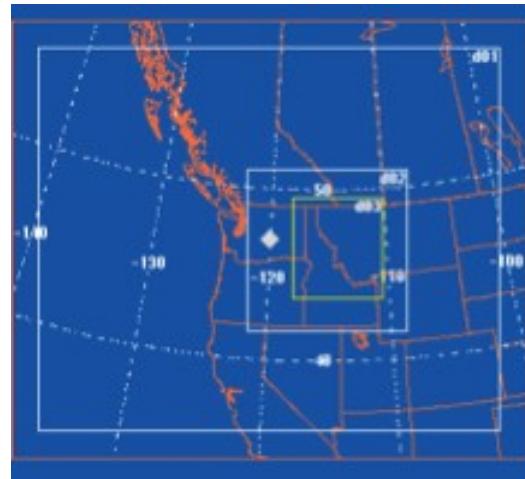


Cray XD1

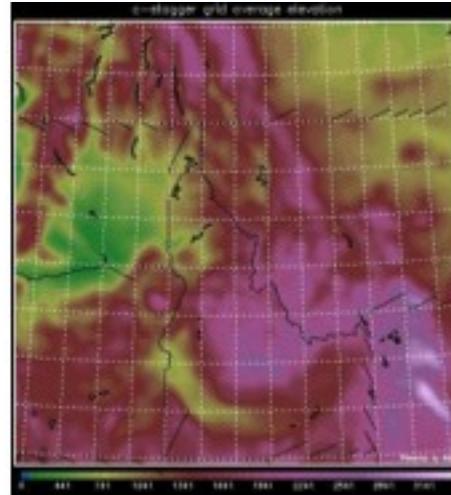
- Missoula USFS purchased small XD1, housed at NWS with usage agreement
- Started getting complicated – moved from bash to Python
- SGE queueing system
- People wanted multiple runs per day
- People wanted AWIPS
- Automated verifications
- Still a monolithic and linear structure



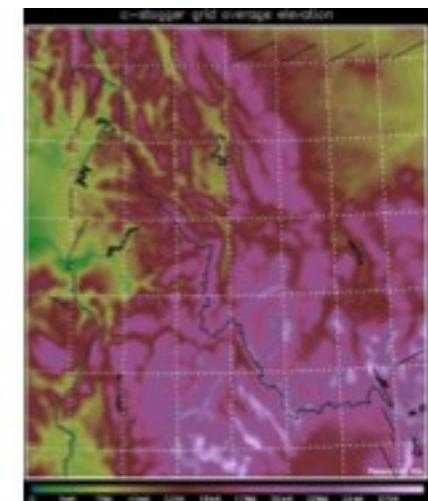
XD1 MSO Production Domains



36 km



12 km



4 km



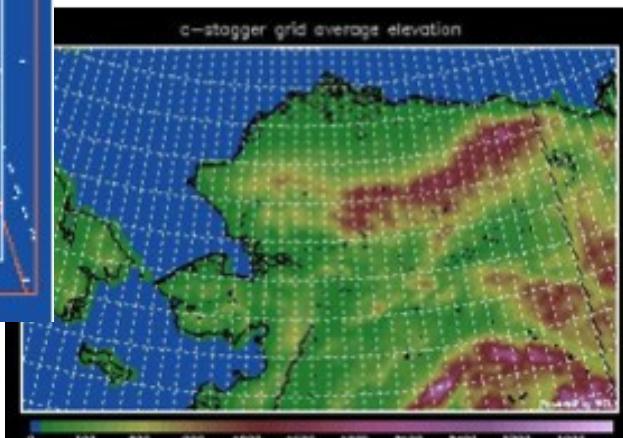
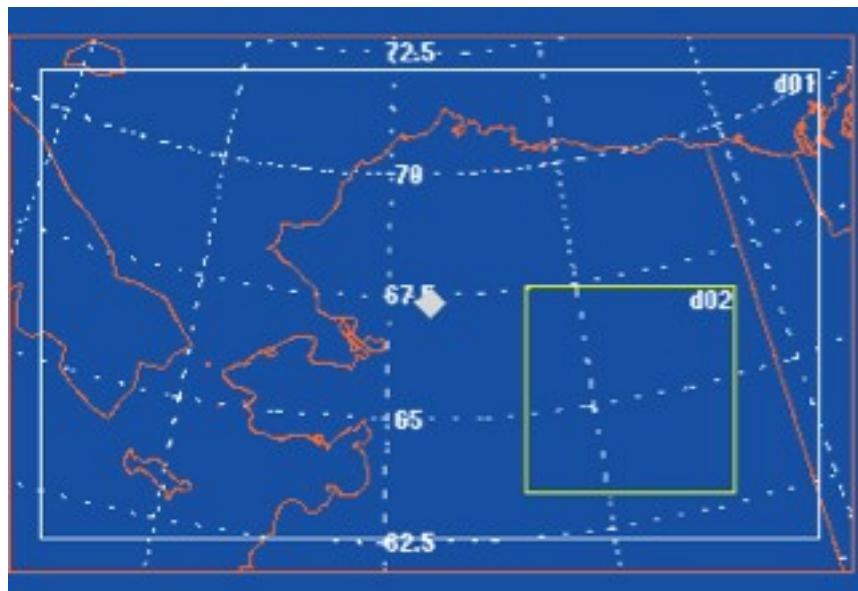
2005 – Alaska forecasts

- ARSC had acquired an XD1 and was interested (along with Cray) in these forecasts as showcase application
- Early summer - XD1 wasn't ready yet, so deployed first on IBM p655 cluster (XL compilers, LoadLeveler queueing system), with intent to move to XD1 when ready
- Late summer - Moved to XD1, more ambitious runs, more I/O issues

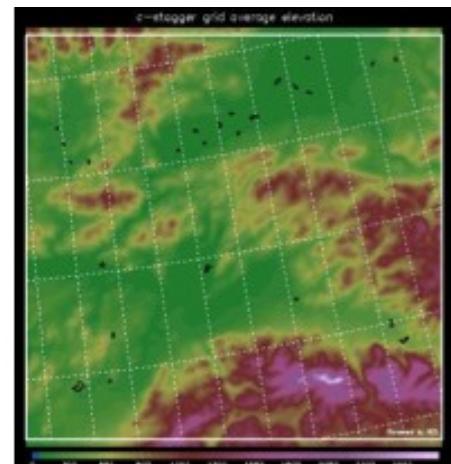


2005 – Alaska Forecasts

- At the time, focused on Northern Alaska forecast area

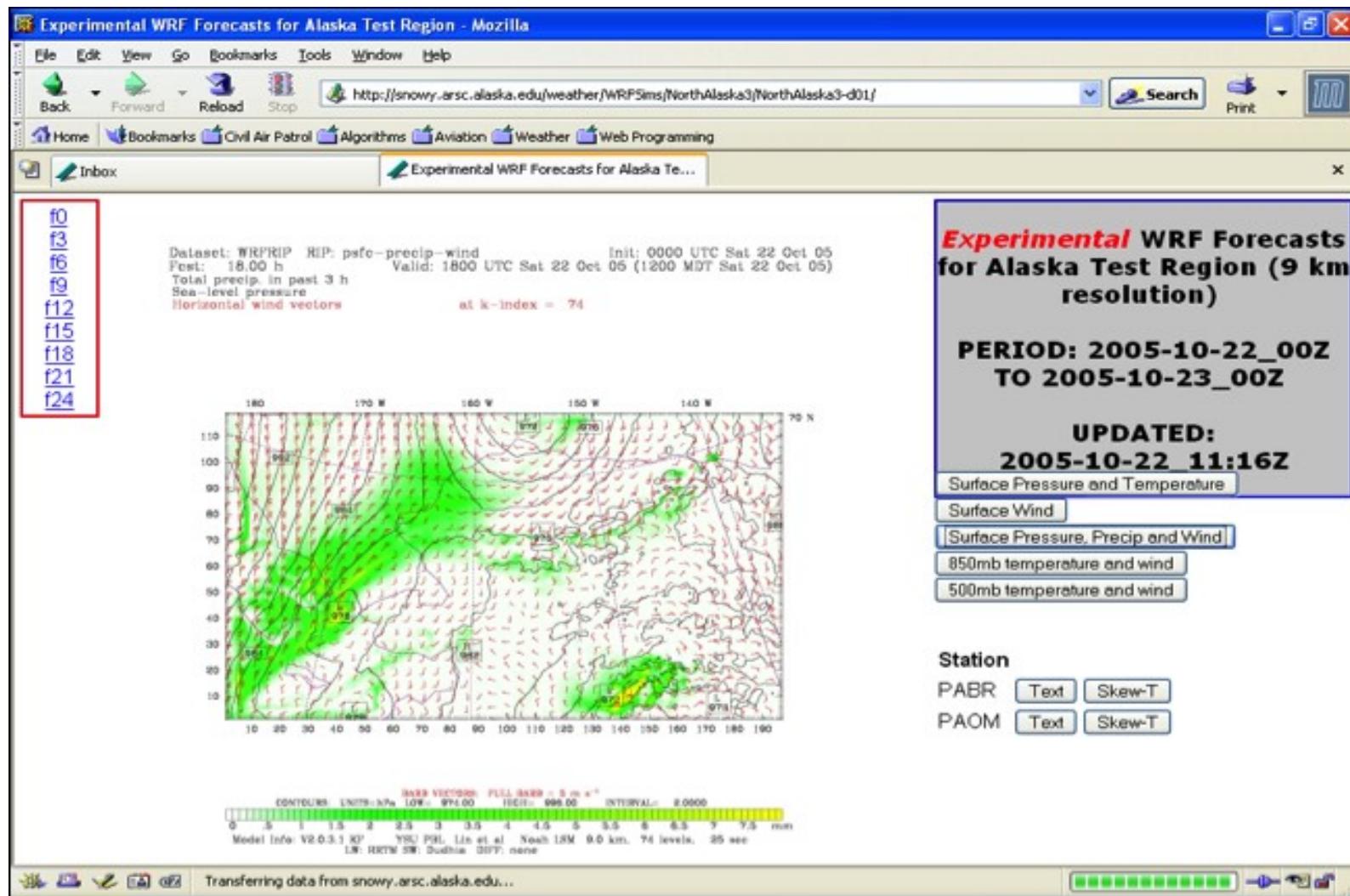


9 km



3 km

Alaska Products



Alaska Products

Experimental WRF Forecasts for Alaska Test Region - Mozilla

File Edit View Go Bookmarks Tools Window Help

Back Forward Reload Stop

http://snowy.arsc.alaska.edu/weather/WRFsim/NorthAlaska3/NorthAlaska3-d02/ Search Print

Home Bookmarks Civil Air Patrol Algorithms Aviation Weather Web Programming

Inbox

PABI

Dataset: f0 f3 f6 f9 f12 f15 f18 f21 f24

Timestamp StaPSFC StaT2 StaWindSpd StaWindDir StaRain

Timestamp	StaPSFC	StaT2	StaWindSpd	StaWindDir	StaRain
2005102200					
2005102203	939.56	3.85	4.64	118.45	0.00
2005102206	939.90	3.63	5.51	111.08	0.00
2005102209	937.77	5.36	8.58	144.39	0.00
2005102212	939.11	5.05	11.81	156.60	0.00
2005102215	937.70	1.64	5.42	106.77	0.00
2005102218	934.35	0.94	14.53	98.26	0.00
2005102221	930.22	5.77	12.52	102.50	0.00

Done

2005-10-22 11:17Z

Surface Pressure and Temperature

Surface Wind

Surface Pressure, Precip and Wind

850mb temperature and wind

500mb temperature and wind

Station

PABI Text Skew-T

PAFA Text Skew-T

PANN Text Skew-T

BARB VECTORS: FULL BARB = 5 m s⁻¹

Model Info: V2.0.3.1 KP YSU PBL Lin et al. Noah LSM 3.0 km. 74 levels. 8 sec

IN: IERTM SW: Dudhia DUFF: none

start Experimental WRF ... 2/639/10:27/dorm... http://snowy.arsc.... Windows Explorer Microsoft Power... 10:36 AM

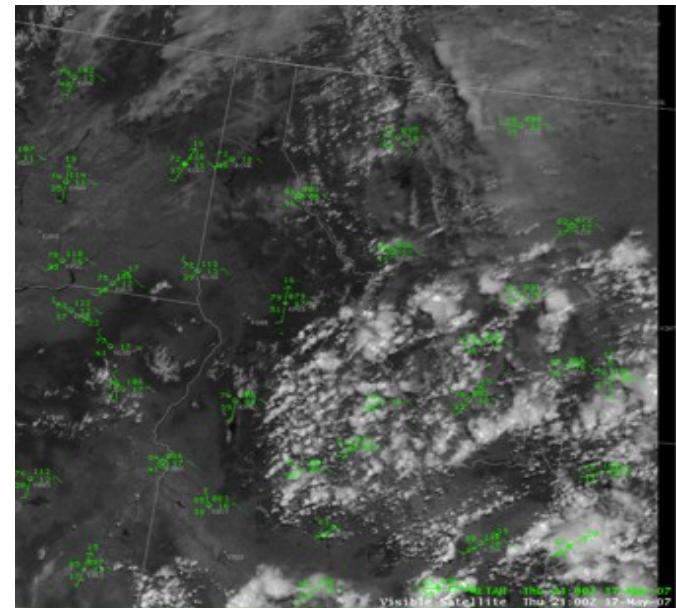
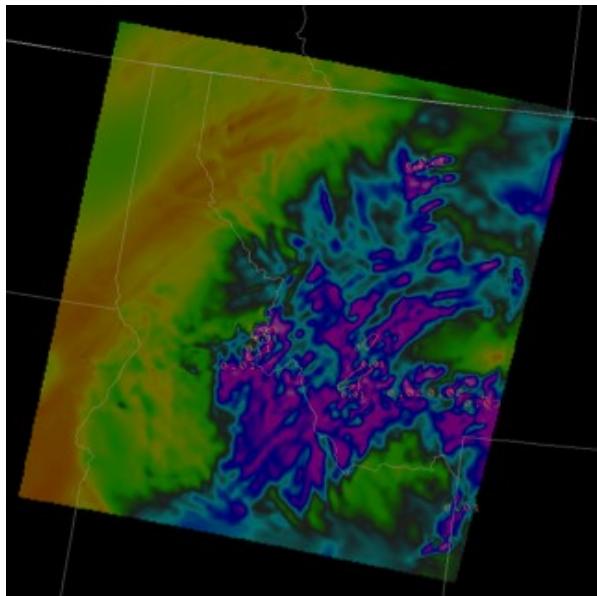
Continued MSO and FAI runs

- MSO runs in high-demand
- NWS wanted more timely runs – e.g. 00Z forecast products starting to come in around 00Z!!
- Multiple runs per day, which meant overlapping of operations
- New, temporary nests
- At ARSC – reserving resources for immediate execution (adjusting of PBS batch scripts)



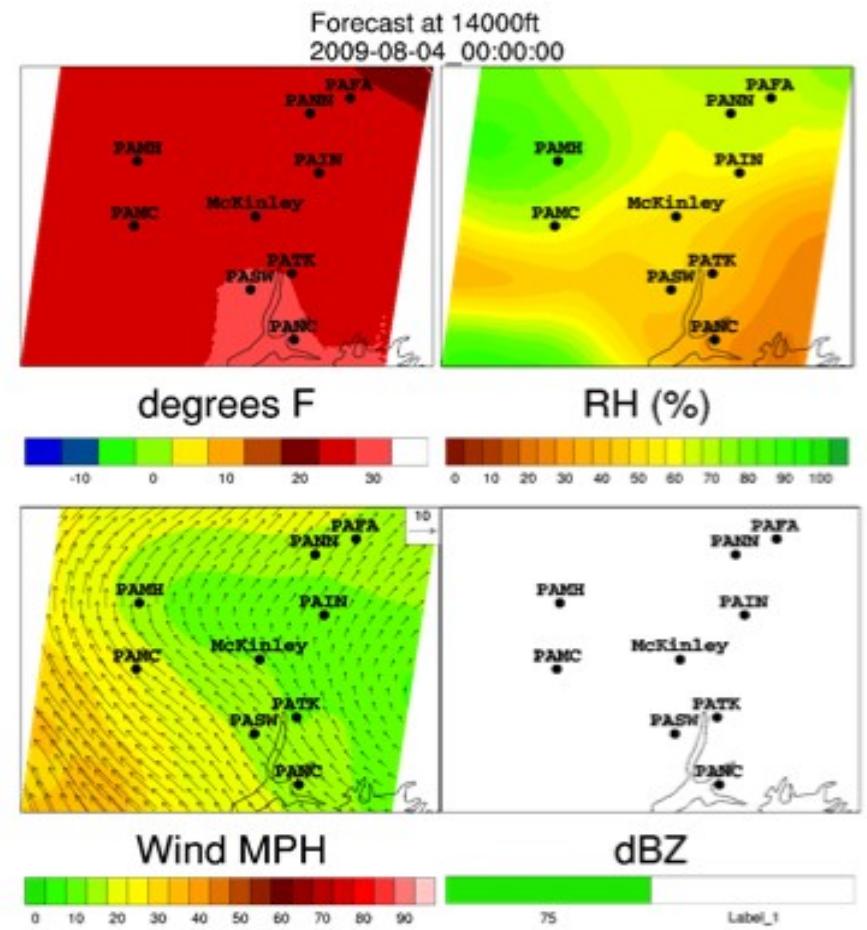
MSO Runs

- Northern Rockies Mesoscale Model (NRMM) (or the Norma Gene)
- Forecasters began to note skill in summer convective activity



Mt. McKinley Forecasts

- Tallest mountain in North America, popular for science and recreation in the summer
- At 63N, experiences bitter, violent weather any time of year
- Daily forecast for 7,000, 14,000, 17,000 and 20,000 feet



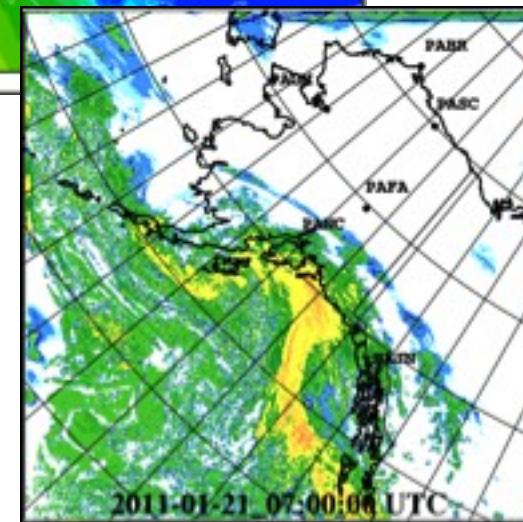
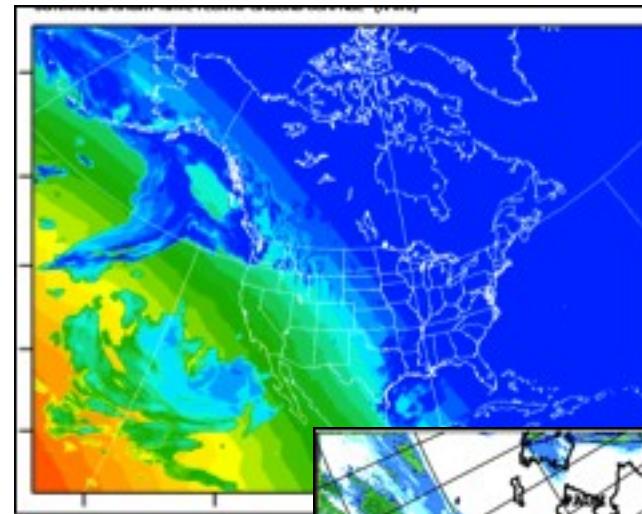
Constant migrations to new systems

- System retirements
- Performance issues
- Making room for other users – time/space constraints

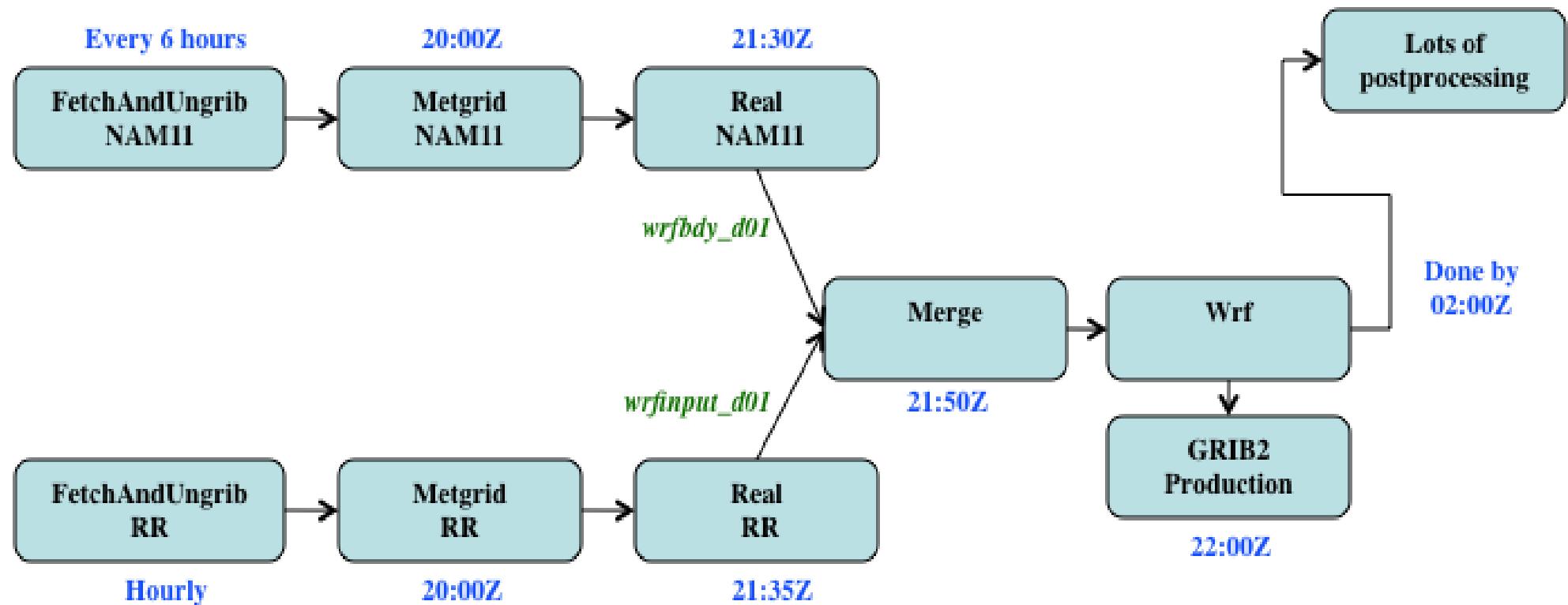


- 48-hr forecasts, 2x daily, 256 cores
- 1050x1050x51 grid points at 3km
- Initialised by RAP, LBC's from AK NAM 12
- Multiple streams of pre-processing now
- Real-time GRIB production during simulation for AWIPS

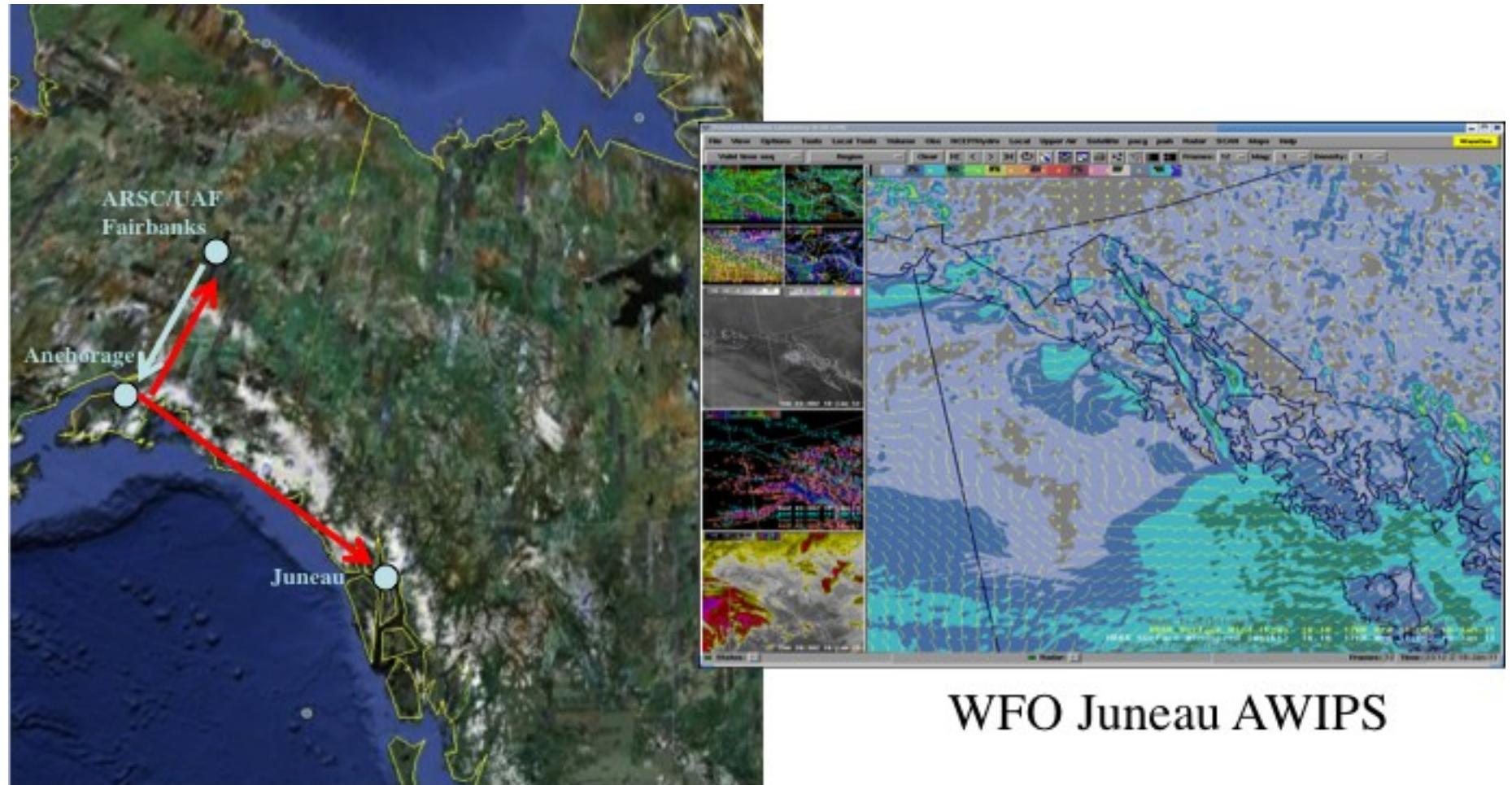
2009 - HRRR-AK



HRRR-AK Process Flow



HRRR-AK Distribution

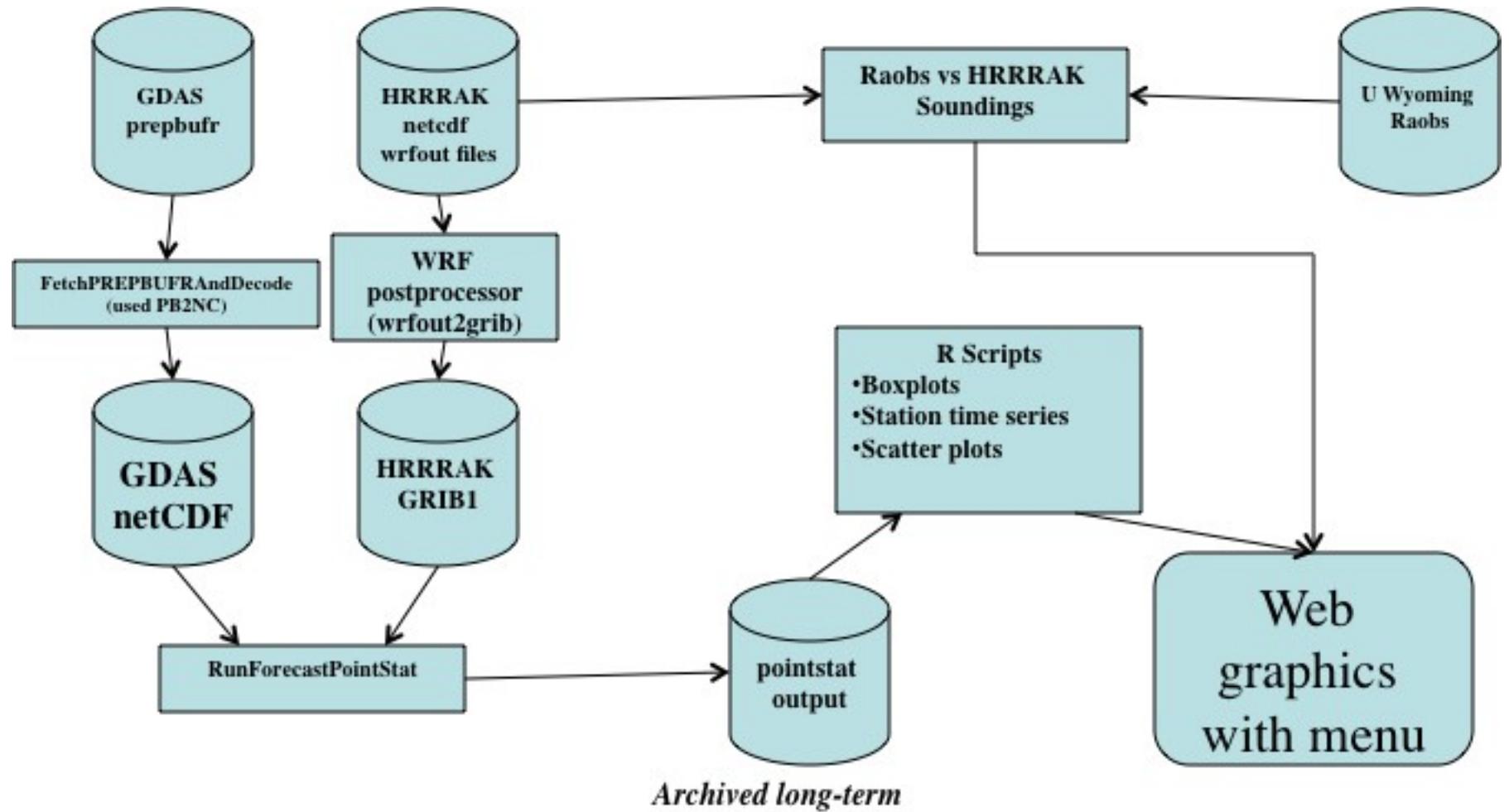


2010 – DTC support

- Model Evaluation Tools (MET)
- Sounding verifications
- Longer term summaries
- Experimental introduction of Gridpoint Statistical Interpolation (GSI) data assimilation



Verification Post Processing



Verification Products

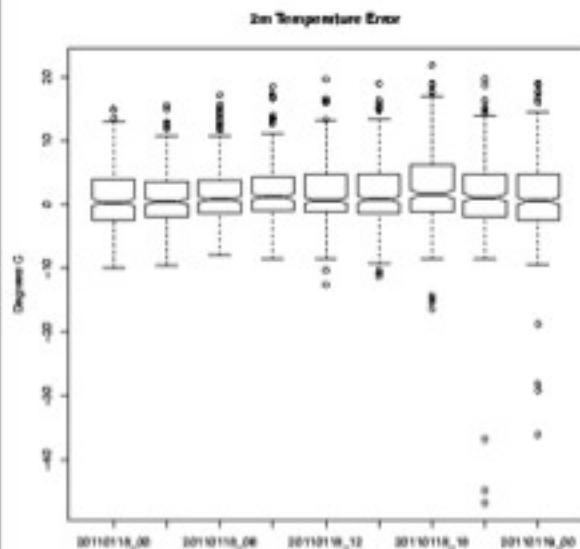
The screenshot shows a web browser window with two tabs: 'Forecast Errors for Ind...' and 'Scatterplot_201101200...'. The active tab displays the URL 'weather.arsc.edu/Operational/HRRR-AK/Pr...'. The main content area has a title 'Forecast Errors for Individual HRRR-AK Forecasts' and a form for entering forecast start time with fields for Year (2011), Month (01), Day (18), Hour (00Z), and a Go button.

The screenshot shows a web browser window titled 'Untitled' with the URL 'about:blank'. The main content area displays the title 'Forecast Errors for Start Time: 2011-01-16_12Z'. Below the title are four buttons: 'Boxplots', 'Station Time Series PAFA', 'Scatterplot for Forecast Hour 00', and 'Sounding Comparison PAFA'. Each button has a dropdown arrow next to it.

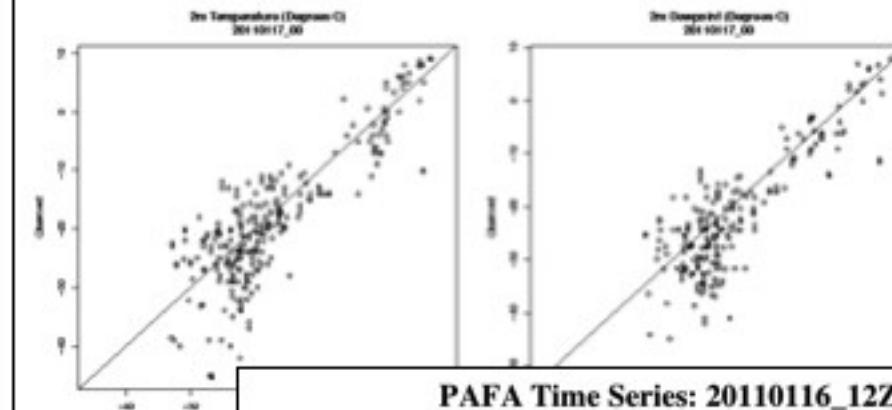


Verification Products

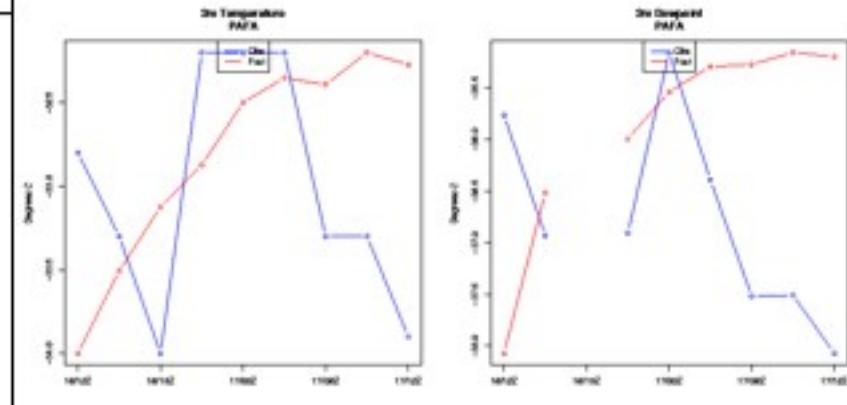
Forecast Start Time: 20110118_00Z



Forecast Start Time: 20110116_12Z
Forecast Hour: 12

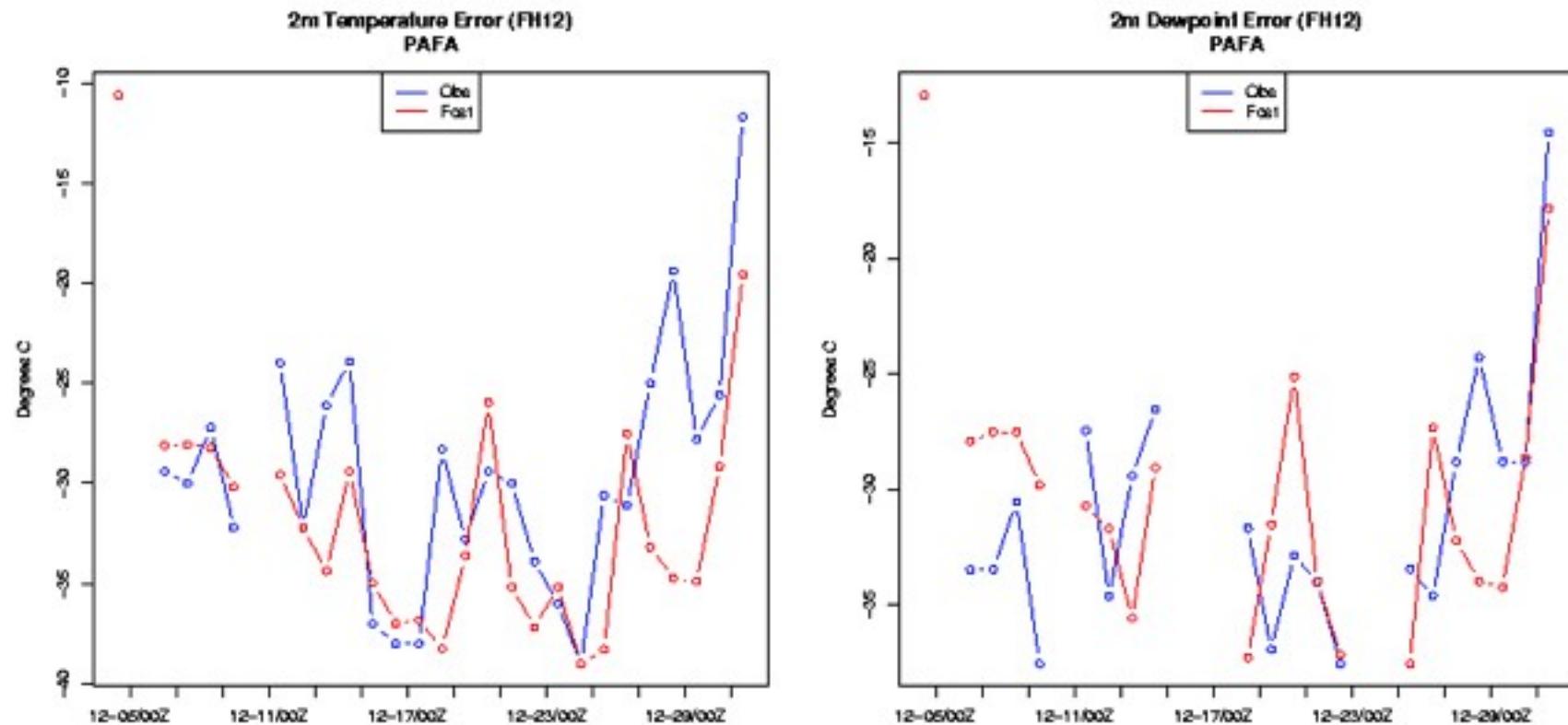


PAFA Time Series: 20110116_12Z

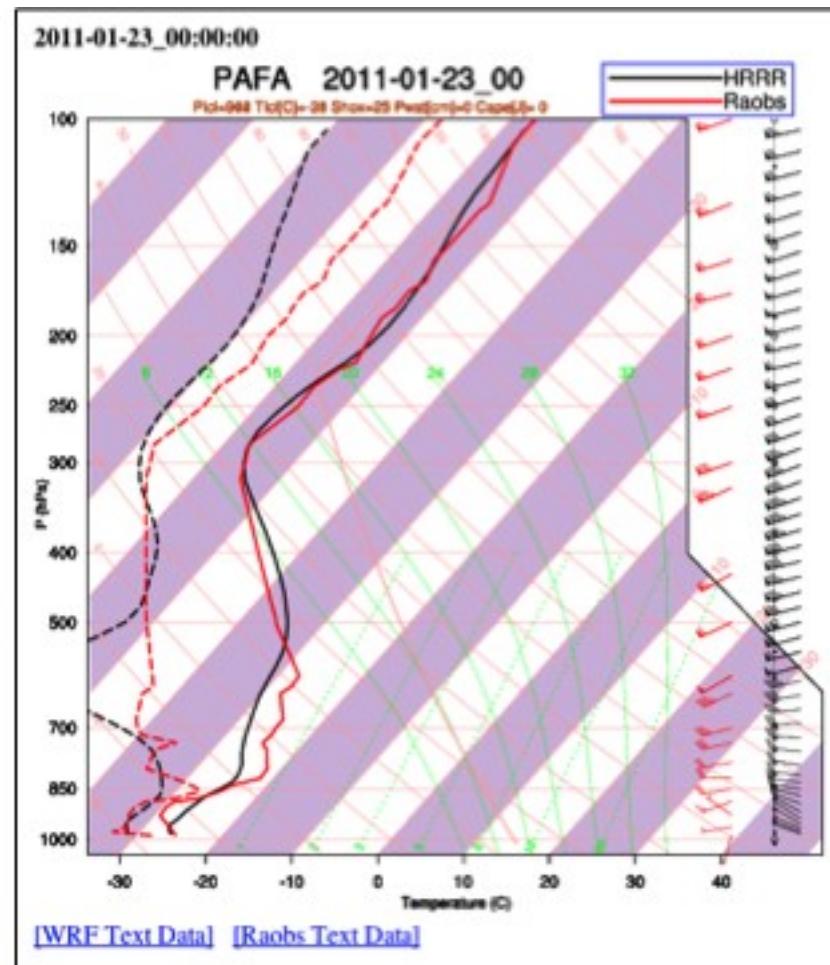
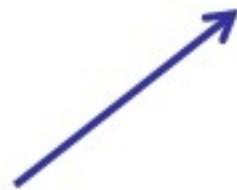
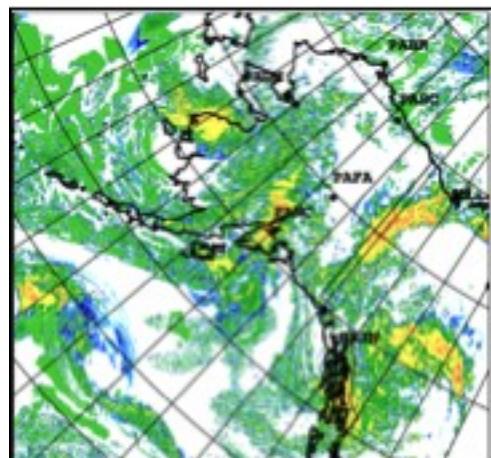


28-day Verifications

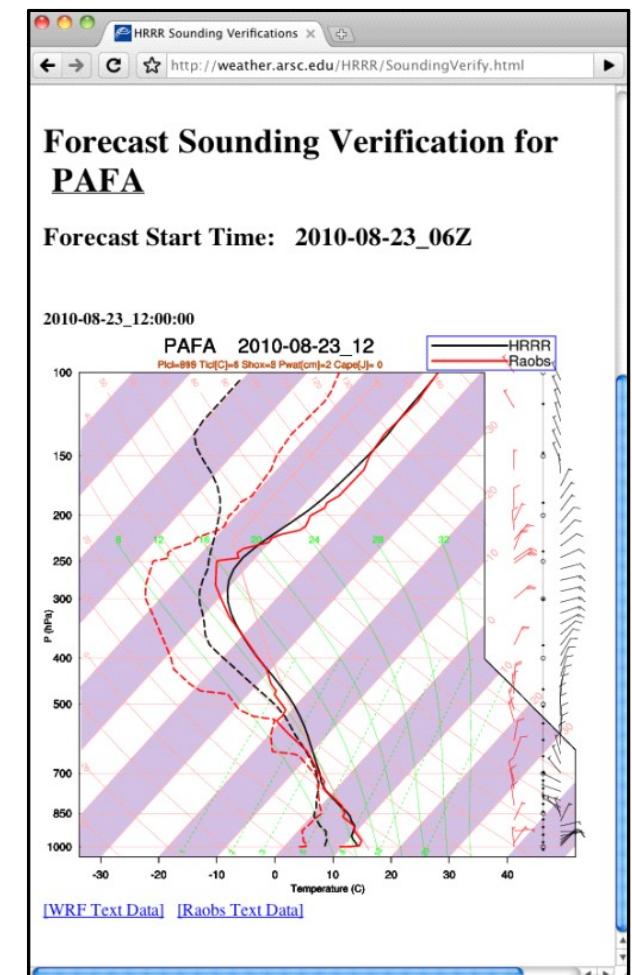
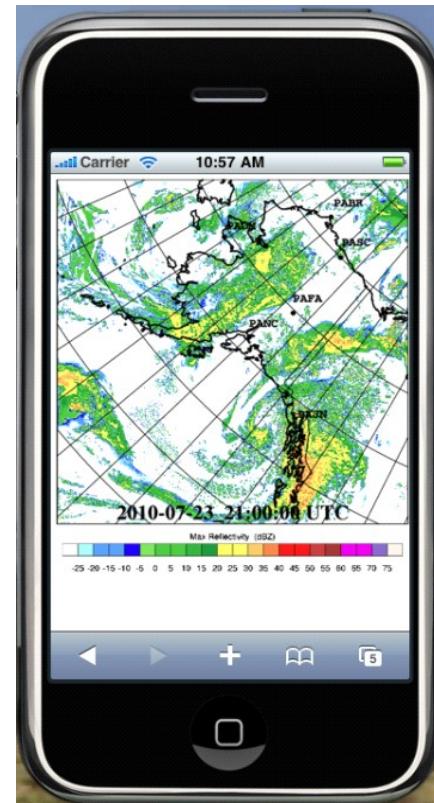
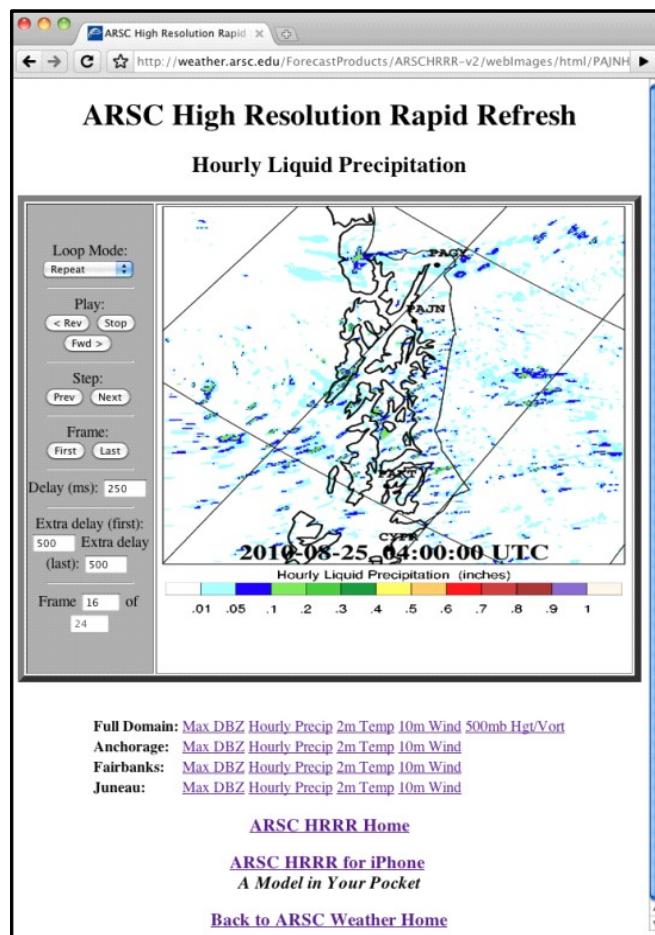
PAFA 28-Day Time Series Starting: 2010-12-04, Cycle: 00, ForecastHour: 12



Raobs vs WRF Soundings

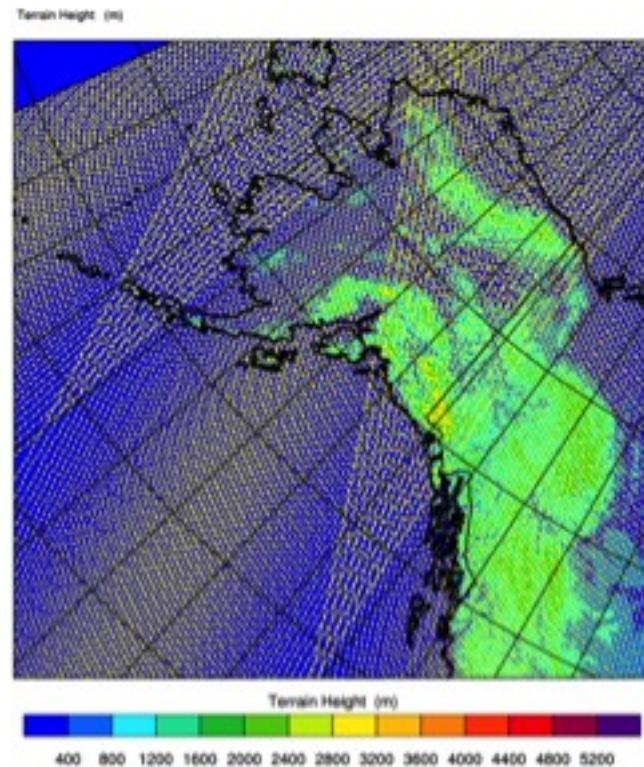


H R R R - A K W e b P r o d u c t s

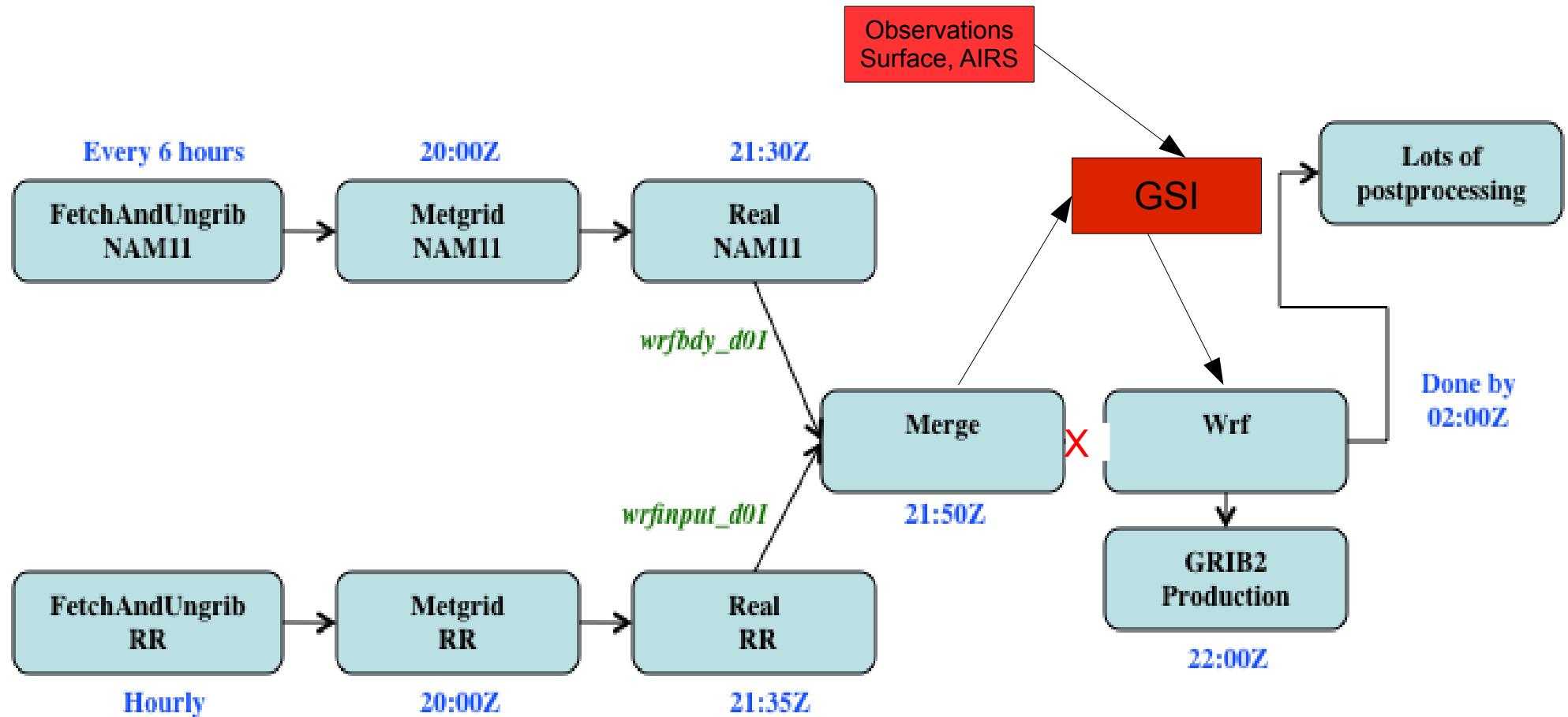


Assimilation of Atmospheric Infrared Sounder (AIRS) profiles

- Comes from instrument aboard NASA's polar-orbiting EOS Aqua satellite
- JPL and NASA SpoRT derive temperature and moisture profiles from radiances
- Short-term experiment to assimilate these, using Gridpoint Statistical Interpolation (GSI) system with HRRR-AK, concurrent with HRRR-AK runs

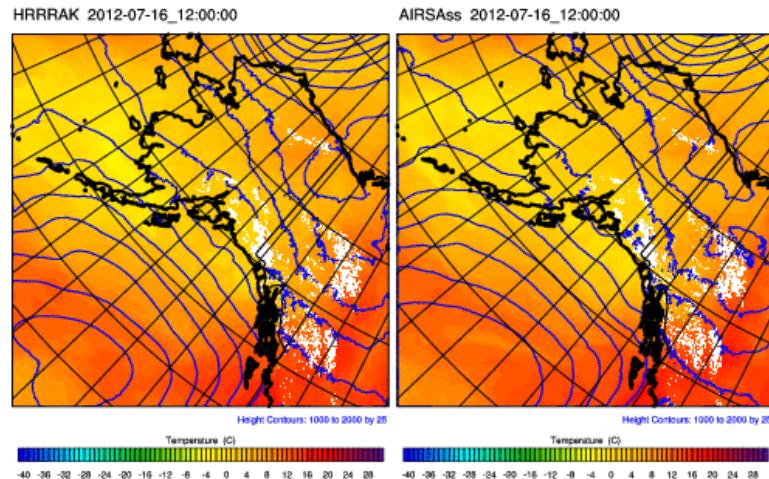


Addition of GSI components

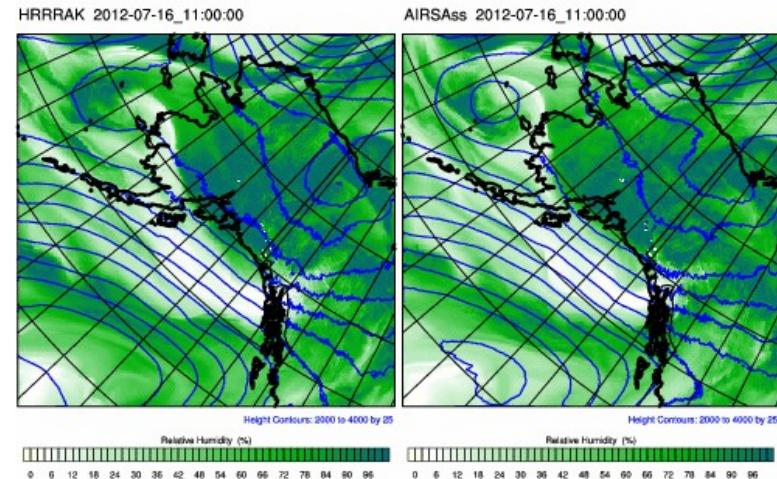


HRRR-AK / AIRSAss Comparisons

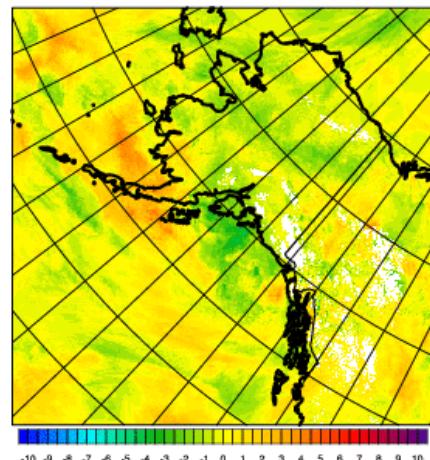
850mb T



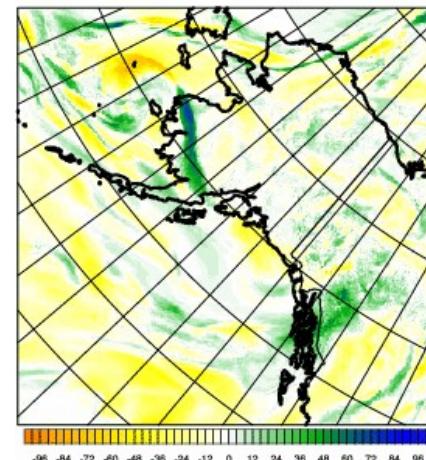
700mb RH



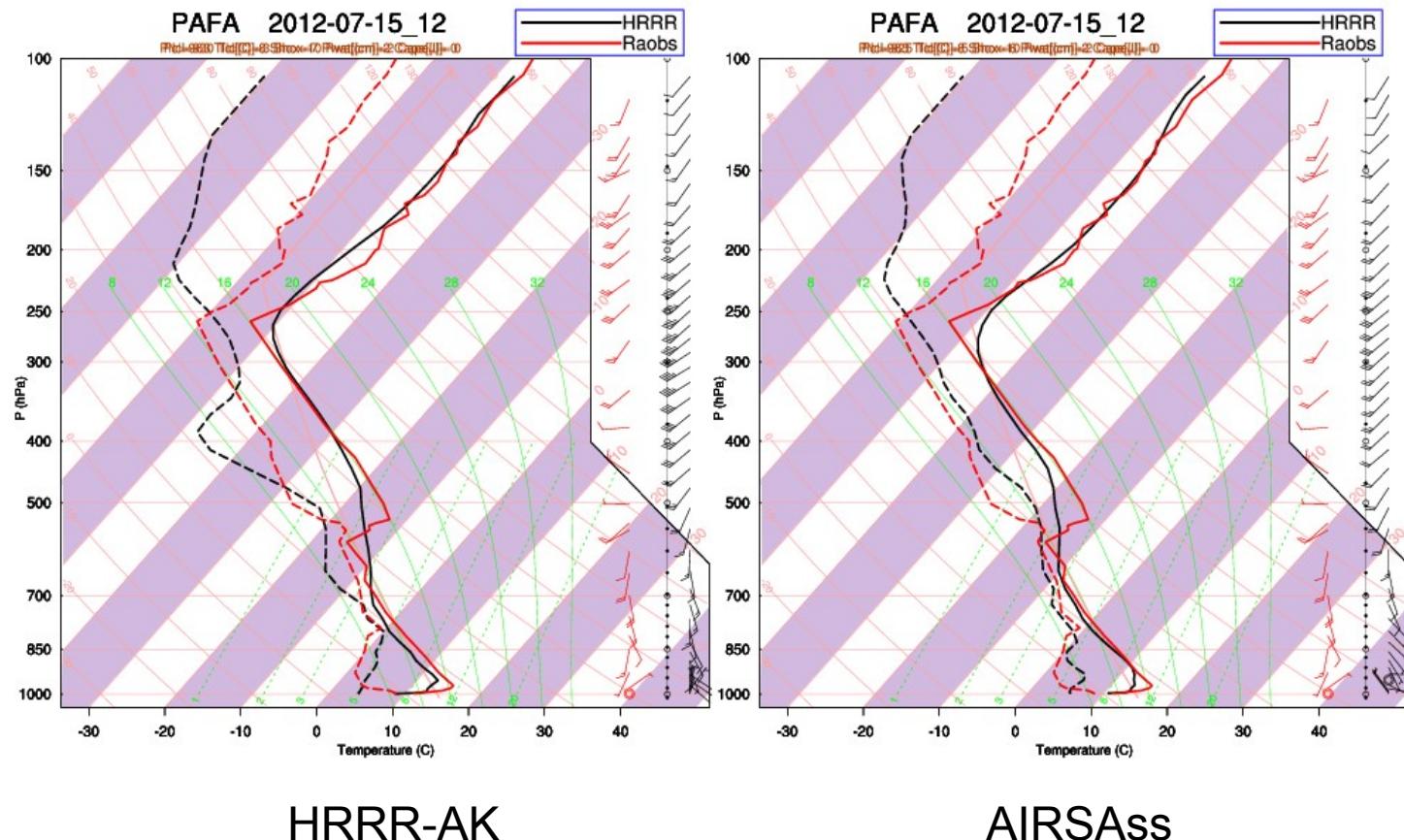
AIRSAss minus HRRRAK 2012-07-16_12:00:00



AIRSAss minus HRRRAK 2012-07-16_11:00:00



HRRR-AK / AIRSAss Sounding/RAOBS

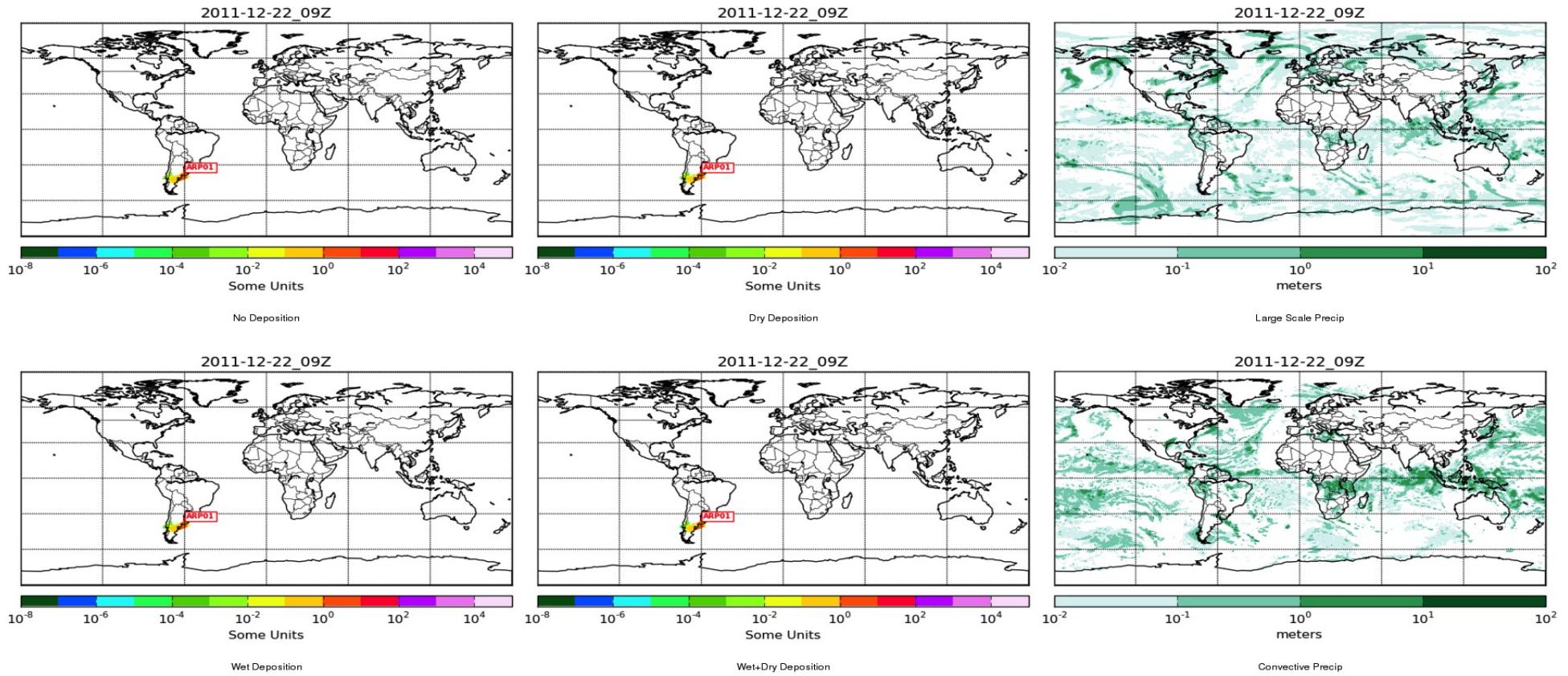


2011 – CTBTO, LPDM

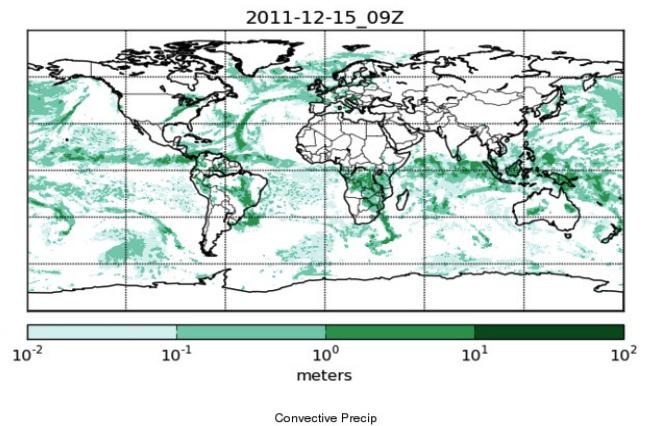
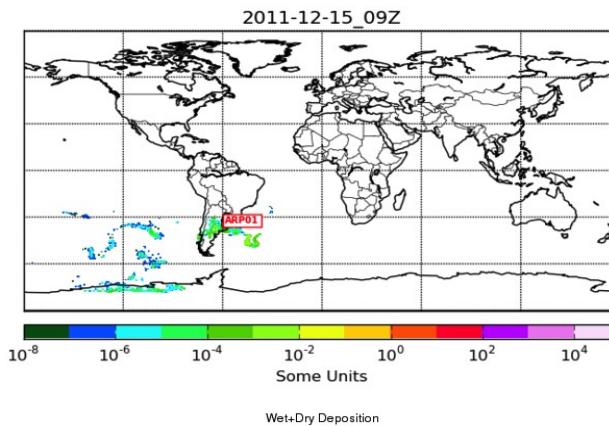
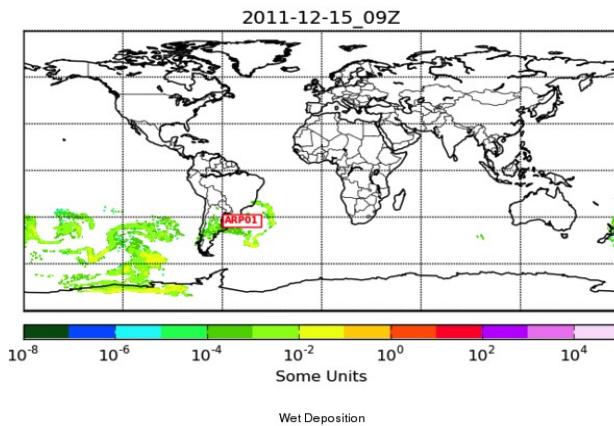
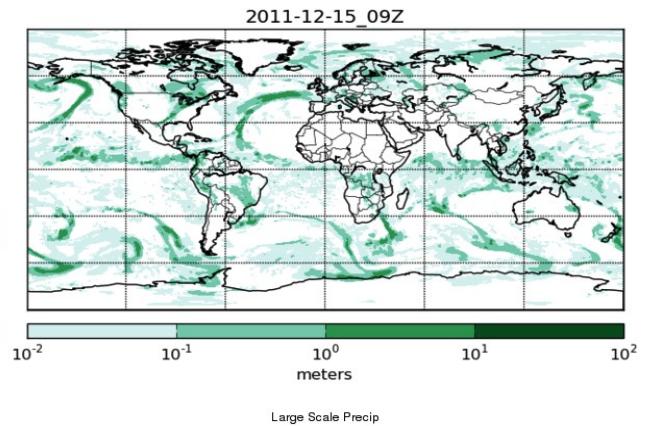
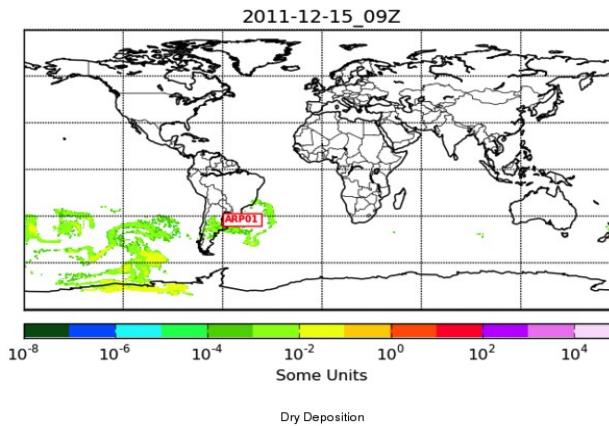
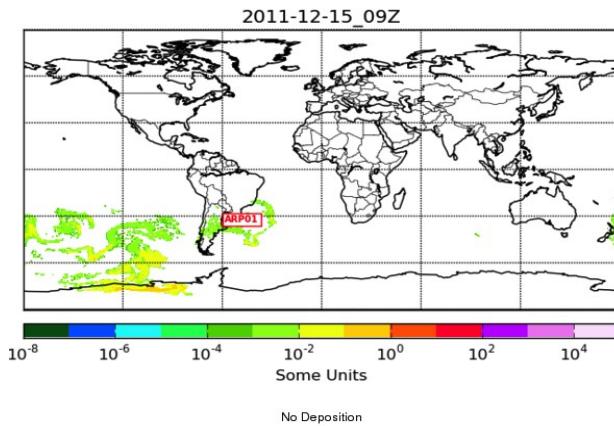
- Operational backwards simulations for United Nations Comprehensive Test Ban Treaty Organisation (CTBTO) with Lagrangian Particle Dispersion Models (LPDM)
- Forward plume simulations – operational smoke forecasts for Alaska – just proof of concept – based on FLEXPART. Required new streams of met data (GFS), new problems
- Rapid deployment of same smoke forecasts for Catalonia



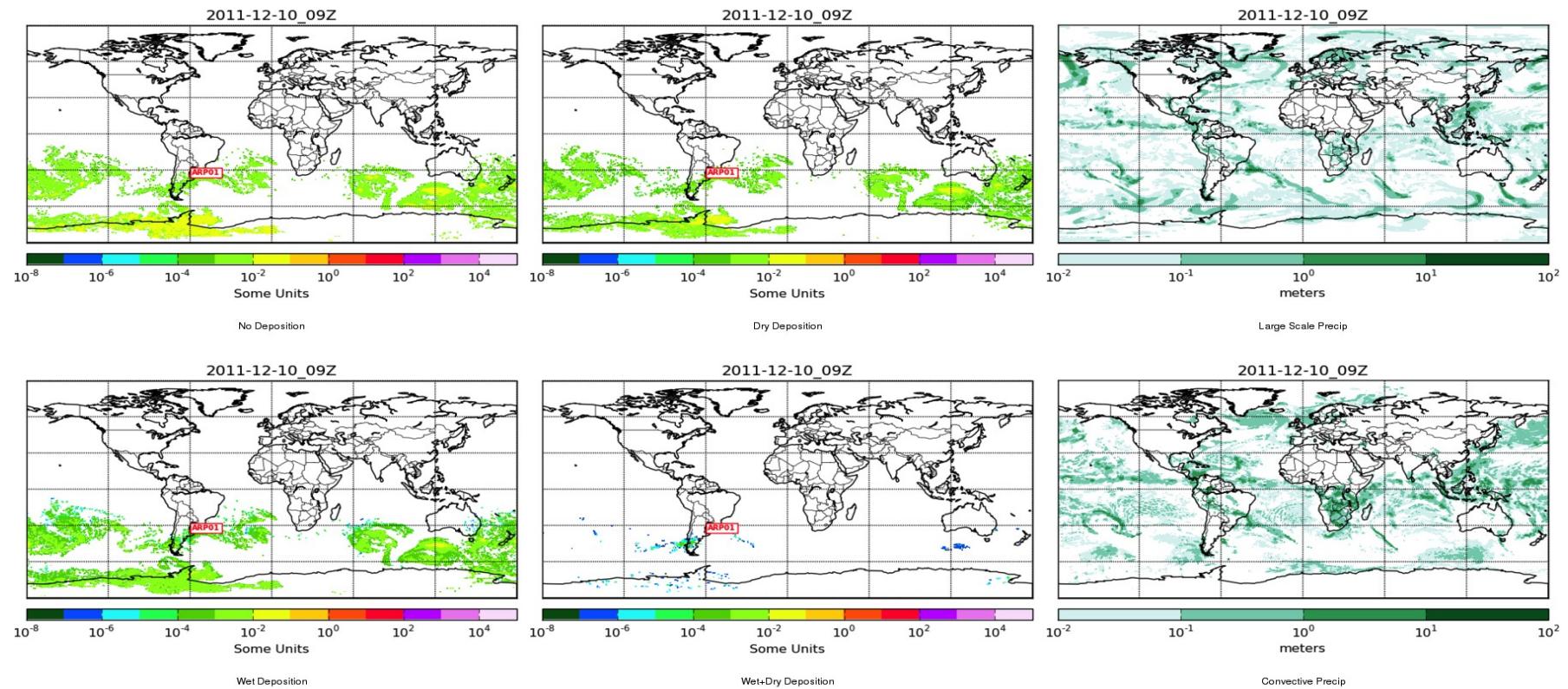
Operational Backwards Simulations



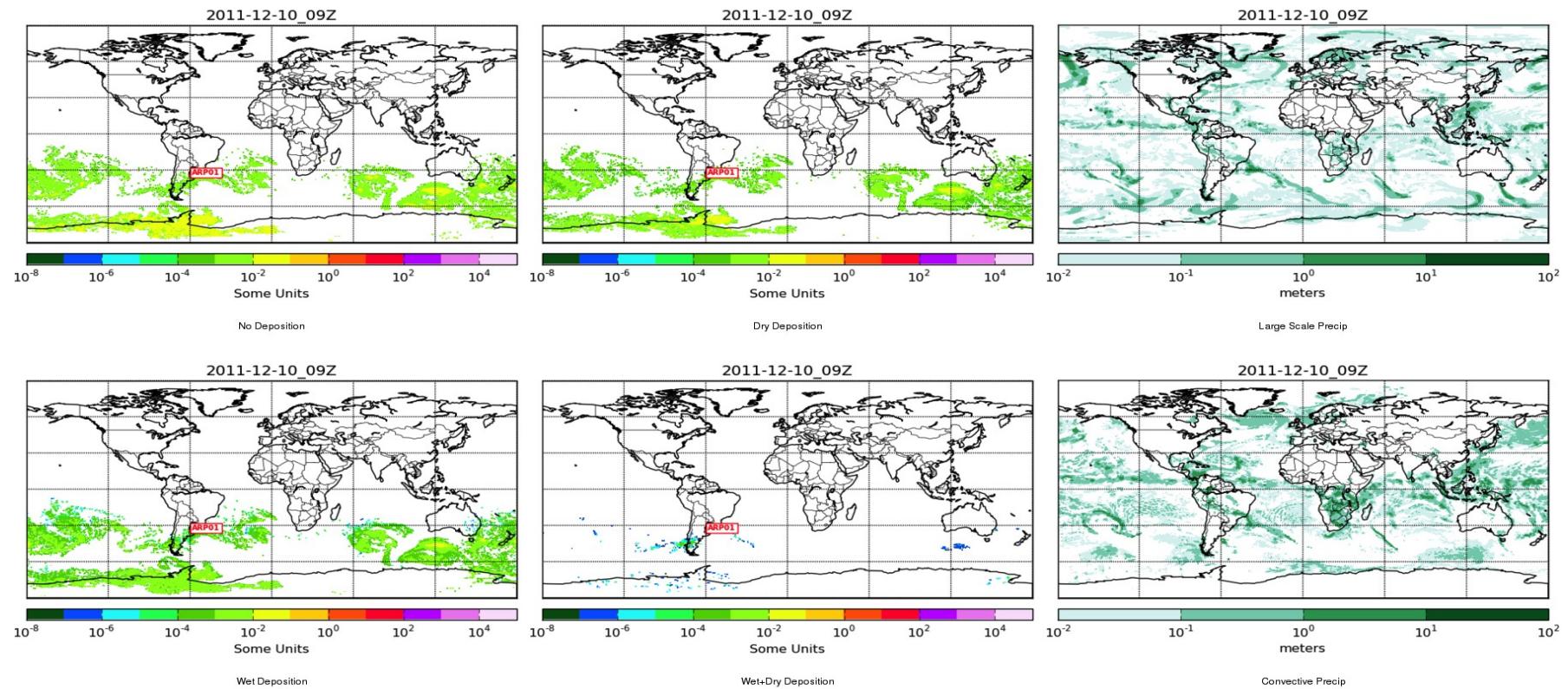
Operational Backwards Simulations



Operational Backwards Simulations

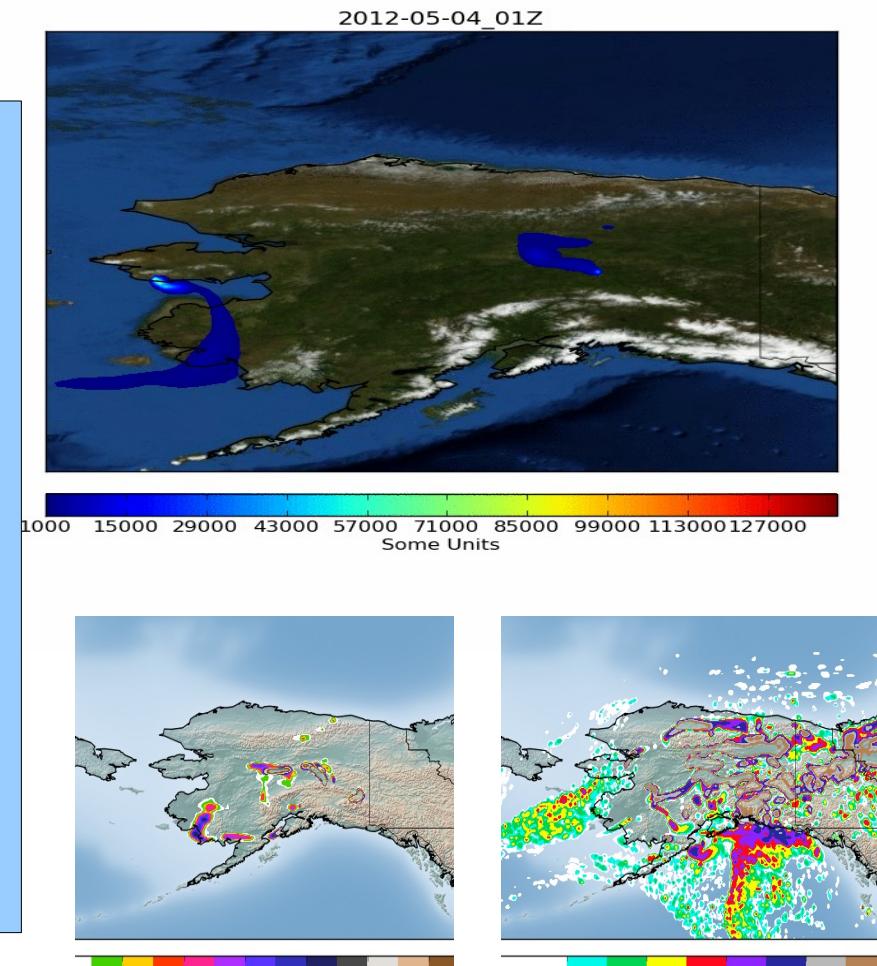
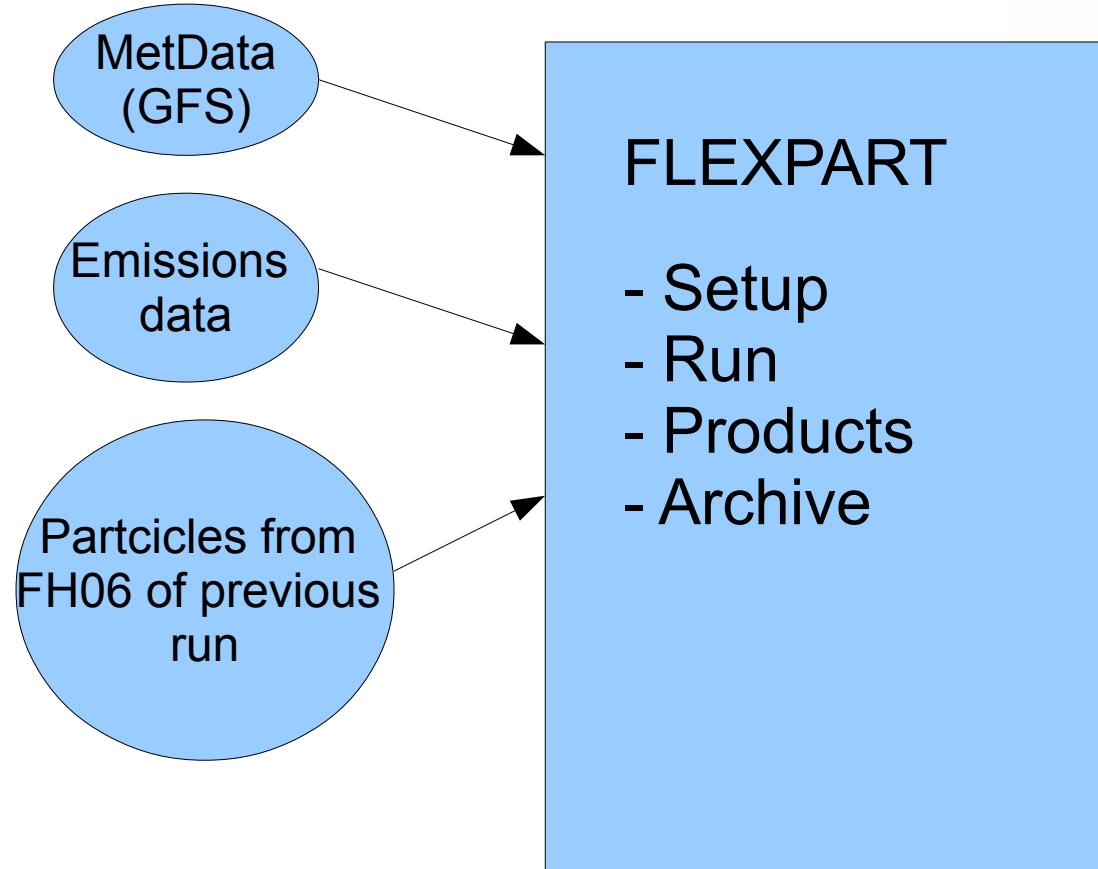


Operational Backwards Simulations



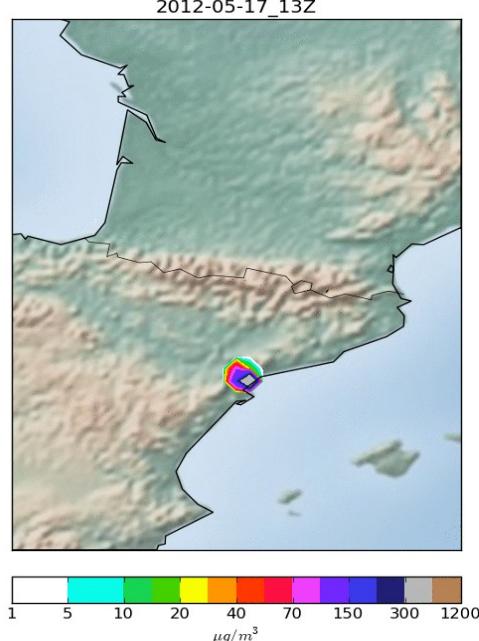
FLEXPART operational runs

- April 2012 - prototyping operational wildfire smoke dispersion - run every six hours



Fires in Catalonia!

- 18 May 2012
- Rapid deployment of operational FLEXPART for Catalonia in 2 hours
- Like the Alaska runs, ran every six hours using same GFS met data



Inbox (2,329) UA Mail - FW UA - Calendar Index of /Op Live Dawg C Any information

https://www.facebook.com/don.morton.7/posts/366087773448370?notif_t=feed_comment

Catalan NCT The Aurora Live (NIC) Google UAF Weather Tethys | Tethys bit UAF VPN Other Bookmarks

Meritxell Pagès > **Don Morton**
52 minutes ago ·

Any information you need, do not hesitate in asking me! This fire is a shame because the area is nice...in fact, I was there last summer :-(

Like · Comment · Unfollow Post · See Friendship

Don Morton Thanks Meritxell! Delia's going to call me from Vienna soon. Right now I'm looking at setting up a 0.2 degree domain from a SW corner of 38N/3.0W and a NE corner of 45N/7.0E. Just for starters.
50 minutes ago · Like

Delia Arnold wonderful! ok, so let's flexpart fires in catalonia!!!! what do you need?
43 minutes ago · Like

Meritxell Pagès Well done even being for starters! Would be synoptic/mesoscale analisys useful for you?
43 minutes ago · Like

Delia Arnold hola meritxell? estas al chat?
See Translation
42 minutes ago · Like

Meritxell Pagès No el tinc connectat, però I'm alive!
42 minutes ago · Like

Delia Arnold hehe
42 minutes ago · Like

Delia Arnold ok, so don is talking with me right now.. he is setting all this up
41 minutes ago · Like

Delia Arnold output is at 0.2 deg, not bad... in the near future we can think about flexpart-wrf
41 minutes ago · Like

Meritxell Pagès In one second I'll connect the chat
41 minutes ago · Like

Delia Arnold juhu!!!! everyone excited and don under pressure!!!
40 minutes ago · Like

Meritxell Pagès :-)
40 minutes ago · Like

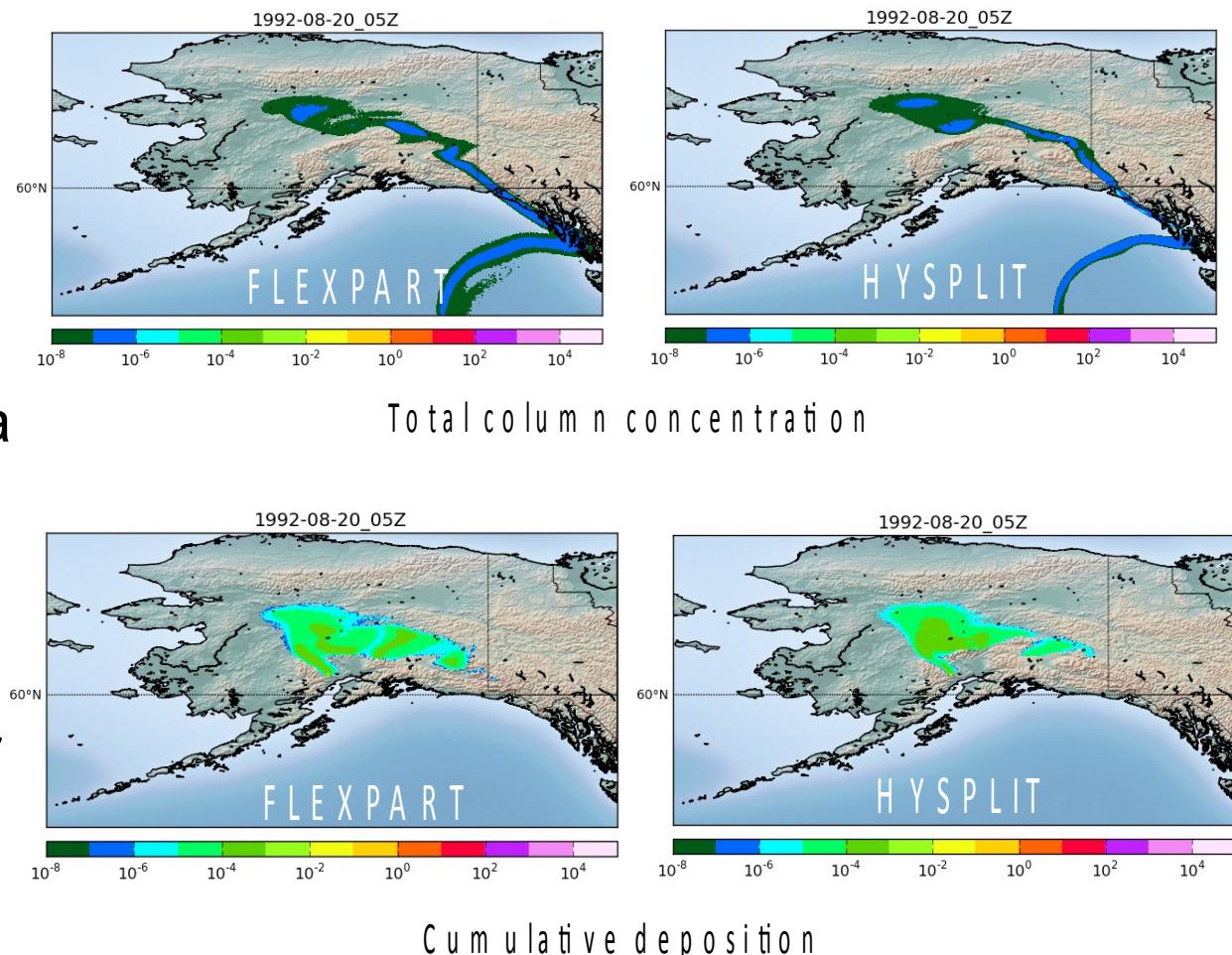
Don Morton Ok, because the 18Z GFS wasn't out yet, I initialised on the 12Z, and we're now up to Forecast Hour 5. This may take a few hours to run. In retrospect, maybe I should have initialised with 1.0 deg GFS rather than 0.5, but that's OK, we'll see how this works :). Every forecast hour is taking longer to compute because we're increasing the amount of smoke in the model. I'm going to set this up so that it automatically runs four times a day, just like the Alaska model (but we don't have any fires yet).
10 minutes ago · Like

Write a comment...

Chat (Offline)

Long-term operational vision

- Multimodel operational runs (begin with FLEXPART and HYSPLIT)
- Outputs in common framework
- Rapid deployment capabilities - literally go to a web page (or even better, have an event trigger the operations), enter coordinates, and launch a multi-model, multi-member set of operational runs until event has passed.

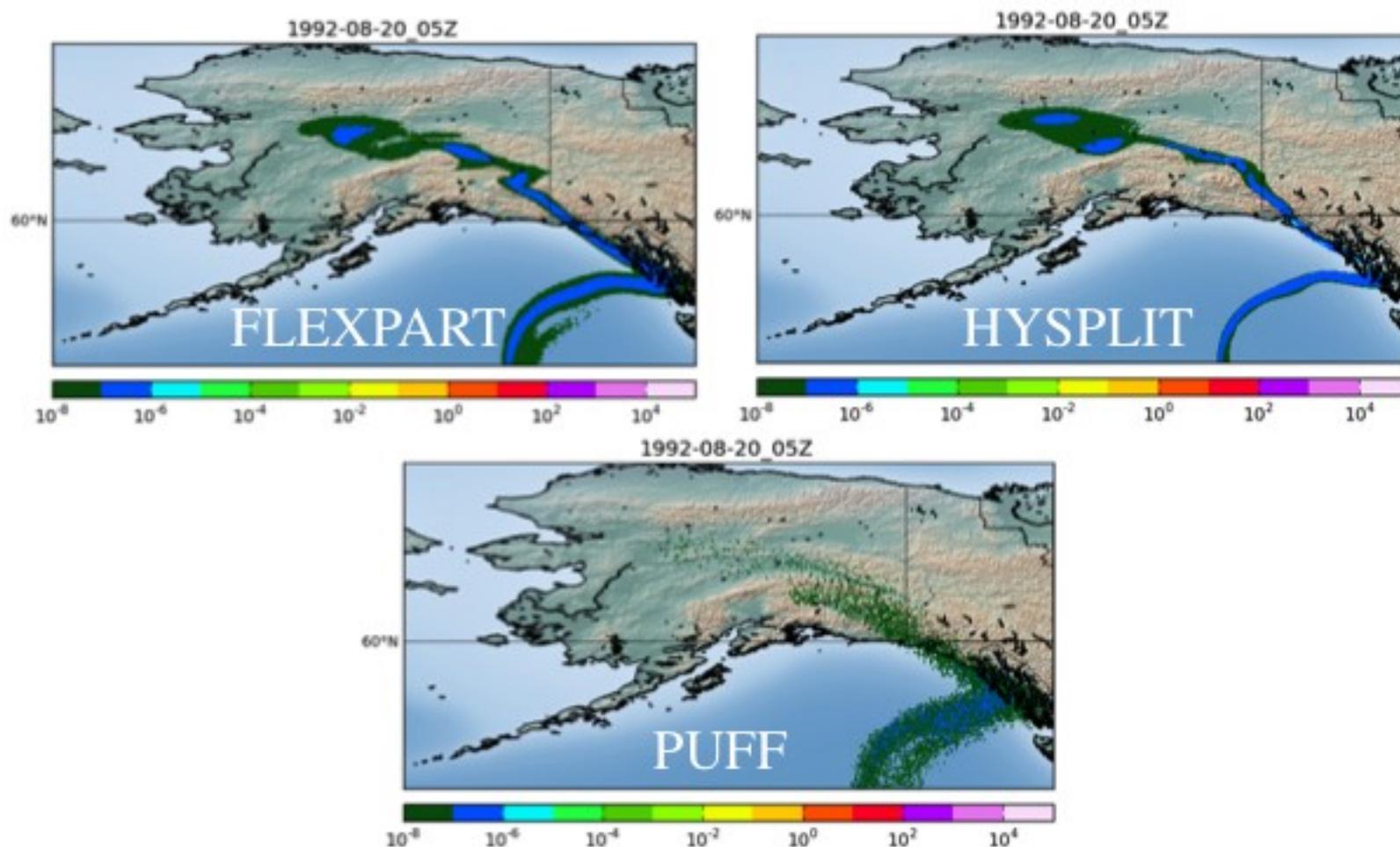


Volcanic ash transport and other emergency response activities

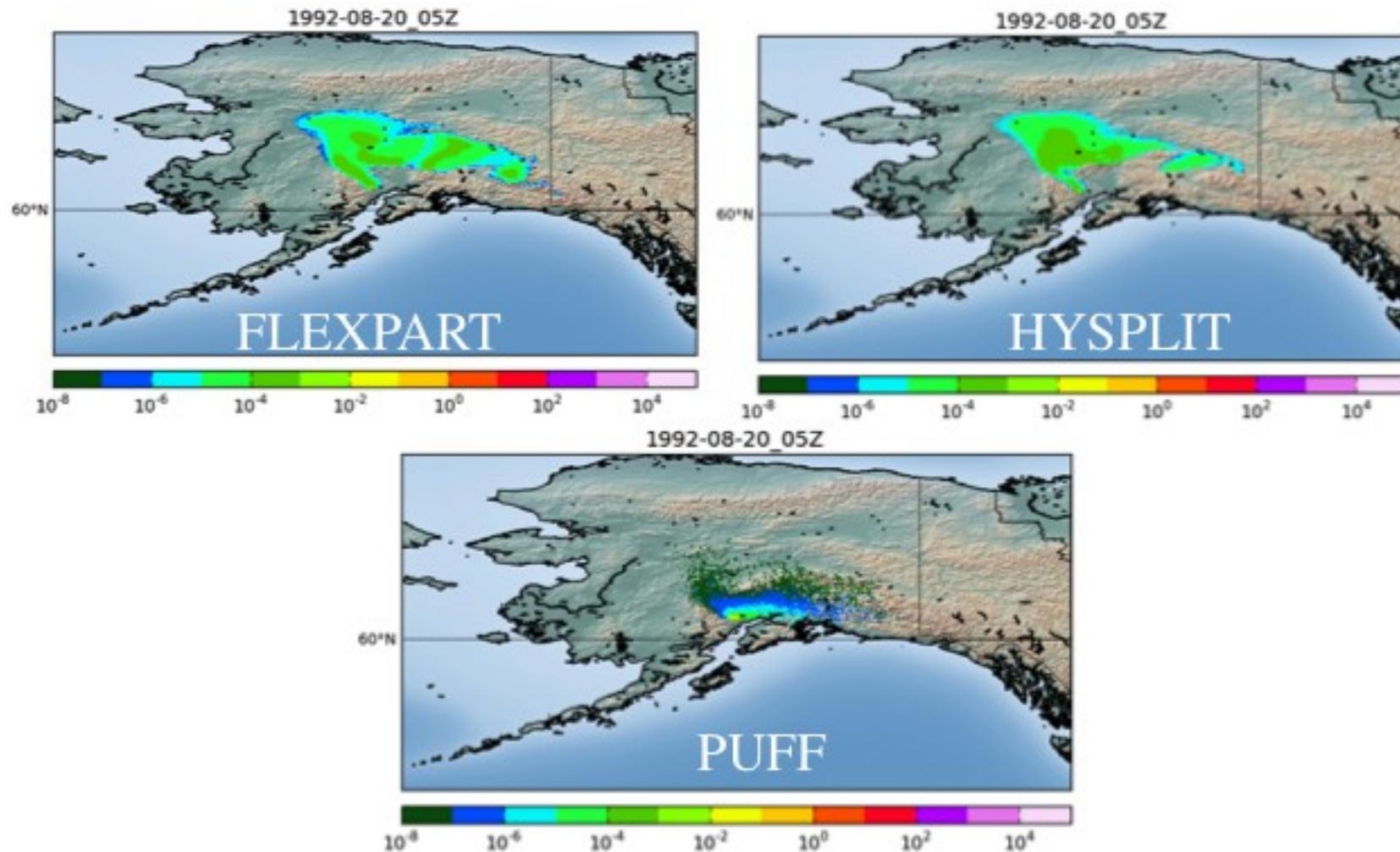
- There is a need to rapidly deploy targeted forecasts for surprise events like nuclear accidents and volcanic eruptions
- Need to be able to conglomerate outputs from different models and configurations into unified environment for rapid assessment
- Vision – literally, with the push of a button, or automated, launch an ensemble of models and configurations and display results in a common framework



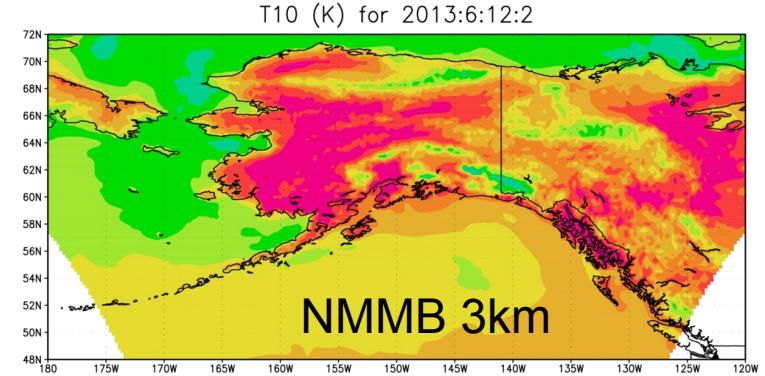
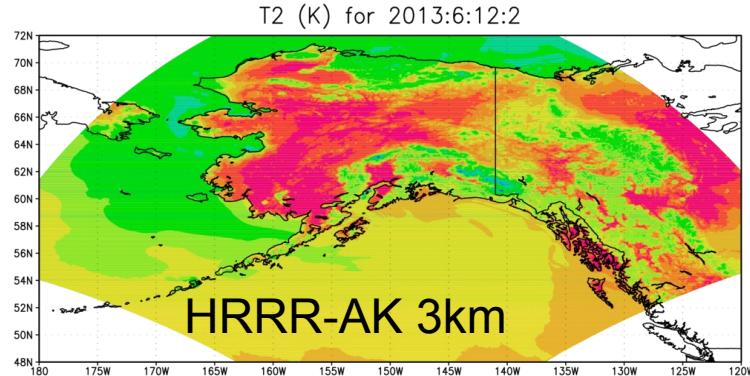
Total integrated column concentration, Mount Spurr eruption 1992



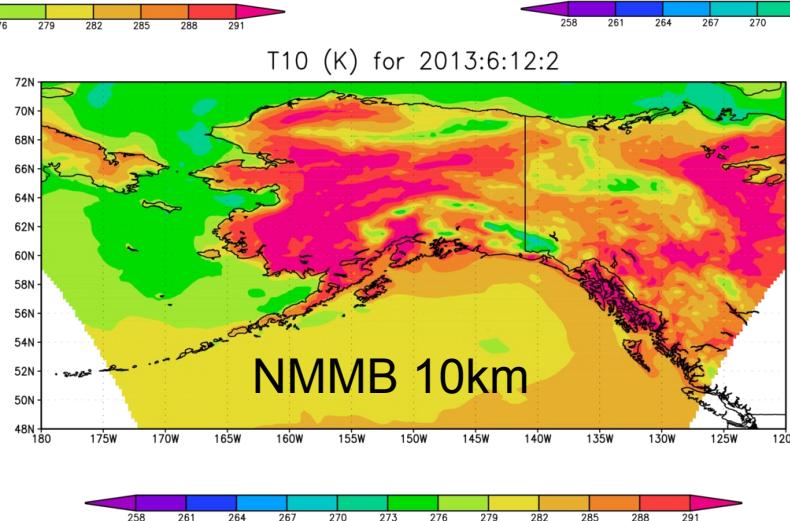
Cumulative deposition, Mount Spurr eruption 1992



Operational NEMS/NMMB for Alaska



- Learning, porting to GNU environments took weeks of work
- Setting up operationally took less than a day
- 10km NMMB 4x per day, 48 hour forecasts



Some software engineering perspectives in a real-world constantly changing environment



From monolithic bash script to loosely coupled Python components

- Various learning experiences over the years - changes were often mandated, requiring "uncomfortable" hacks
- Examples
 - From one run to multiple runs per day - conflicts in timing, filesystems
 - NWS requests
 - Start a forecast (e.g. 00Z) before the forecast start time (e.g. 00Z)
 - Get GRIB output files delivered to LDM as soon as each forecast hour completes
 - Special requests for new, temporary nests



From monolithic bash script to loosely coupled Python components

- More Examples
 - Frequent changes from one supercomputer to another
 - Changes in filesystems often mean new locations for scratch work, performance changes, etc.
 - On supercomputers, not all filesystems are available from all processors, meaning we frequently need to stage files from one place to another
 - e.g. We typically want our input and output files on archive filesystems, but compute nodes only have access to high-performance filesystems
 - high-performance filesystems have limited storage
 - Need to do a lot of juggling
 - Timing is critical - if input data is late or incomplete, what do we do? We need to have everything cleared for next forecast



A paradigm of loosely-coupled Python components

- The Unix philosophy. Excerpts from *The Art of Unix Programming*, by Eric S. Raymond
 - Rule of Modularity: Developers should build a program out of simple parts connected by well defined interfaces, so problems are local, and parts of the program can be replaced in future versions to support new features. This rule aims to save time on debugging code that is complex, long, and unreadable.



A paradigm of loosely-coupled Python components

- The Unix philosophy
 - Rule of Composition: Developers should write programs that can communicate easily with other programs. This rule aims to allow developers to break down projects into small, simple programs rather than overly complex monolithic programs.
 - Rule of Simplicity: Developers should design for simplicity by looking for ways to break up program systems into small, straightforward cooperating pieces. This rule aims to discourage developers' affection for writing "intricate and beautiful complexities" that are in reality bug prone programs.



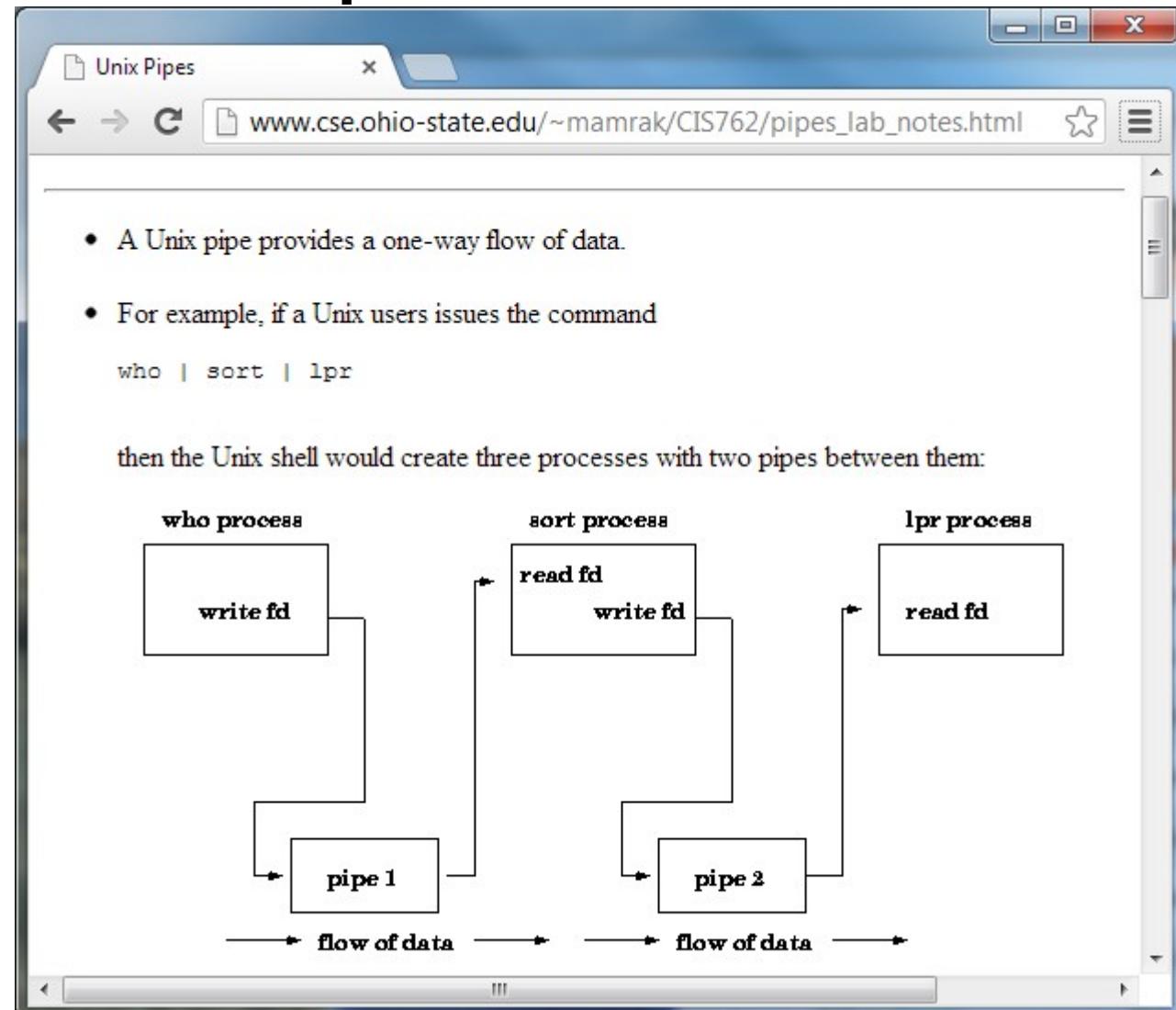
A paradigm of loosely-coupled Python components

- The Unix philosophy
 - Rule of Parsimony: Developers should avoid writing big programs. This rule aims to prevent overinvestment of development time in failed or suboptimal approaches caused by the owners of the program's reluctance to throw away visibly large pieces of work. Smaller programs are not only easier to optimize and maintain; they are easier to delete when deprecated.



A paradigm of loosely-coupled Python components

- The Unix philosophy

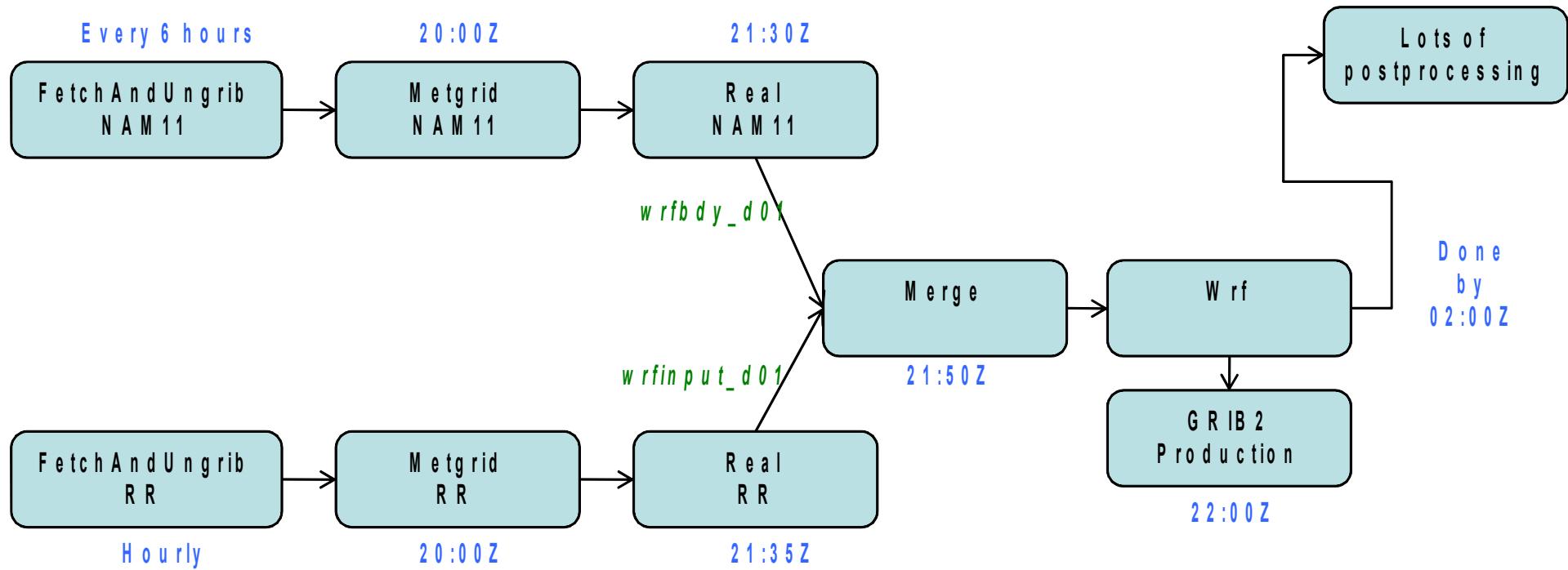


A paradigm of loosely-coupled Python components

- Ideally, our operational system is built up of loosely-coupled, cohesive components
 - Each component has a specific job to do, and its function is self-contained within the component (cohesive)
 - We try to minimise interactions (side effects) by making the components loosely-coupled. Interfaces are well-designed and simple and, ideally, we think in terms of a single input stream and a single output stream



H R R R - A K Process Flow (00Z Forecast)



In the real world

- People who want operational systems aren't necessarily interested in the investment needed for a good, robust design
- Often, we are not fully aware of our requirements when we start developing, and need to be ready to modify certain functionalities.
- Thinking in terms of small, functional, loosely-coupled components is, of course, complex, but it also gives us flexibility not found in a huge, monolithic system
- Software design and deployment is all about recognising imperfection and managing many tradeoffs

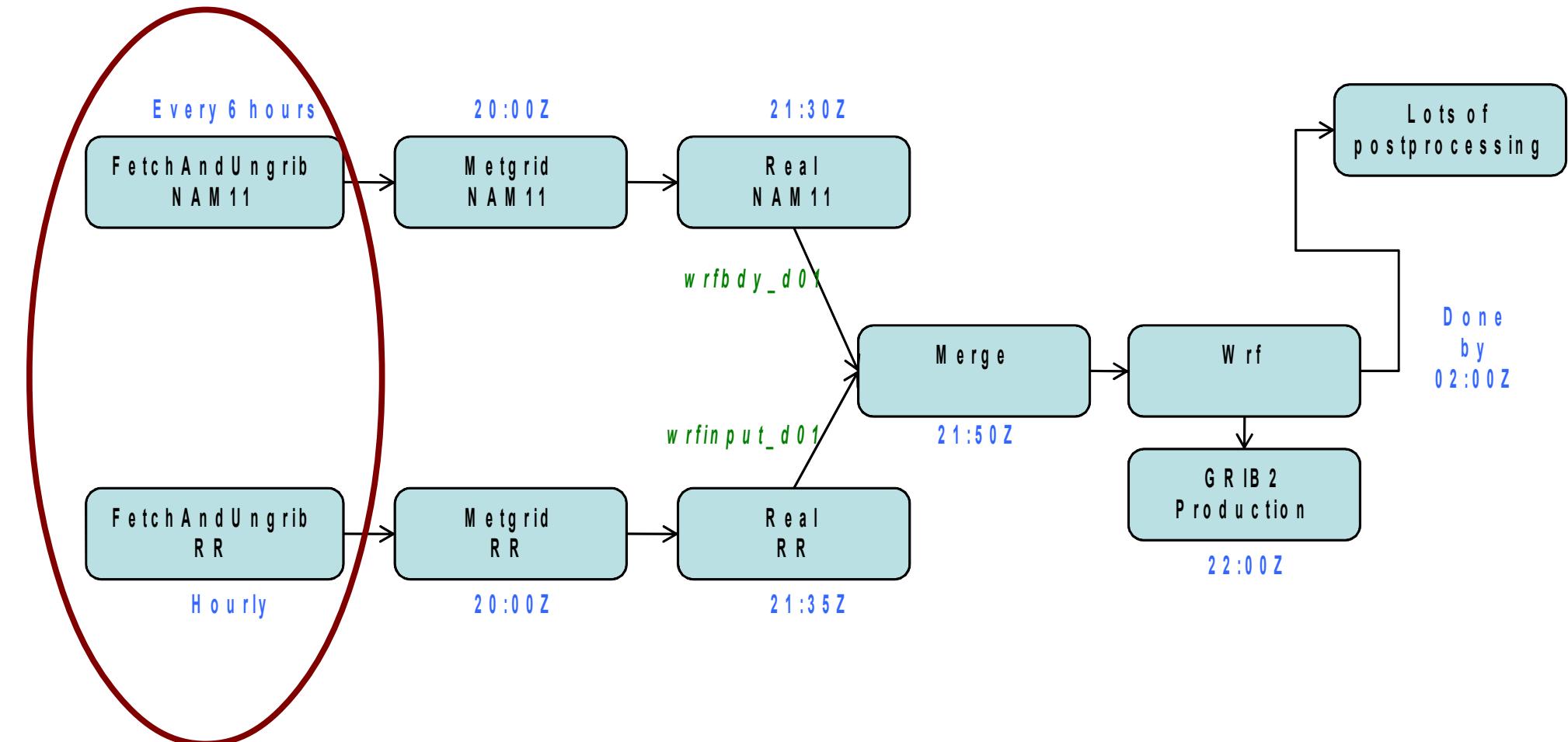


A closer look at our HRRRAK operational system

- Want to make you aware of the many issues and potential problems we face
 - We try to handle some potential problems beforehand
 - With other problems we gamble and hope for the best
 - Tradeoffs - there comes a point where increased error handling requires a huge resource commitment, and we must weigh this against the odds and the ramifications of potential problems



H R R R - A K Process Flow (00Z Forecast)



HRRRAK input data

- Two sources, one every six hours, one hourly
- Sometimes the sources are delayed, so we need to keep trying (but not forever)
- How do we know when a remote file is "complete?"
Depends on server capabilities, but sometimes we have to download it first and make sure it exceeds a predetermined threshold size. If not, we try again a few minutes later
- Sometimes file transfers simply fail, so we need to be able to recover, and try again, but not forever



HRRRAK Input Data

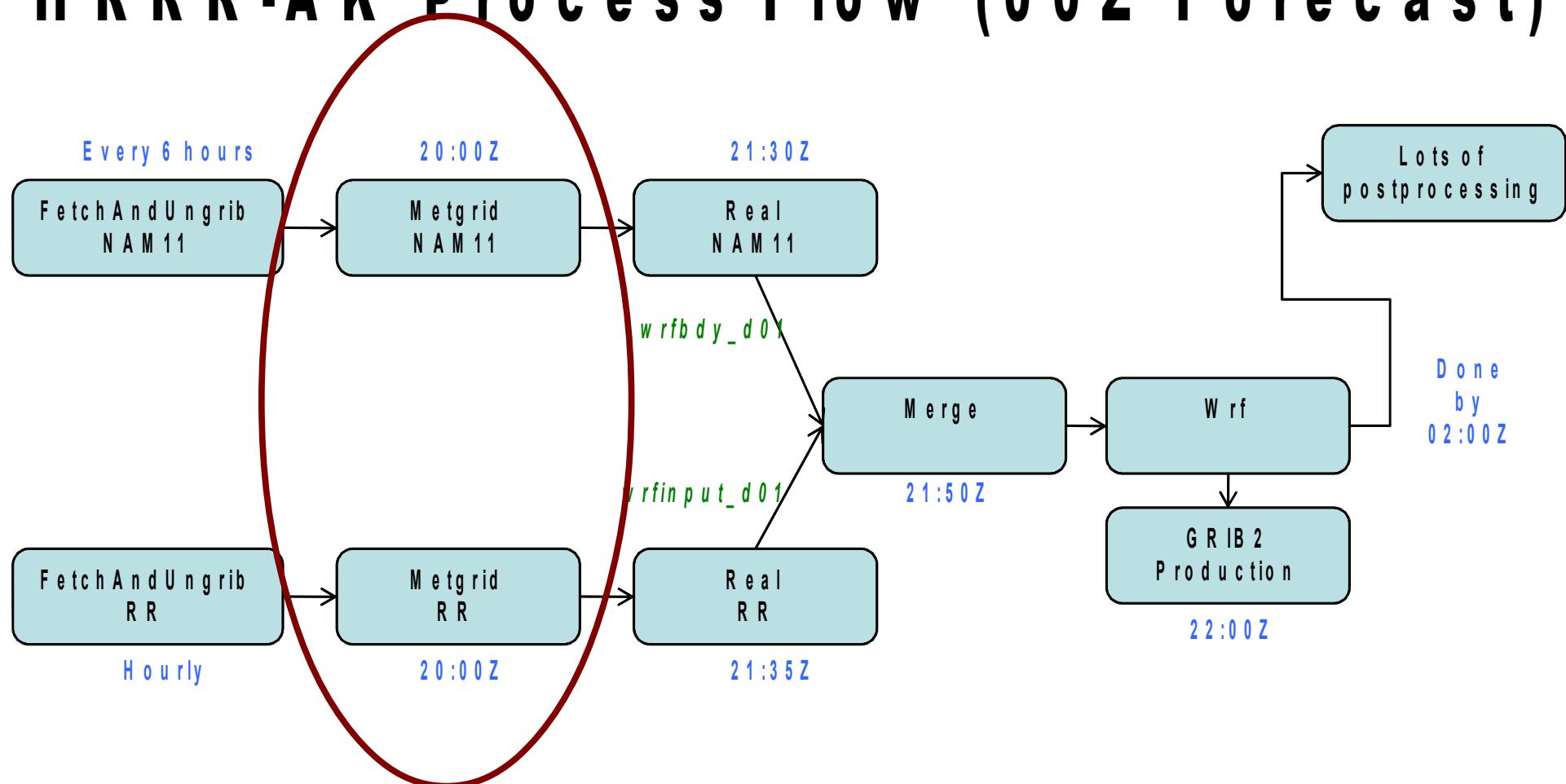
```
##### Various criteria for success, retry, or giving up
# If the file size on the server is not at least this big, then don't retrieve
MIN_FILESIZE_THRESHOLD_BYTES = 50000000

MAX_PROCESS_AGE_MINUTES = 120      # Age at which to terminate process

FILESIZE_WAIT_TIME_SECONDS = 120    # How long to wait before trying filesize again
FILESIZE_WAIT_MAX_TRIES = 10        # How many times to try on filesize before
                                    # giving up
FILERETRIEVE_WAIT_TIME_SECONDS = 60 # How long to wait before trying file retrieve
again
FILERETRIEVE_WAIT_MAX_TRIES = 5    # How many times to try on file retrieve before
                                    # giving up
SOCKET_TIMEOUT_SECONDS = 60        # How many seconds before a socket times out
```



H R R R - A K Process Flow (00Z Forecast)

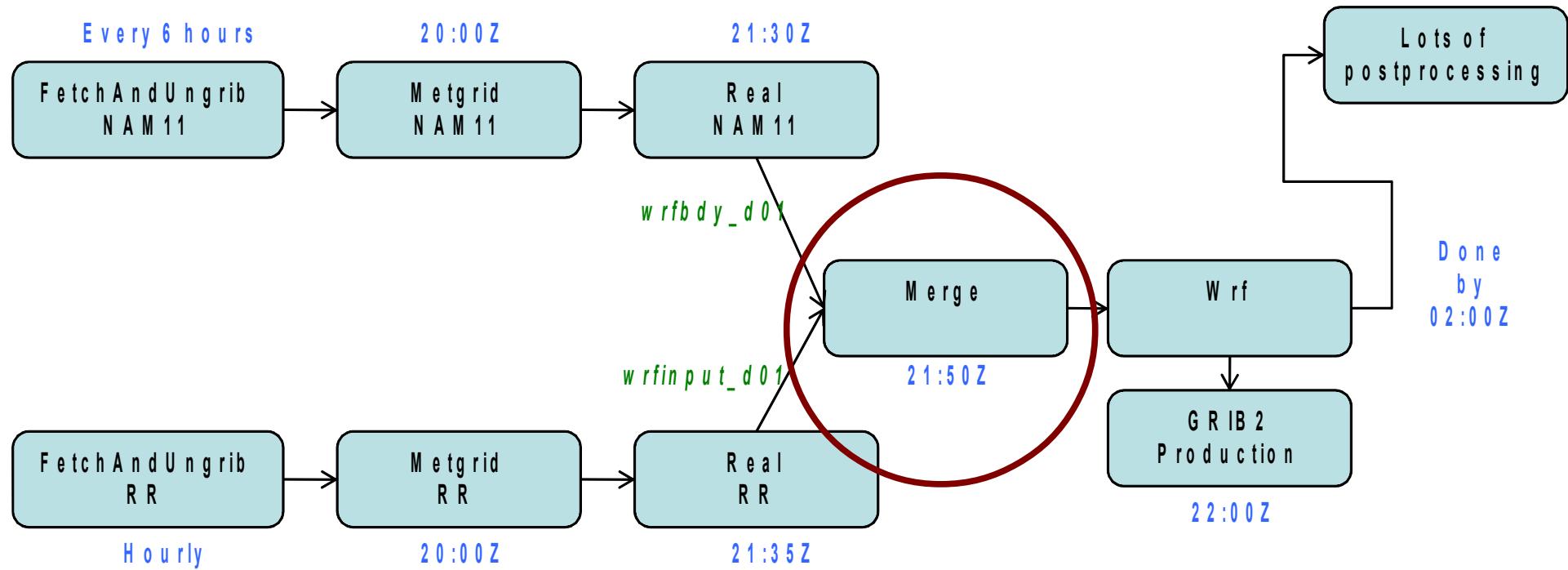


HRRAK Metgrid

- Transforms input data to a specified computational grid
 - This is done for both NAM and RAP input data
 - Metgrid for the 00Z run needs to run at 20Z, or not much later
 - The computational grid is static and pre-specified
 - The input data may or may not be there
 - If it's there, but it's not complete, or it's not there at all what do we do? Settle for a shorter forecast? Or maybe we try the previous set of input data (try starting at forecast hour 6 of the 18Z input data). What if the previous set of input data isn't there, or is incomplete? Go back to previous one?



H R R R - A K Process Flow (00Z Forecast)

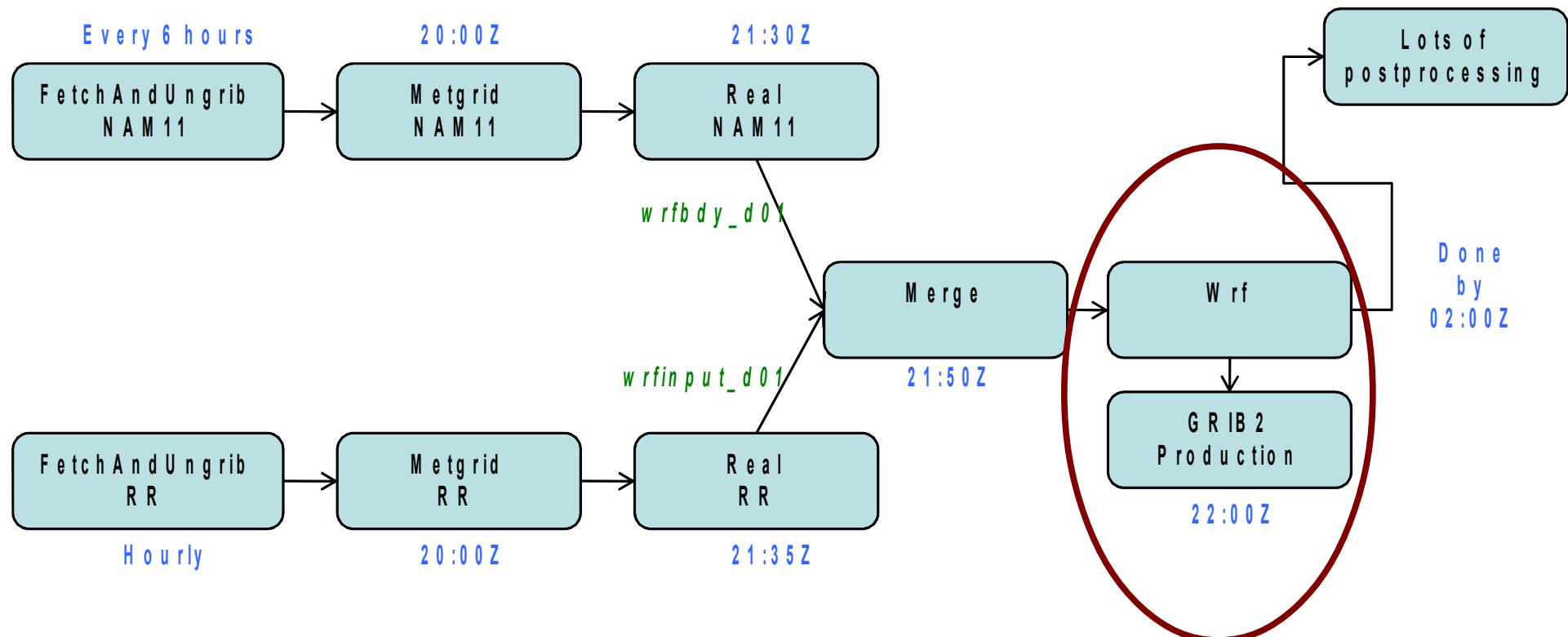


HRRRAK Merge

- Merges the NAM LBC's and the RAP initial conditions
 - We need the outputs of real by about 21:50Z
 - What if they're not available? Wait and try again, but not forever? We need to move with the WRF run
 - What if one is available, especially the NAM? Use its initial condition file instead of the RAP, and run the HRNAMAK instead?



H R R R - A K Process Flow (00Z Forecast)



HRRRAK WRF run and GRIB production

- 00Z WRF run needs its initial and lateral boundary conditions by 22Z. If not available, then we abort, because we don't want to delay the 06Z forecast
- After WRF is launched, we launch processes to monitor the presence of wrfout netCDF files. When the next one is available, numerous processes start up to create GRIB files and push them to an LDM server
- We used to do GRIB production in a single, monolithic program, but performance was horrible. Now, for each forecast hour, we launch 9 GRIB2 production processes, each using 16 cores of supercomputer

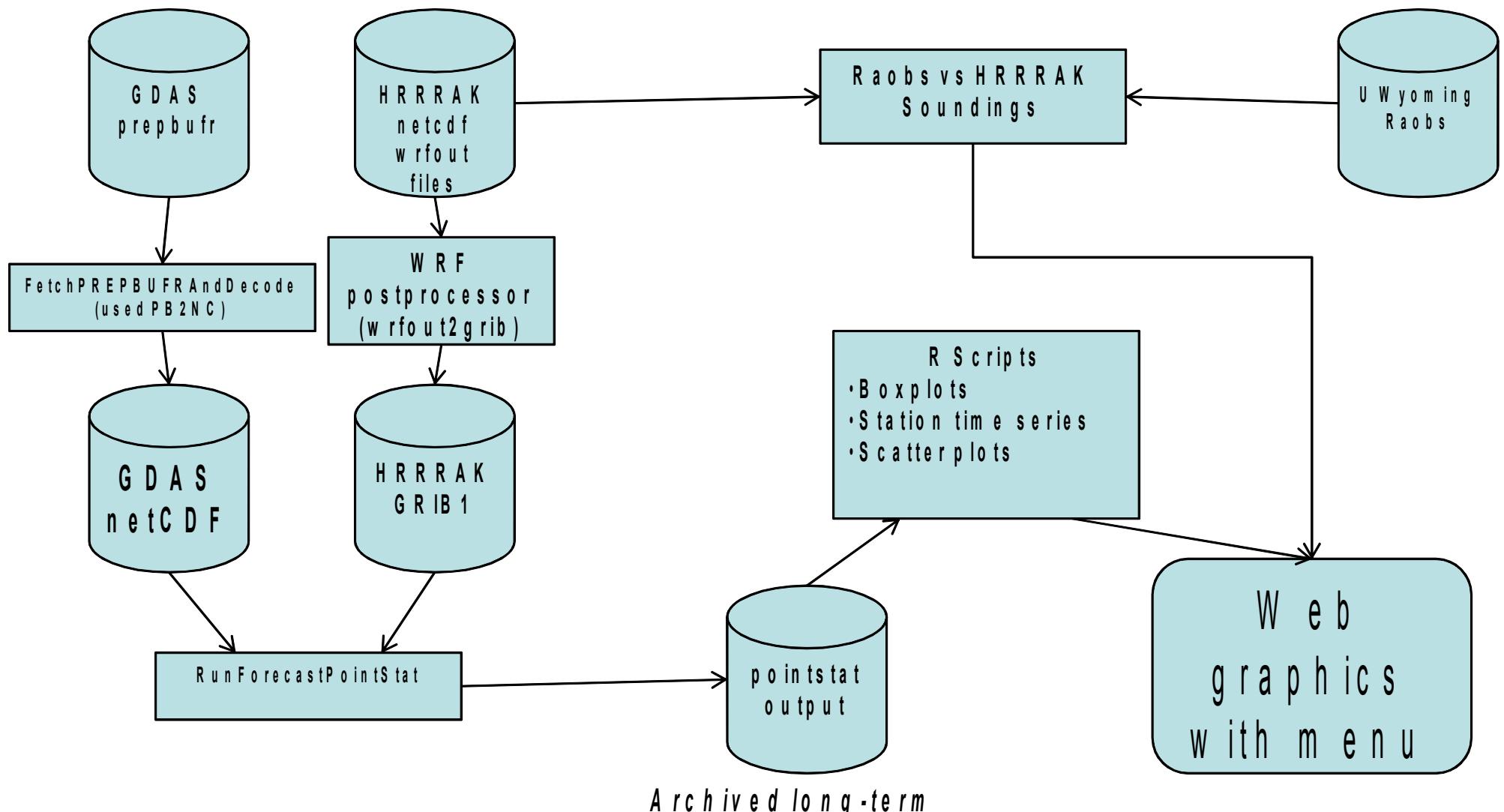


HRRRAK Postprocessing

- Real-time GRIB files for forecast offices
- Close to real-time web products for public and PR purposes
- Forecast verification
 - Downloading of observations
 - Extracting point forecasts from the wrfout files
 - Producing tables and graphics of comparisons
 - Also comparing WRF soundings with raobs
- Weekly, monthly, seasonal verifications
- Archiving



Verification Post Processing



Additional postprocessing

- Now converting wrfout files to ARL format for HYSPLIT (and AER, for time averaged fields) - every forecast
- Cleanup - each forecast uses 3-5 TBytes (3-5 million Gigabytes) of storage
 - Need to purge soon, but not too soon
 - Need to make sure they stay around long enough



Motivations for Python (reference the BAMS)

- BAMS, December 2012, Why Python is the Next Wave in Earth Sciences Computing, Johnny Wei-Bing Lin
- “Python is executable pseudocode” - programs are clear and easy to read
- Interpreted, object-oriented
- Open source
- Very portable



Introduction to Python

- General-purpose, high-level programming language
- Design philosophy emphasises code readability
- Allows programmers to express concepts in fewer lines of code than many other languages like C and Fortran
- Supports object-oriented (e.g. Java), imperative (e.g. Fortran, C) and functional programming (e.g. LISP) styles



Introduction to Python

- With a relatively simple core, designed to be highly extensible
- Core philosophies
 - Beautiful is better than ugly.
 - Explicit is better than implicit.
 - Simple is better than complex.
 - Complex is better than complicated.
 - Readability counts.
- “To describe something as clever is *not* considered a compliment in the Python culture”, Alex Martelli



Introduction to Python

- Conceived in late 1980's with implementation beginning in 1989
- Guido van Rossum is Python's principal author and continues to play a central role. Affectionately referred to in the Python community as Benevolent Dictator for Life
- Name derived from *Monty Python's Flying Circus*
- Python 2.0 released in 2000
 - Last version 2.7
- Python 3.0 released 2008
 - Current production version 3.3.1
 - Development version 3.4



Python 2 or 3?

- Python 3.0 is known as the first ever *intentionally backward incompatible* Python release.
- Python 2 final version is 2.7 with statement of extended support. 2.x branch will see no new major releases.
- Python 3.x under active development, and improvements in libraries won't be reflected back into 2.x
- As a language, Python 3.x is ready, but external library support is still lacking in some areas (e.g. PIL).
- Python 2.x is better supported in IT environments
- Don's opinion - these are days where you might want to consider Python 3, but be prepared for some challenges, particularly in the area of third-party library support.



Python's “Beautiful Heart”

- Reference: *Programming in Python 3: A Complete Introduction to the Python Language*, Mark Summerfield
- Python is incredibly rich in features with many depths of complexity that can seem overwhelming
- However, understanding a few key features allows you to do almost anything you want, and that's where our focus will be
- Additional features will be learned over time as need and interest dictate



Python's “Beautiful Heart”

- Eight key features
 - Data Types
 - Object References
 - Collection Data Types
 - Local Operations
 - Control Flow Statements
 - Arithmetic Operators
 - Input/Output
 - Creating and calling functions



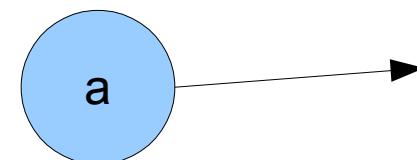
Object Implementation

- Everything in Python is an object
- In short, an object is an entity with data and methods (for accessing the data)
- But, we often use the terms variable and object interchangeably

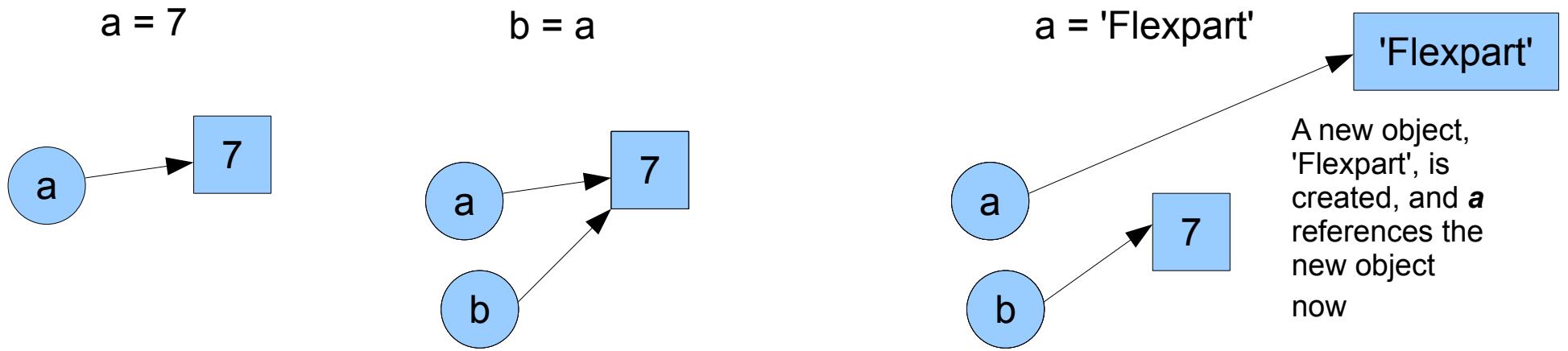


Object Implementation

```
>>> a = 18
>>> dir(a)
['__abs__', '__add__', '__and__', '__class__', '__cmp__', '__coerce__', '__delattr__',
 '__div__', '__divmod__', '__doc__', '__float__', '__floordiv__', '__format__',
 '__getattribute__', '__getnewargs__', '__hash__', '__hex__', '__index__', '__init__',
 '__int__', '__invert__', '__long__', '__lshift__', '__mod__', '__mul__', '__neg__',
 '__new__', '__nonzero__', '__oct__', '__or__', '__pos__', '__pow__', '__radd__',
 '__rand__', '__rdiv__', '__rdivmod__', '__reduce__', '__reduce_ex__', '__repr__',
 '__rfloordiv__', '__rlshift__', '__rmod__', '__rmul__', '__ror__', '__rpow__',
 '__rrshift__', '__rshift__', '__rsub__', '__rtruediv__', '__rxor__', '__setattr__',
 '__sizeof__', '__str__', '__sub__', '__subclasshook__', '__truediv__', '__trunc__',
 '__xor__', 'bit_length', 'conjugate', 'denominator', 'imag', 'numerator', 'real']
>>> x = a.__hex__()
>>> print x
0x12
>>> y = a.__float__()
>>> print y
18.0
>>> z = a.__add__(32)
>>> print z
50
```



Object References



Example of reading in data and storing in collection data types

Reading in real-time fire emissions data (Fire Information for Resource Management System - FIRMS)



Active Fire Data | EOSDI x https://earthdata.nasa.gov/data/near-real-time-data/firms/active-fire-data#tab-content-4

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Near Real-Time Data Land Atmosphere Near Real-time Capability for EOS

Home > Data > Near Real-Time Data > FIRMS > Active Fire Data

Download Active Fire Data

Data for the last 24, 48 hours and 7 days can be downloaded in shapefile, KML, WMS or text file formats.

Use the Archive Download Tool to download older MODIS fire data.

SHP KML WMS TXT Archive Download Tool Global Fire Maps

TXT

Text files, in CSV format, of the active fire data for the last 24 and 48 hours, and 7 days are available below. Daily text files for the last two months are available via FTP. To keep file sizes to a minimum, the data are provided by region. For more information, please see the [TXT README \(PDF\)](#).

Register to start downloading data.

Feedback

Alaska Canada Europe Russia and Asia

Conterminous USA and Hawaii Central America Northern and Central Africa South Asia

South America Southern Africa South East Asia Australia and New Zealand

World Alaska Australia and New Zealand Canada Central America Europe Northern and Central Africa Russia and Asia South America South Asia South East Asia Southern Africa U.S.A. (Conterminous) and Hawaii

	24h	48h	7d
Alaska	24h	48h	7d
Australia and New Zealand	24h	48h	7d
Canada	24h	48h	7d
Central America	24h	48h	7d
Europe	24h	48h	7d
Northern and Central Africa	24h	48h	7d
Russia and Asia	24h	48h	7d
South America	24h	48h	7d
South Asia	24h	48h	7d
South East Asia	24h	48h	7d
Southern Africa	24h	48h	7d
U.S.A. (Conterminous) and Hawaii	24h	48h	7d

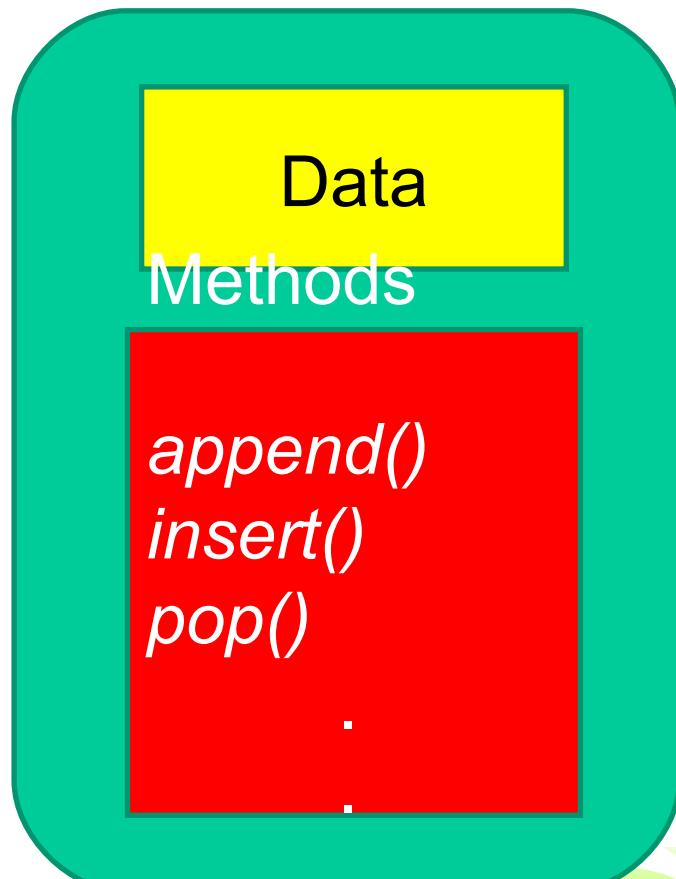
Overview of modules and objects

- The core of Python is relatively small and simple, though you can do many things with just this much.
- Most of the power of the Python environment comes from its standard library in the form of modules, 3rd party packages and modules, and your own user-defined modules



Concepts of Object-Oriented Programming

Abstract *list()* class



objects from the *list()* class.

- These objects can be viewed as a black box
 - Details hidden from user
 - Well-defined and controlled access to the data, via methods

Concepts of Object-Oriented Programming

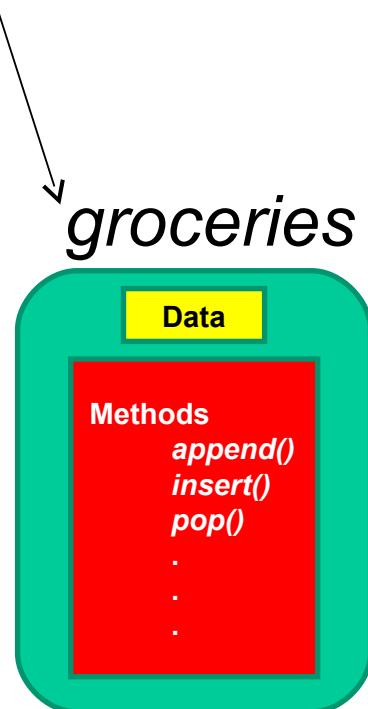
- We can create instantiations of the list class

```
groceries = list()
```

```
groceries.append('pizza')
```

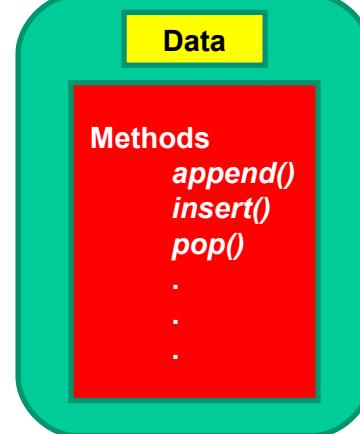
```
groceries.append('beer')
```

Data section
contains
'pizza' and
'beer'



```
waitlist = ['Kim', 'Eric', 'Andrew', 'Nell']
```

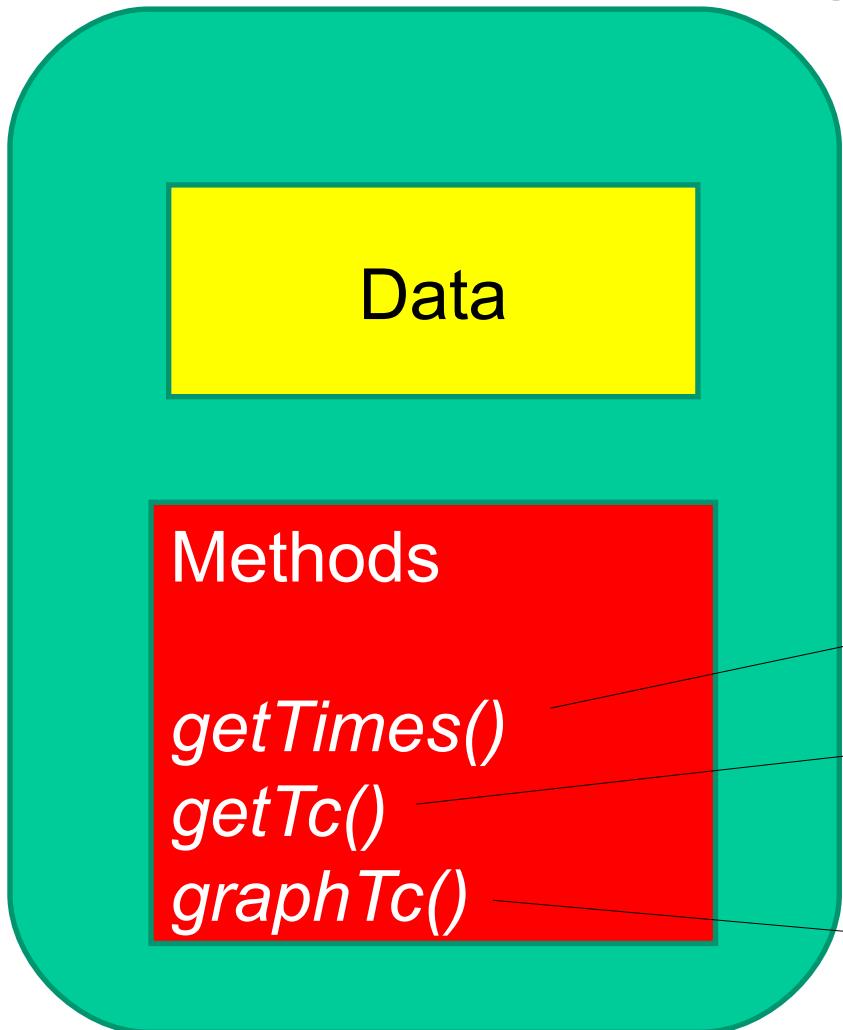
waitlist



Data section
contains 'Kim'
'Eric'
'Andrew'
'Nell'

Concepts of Object-Oriented Programming

Abstract *MetFetcher()* class



- Here is an abstract representation of a simple class that I created, called *MetFetcher()*
- *MetFetcher* serves as a *black box*, getting recent meteorological observations from a specified weather station.

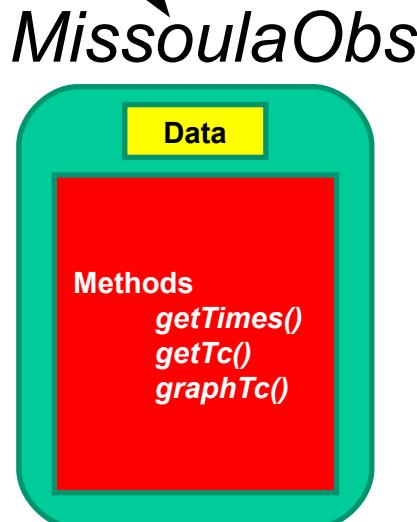
Concepts of Object-Oriented Programming

- MetFetcher serves as a black box, getting recent meteorological observations from a specified weather station.
- You can create instances of the class, as follows

MissoulaObs = MetFetcher("KMSO")

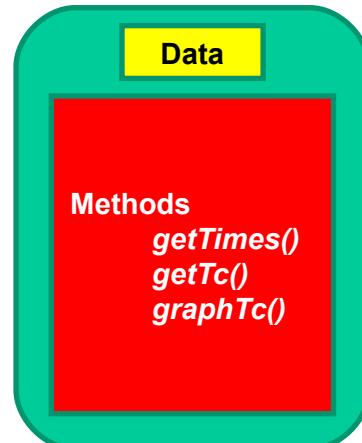
FairbanksObs = MetFetcher("PAFA")

Data section
contains met
data for
Missoula



FairbanksObs

Data section
contains met
data for
Fairbanks



Concepts of Object-Oriented Programming

MissoulaObs() object

Data

Methods

getTimes()
getTc()
graphTc()

This is what some of the raw data looks like inside the object, but you don't care about that. You only care about using the supplied methods to get certain data out in a certain format

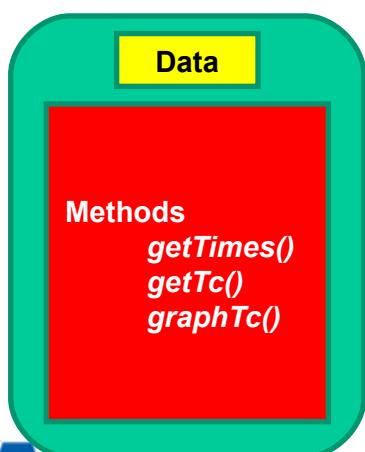
```
[<pre>StnId,Lat,Lon,Elev,Time,Tmp,Dwpt,RH,WndDir,WndSpd,WndGst,Vsby,WX,WX,Clds,SLP,AI  
tim,StnPrs,Pcpn1hr,Pcpn3hr,Pcpn6hr,Pcpn24hr,6hrMaxT,6hrMinT,24hrMaxT,24hrMinT,QC,\n',  
'KMSO,46.92,-  
114.09,3199,200809051853,55,41,59,210,3,,10.00,Clear,,OVC060,1020.9,30.15,26.837,T,,,,,OK,\n',  
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'KMSO,46.92,-114.09,3199,200809060353,52,45,77,320,3,,1
```

Concepts of Object-Oriented Programming

Sequence of Python statements

```
MissoulaObs = MetFetcher("KMSO")  
  
theTimes = MissoulaObs.getTimes()  
theTemps = MissoulaObs.getTc()  
  
print theTimes  
print theTemps
```

MissoulaObs



Output

A screenshot of a Windows-style Python Shell window titled "Python Shell". The window shows the command-line input and its corresponding output. The output consists of two large lists of numerical data, representing the times and temperatures collected by the MissoulaObs object.

```
[200809051853, '200809051953', '200809052053', '200809052153', '200809052253',  
'200809052353', '200809060053', '200809060153', '200809060253', '200809060353',  
'200809060453', '200809060553', '200809060653', '200809060753', '200809060853',  
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[12.777777777777779, 12.777777777777779, 13.33333333333334, 12.777777777777779,  
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10.5555555555555555, 10.0, 7.777777777777786, 7.777777777777786, 8.333333333333  
33339, 7.777777777777786, 6.66666666666667, 7.22222222222223, 9.444444444444  
4446, 11.66666666666668, 13.88888888888889, 16.11111111111111, 17.777777777777  
7779, 19.44444444444446, 19.44444444444446, 19.44444444444446, 19.444444444444  
4446, 13.88888888888889, 14.44444444444445, 15.0, 15.0, 15.0, 14.444444444444  
45, 13.88888888888889, 12.77777777777779, 13.33333333333334, 12.222222222222  
23, 12.22222222222223, 12.22222222222223, 12.22222222222223, 12.222222222222  
23, 11.11111111111111, 11.11111111111111, 10.5555555555555555, 11.11111111111111  
11, 11.11111111111111, 11.11111111111111, 11.11111111111111, 10.0, 10.0, 10.0  
, 10.0, 10.0, 11.11111111111111, 12.77777777777779, 13.33333333333334, 14.44  
444444444445]  
>>>
```

Concepts of Object-Oriented Programming

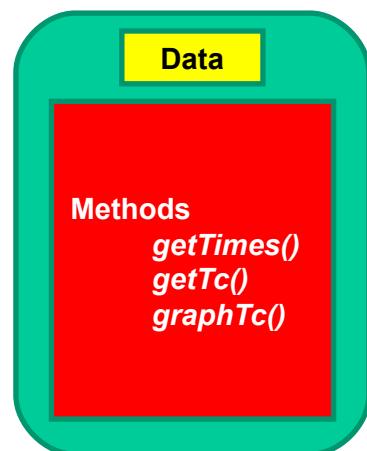
Sequence of Python statements

```
FairbanksObs = MetFetcher("PAFA")
```

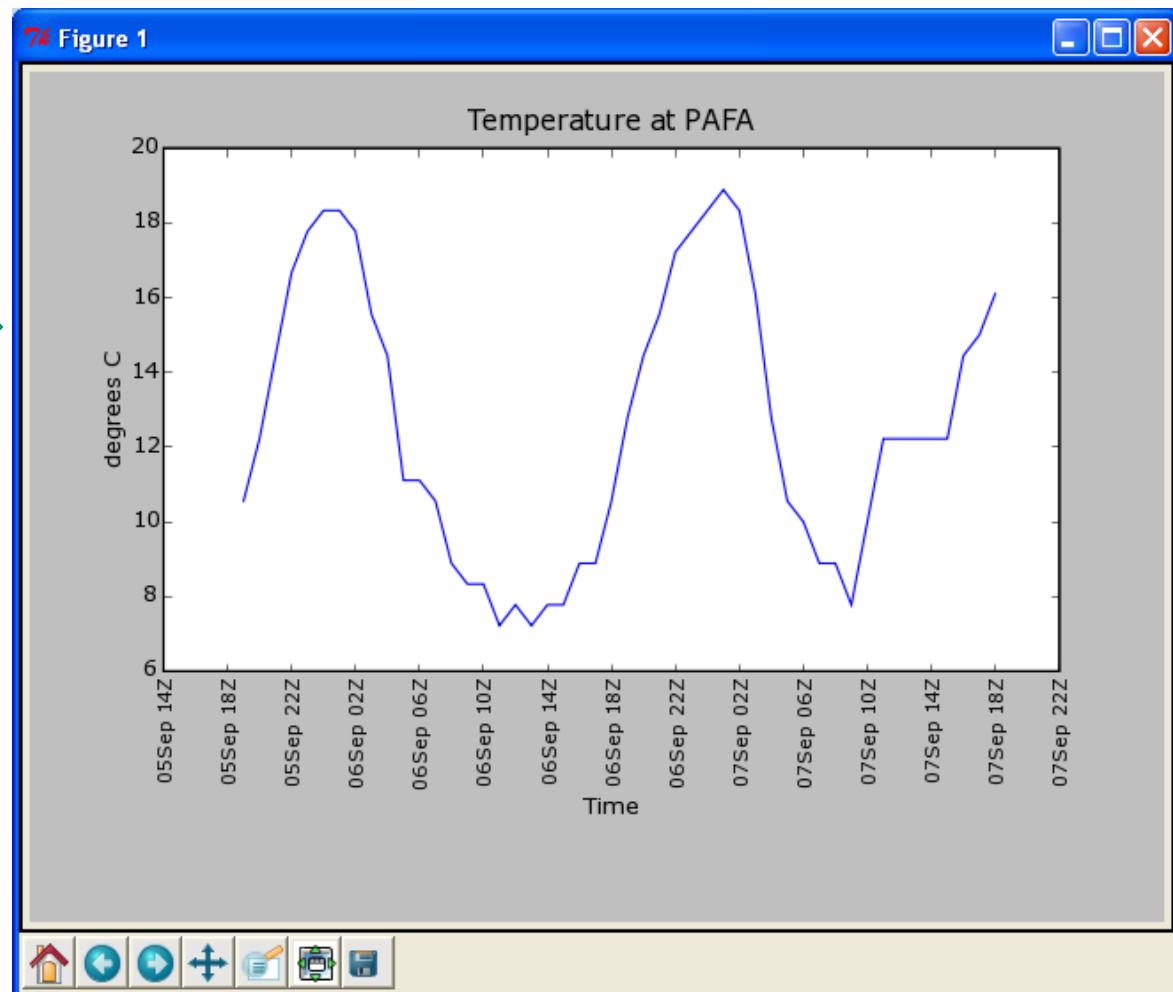
```
FairbanksObs.graphTc()
```



FairbanksObs



Output



Preparing for next lab

(From operational Python workshop, April 2012, Vienna, Austria)

- I have prepared a class, *FIRMS*, for retrieving satellite data to be used as fire emissions
- You will import the class, study it, use it, and make minor enhancements to it
- This class will be a part of the overall FLEXPART operational system



FIRMS Data Access page

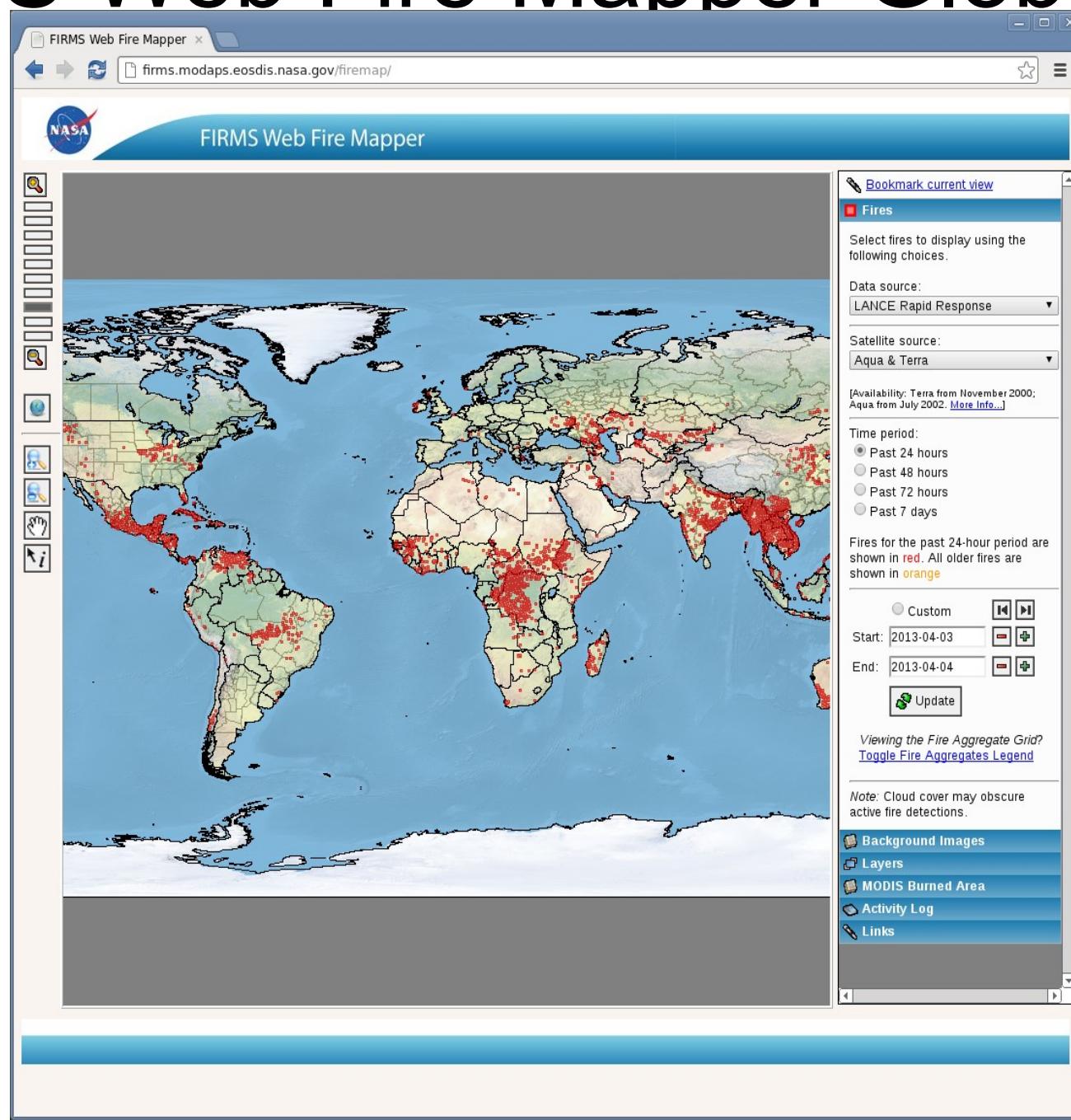
FIRMS: Fire Information for Resource Management System

Delivers MODIS hotspots / fire information in easy to use formats.

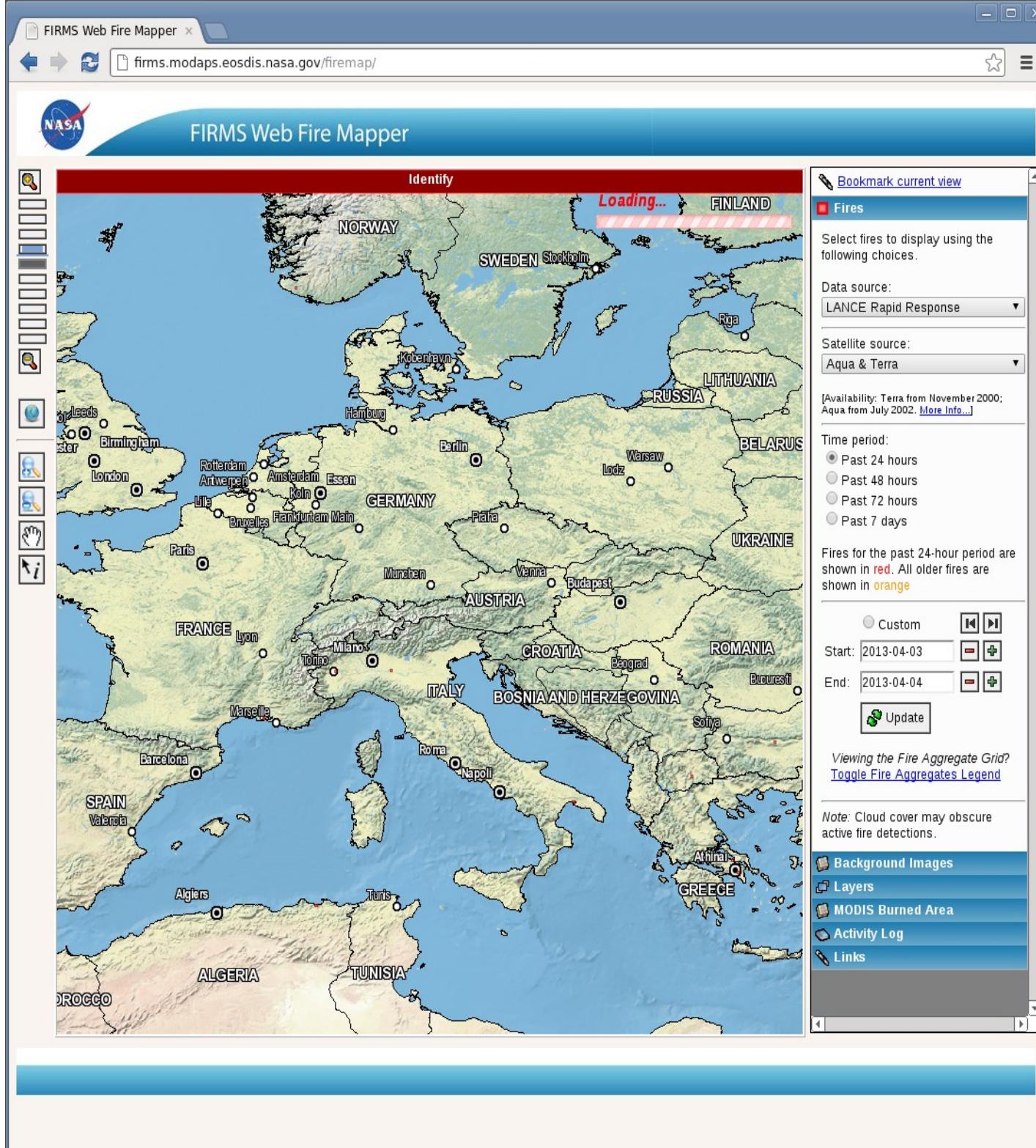
The screenshot shows the EOSDIS Near Real-Time Data FIRMS Active Fire Data page. The left sidebar has a 'Feedback' button. The main content area shows a map of the world with regions labeled: Alaska, Conterminous USA and Hawaii, Canada, Europe, Russia and Asia, Central America, Northern and Central Africa, South Asia, South East Asia, Southern Africa, Australia and New Zealand, and South America. Below the map is a table of download options for different regions and the world.

Region	24h	48h	7d
World	24h	48h	7d
Alaska	24h	48h	7d
Australia and New Zealand	24h	48h	7d
Canada	24h	48h	7d
Central America	24h	48h	7d
Europe	24h	48h	7d
Northern and Central Africa	24h	48h	7d
Russia and Asia	24h	48h	7d
South America	24h	48h	7d
South Asia	24h	48h	7d
South East Asia	24h	48h	7d
Southern Africa	24h	48h	7d
U.S.A. (Conterminous) and Hawaii	24h	48h	7d

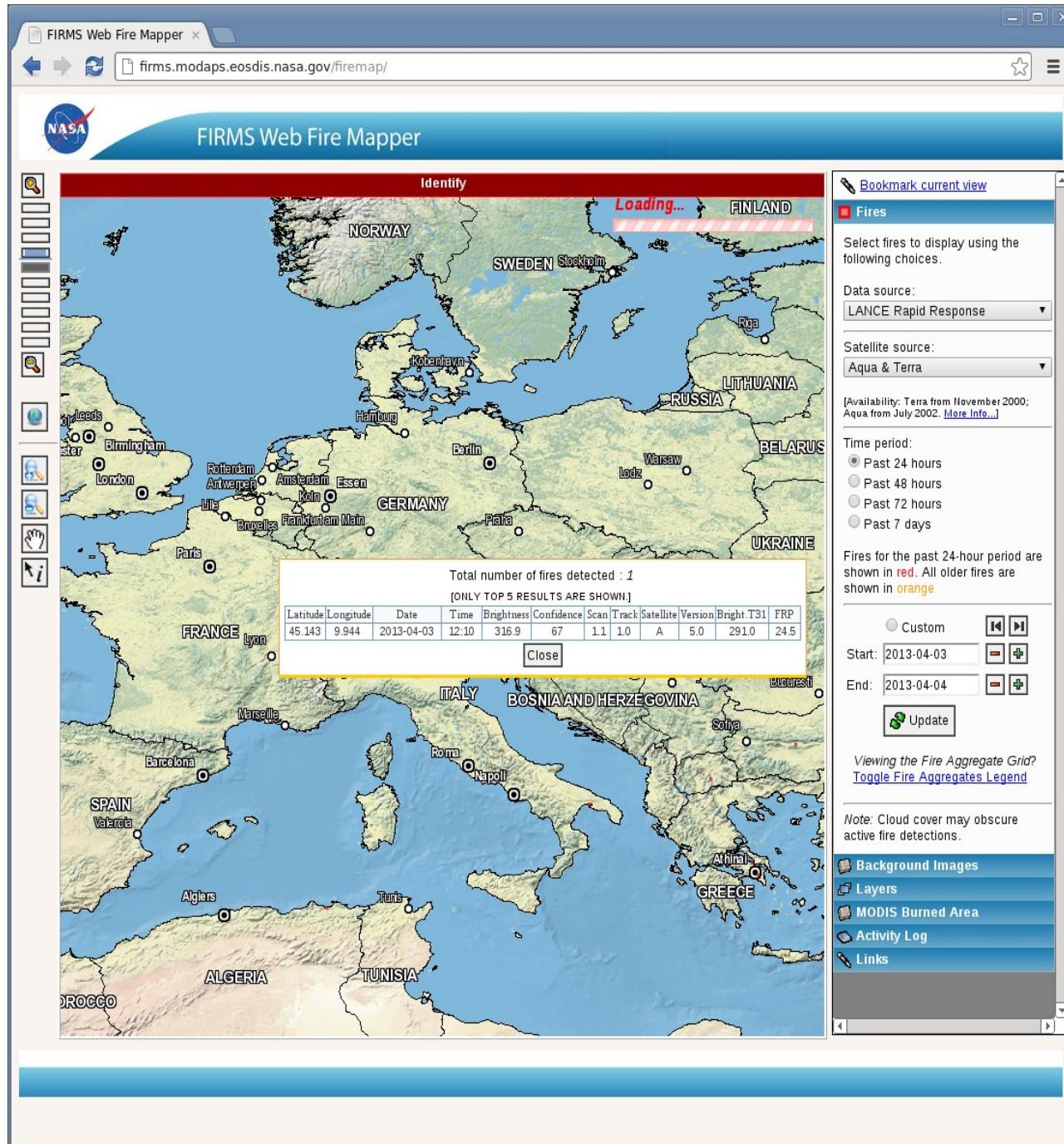
FIRMS Web Fire Mapper Global



Web Fire Mapper Zoom EU



Web Fire Mapper Zoom EU Italy Info



FIRMS Raw EU Data

```
morton@localhost:~/Administrative/ZAMGCourse2013/LabExerciseData/FIRMS
latitude,longitude,brightness,scan,track,acq_date,acq_time,satellite,confidence,version,bright_t30,frp
40.513,17.215,305,2,1,1,1,2013-04-03,0105,A,22,5.0,.283,2,16,7
36.88,6.953,309,3,1,8,1,3,2013-04-03,0105,A,65,5.0,.283,9,46,3
49.786,32.784,312,4,1,1,1,2013-04-03,0845,T,54,5.0,.279,7,22,3
47.859,33.448,315,3,1,1,2013-04-03,0845,T,54,5.0,.292,5,16,3
36.879,6.943,341,8,1,1,2013-04-03,1030,T,69,5.0,.297,9,53
36.877,6.954,333,8,1,1,2013-04-03,1030,T,65,5.0,.298,2,38,6
47.676,34.736,313,9,1,1,2013-04-03,1030,A,59,5.0,.294,8,12,9
47.627,34.32,313,2,1,1,1,2013-04-03,1030,A,57,5.0,.294,3,12,8
48.81,32.536,321,2,1,1,1,2013-04-03,1030,A,74,5.0,.289,7,31,5
49.89,34.229,314,2,1,1,2013-04-03,1030,A,60,5.0,.286,19,2
43.434,4.887,310,3,2,1,1,4,2013-04-03,1210,A,31,5.0,.296,3,32,1
45.143,9.944,316,9,1,1,1,2013-04-03,1210,A,67,5.0,.291,24,5
45.106,7.725,311,1,1,3,1,1,2013-04-03,1210,A,17,5.0,.293,6,14,5
51.031,2.27,322,7,1,7,1,3,2013-04-03,1210,A,57,5.0,.283,3,64,9
57.316,-5.573,379,6,1,2,1,1,2013-04-03,1200,T,87,5.0,.280,4,212,6
56.879,-5.52,310,4,1,2,1,1,2013-04-03,1200,T,38,5.0,.288,3,22,7
54.091,-8.969,313,7,1,1,1,2013-04-03,1205,T,59,5.0,.282,7,23,8
36.869,6.951,316,5,2,3,1,5,2013-04-03,1205,A,66,5.0,.299,1,49,5
36.878,6.943,326,7,2,3,1,5,2013-04-03,1205,A,0,5.0,.295,2,93,9
52.134,-9.446,331,1,1,3,1,1,2013-04-03,1350,A,84,5.0,.284,8,61,8
53.356,-9.302,311,3,1,4,1,2,2013-04-03,1350,A,48,5.0,.284,4,26,8
54.082,-8.977,314,6,1,4,1,2,2013-04-03,1350,A,61,5.0,.282,5,35,7
```

1.1

Top

We can download raw text data for past 24 hours, and for different regions - e.g.
http://firms.modaps.eosdis.nasa.gov/active_fire/text/Europe_24h.csv



Use of FIRMS data

- We will use FIRMS data - in a tutorial setting - to estimate fire emissions. Simplifying assumptions, for academic purposes
 - Just to get relative emissions values, we will multiply a pixel's Fire Radiance Power (FRP) to represent the mass emissions over the duration of the FLEXPART simulation
 - We assume that FRP is constant throughout the simulation
 - Though the data contains confidence measures, we will assume all data is good



Use of FIRMS data

- The approach - use a Python class, FIRMS, in which to store and access this data.

MODIS/Active fire data

Methods:

FIRMS(sourceOfData)
printFullListing()
setRegion(boundingBox)
printRegionalListing()
getNumRegionalReleases()
getRegionalData()
plotRegion(imageFilename)



Sample program using FIRMS.py

```
#!/usr/bin/env python

import FIRMS

# European source data
EU_SOURCE_DATA =
'http://firms.modaps.eosdis.nasa.gov/active_fire/text/Europe_24h.csv'

# Alaskan source data
AK_SOURCE_DATA =
'http://firms.modaps.eosdis.nasa.gov/active_fire/text/Alaska_24h.csv'

# Instantiate an object with Europe fire data - this will automatically go
# to the source, retrieve, and store the data
EUDat = FIRMS.FIRMS(EU_SOURCE_DATA)

# Instantiate an object with Alaska fire data
AlaskaDat = FIRMS.FIRMS(AK_SOURCE_DATA)
```



Sample program using FIRMS.py

```
# Print the full listing of the Alaska data
AlaskaDat.printFullListing()

# Create a bounding box (lower left and upper right corners) for EU region
llLat = 40.0; llLon = 0.0; urLat = 55.0; urLon = 20.0

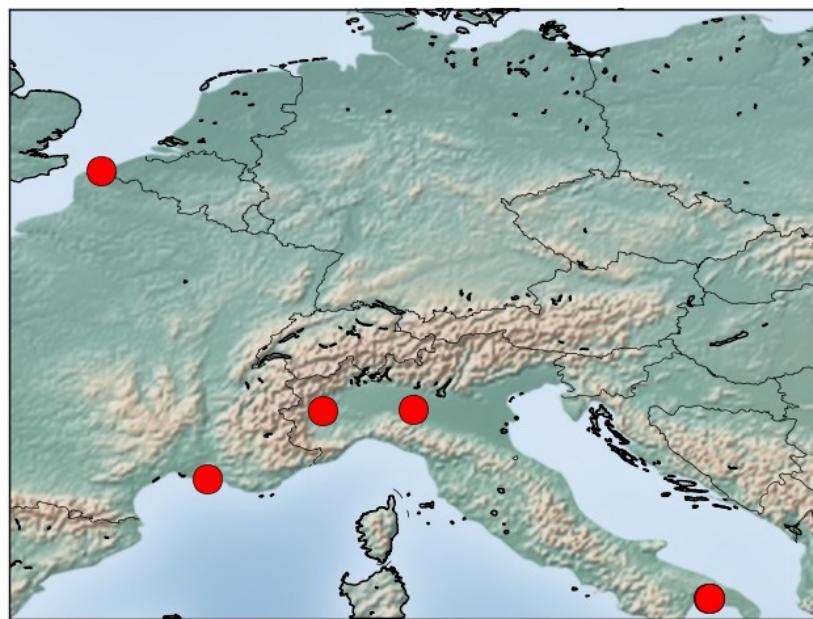
# Store this in a tuple (LL_LAT, LL_LON, UR_LAT, UR_LON)
boundingBox = (llLat, llLon, urLat, urLon)

# Set region, then print listing
EUDat.setRegion(boundingBox)
EUDat.printRegionalListing()

# Plot the region
EUDat.plotRegion('EUFires.png')
```



FIRMS EU BB (40,0,55,20)



Miscellaneous Python Solutions

- Submission of jobs into batch queue systems
- Image production with NCL
- Python Numerical and visualisation capabilities
- File retrievals – timeouts and retries
- Process logging



Abstracting batch queue submission for portability

```
import BatchJob

.
.

def runMetgrid(metgridDir, numTasks, queueName, groupName, wallMinutes, binaryPath):
    print 'in runMetgrid()...'
    os.chdir( metgridDir )

    theJob = BatchJob.BatchJob('metgridrun', 'pacman_4')

    theJob.setCores(numTasks)
    theJob.setQueue(queueName)
    theJob.setGroup(groupName)
    theJob.setWallTimeMinutes(wallMinutes)
    theJob.setRunCommand(binaryPath)

    theStatements = []
    theStatements.append('echo \'Start: \' `date`')
    theJob.setPreJobStatements(theStatements)

    theStatements = []
    theStatements.append('echo \'Finish: \' `date`')
    theJob.setPostJobStatements(theStatements)

    theJob.blockingSubmit()
```



Implementation of blocking submit

```
def blockingSubmit(self):
    print 'In blockingSubmit()'
    print '.... creating the batch script...'
    if self._systemName in ['pacman_4']:
        self._createBatchScript_pacman_4()
        jobID = str( self._submitJob_pacman_4() )
        self._jobSubmitTime = time.time()
        print 'Submitted Job: ' + jobID
        # Sleep for 10 seconds, just giving job plenty of time to
        # be viewable in qstat
        time.sleep(10)

        # Wait for job to finish, or timeout
        # As long as status is 'Q' or 'R', keep waiting, unless timeout
    batchJobDone = False
    while not batchJobDone:
        jobStatus = self._jobStatus_pacman_4(jobID)
        #print 'jobStatus: ' + jobStatus
        if jobStatus in ['Q', 'R']:
            time.sleep(self._blockingSleepTimeMinutes*60)
            if self._isTimedOut():
                print __name__ + '.blockingSubmit(): block timed out...'
                batchJobDone = True
        else:
            batchJobDone = True

    elif self._systemName in ['tana']:
        ...
```



PBS Batch Files

pacman – Penguin computing cluster

```
#PBS -q arscwthr
#PBS -l nodes=4:ppn=4
#PBS -l walltime=00:20:00
#PBS -j oe

cd $PBS_O_WORKDIR
echo 'Start: ' `date`
mpirun -np 16 ./metgrid.exe
echo 'Finish: ' `date`
```

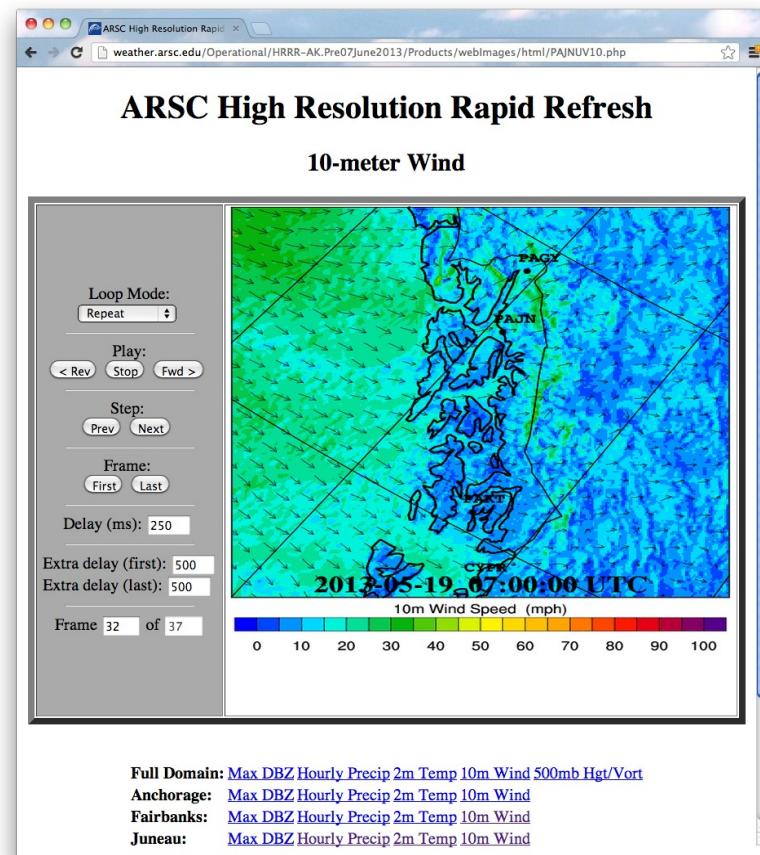
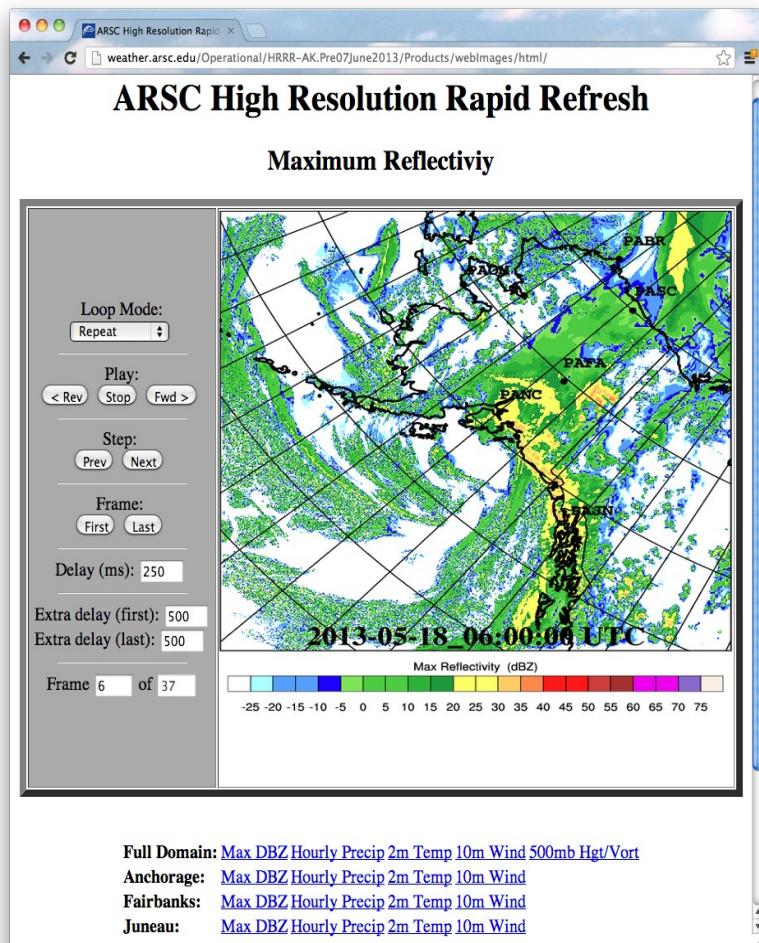
tana – Cray XE6

```
#PBS -q standard
#PBS -l mppwidth=16
#PBS -l walltime=00:20:00
#PBS -W block=true
#PBS -j oe

cd $PBS_O_WORKDIR
echo 'Start: ' `date`
aprun -n 16 ./metgrid.exe
echo 'Finish: ' `date`
```



Image Production



```
[wrfuser@pacman1:~/WX/Operational/HRRR-AK/PostProcScripts/gifImages] $ ls
CommonDefs.py      fullDBZ.ncl          fullUV10.ncl    localHourlyPrecip.ncl
Common.py          fullHourlyPrecip.ncl   gifDriver.py   localT2.ncl
full1500mb.ncl    fullT2.ncl          localDBZ.ncl   localUV10.ncl
```



localT2.ncl excerpt

```
;;;;;; These are used for operational - comment out when testing ;;;;;;  
wrfoutFile = getenv("wrfoutFile") ; Full path to the wrfout file  
imageName = getenv("imageName") ; File name of image file (without ext)  
REGION_NAME = getenv("regionName")  
;;;;;;;;;;;;;;  
  
;; Set up the locale-specific parameters  
if (REGION_NAME .eq. "PAJN") then  
    ; Zoom box  
    LL_LAT = 52.0  
    LL_LON = -135.0  
    UR_LAT = 62.0  
    UR_LON = -132.0  
  
    LOCATIONS = (/"PAJN", "CYPR", "PAGY", "PAKT"/)  
    LATS =      (/58.35, 54.30, 59.46, 55.36/)  
    LONS =      (/−134.58, −130.43, −135.32, −131.71/)  
  
else  
    if (REGION_NAME .eq. "PAFA") then  
        .  
        .  
        .
```



```

def createImages(field, domainList, NCL_SCRIPT_DIR, wrfoutFileList,
                WRFOUT_DIR, nestNum, CONVERT_BIN, ARCHIVE_PRODUCT_DIR,
                MOBILE_ANIMATED_GIF_DIR, WEB_GIF_DIR, wrkdirname):

    for domainName in domainList:
        plotName = domainName + field
        os.environ['regionName'] = domainName

        if domainName == 'Full':
            nclScriptName = 'full' + field + '.ncl'
        else:
            nclScriptName = 'local' + field + '.ncl'

        nclScript = NCL_SCRIPT_DIR + '/' + nclScriptName
        for file in wrfoutFileList:

            # Call NCL script to produce eps file
            os.environ['wrfoutFile'] = WRFOUT_DIR + '/' + \
                                      file + '.nc'

            imageFileName = 'image-' + plotName + '-d0' + \
                           str(nestNum) + '-' + str(imageNum).zfill(3)

            os.environ['imageName'] = imageFileName

            print 'producing: ' + imageFileName

            theCommand = NCL + ' ' + nclScript
            print 'executing: ' + theCommand
            os.system(theCommand)

            # Create the GIF
            outputFilename = imageFileName + '.gif'
            gifCommand = CONVERT_BIN + \
                         ' -compress LZW ' + \
                         ' -geometry 100%x100% ' + \
                         ' -background white '

            gifCommand = gifCommand + imageFileName + '.eps ' + \
                         outputFilename
            print 'executing: ' + gifCommand
            os.system(gifCommand)

            ImageNum += 1

```



Python Numerical and Visualisation

- Python is an interpreted language with little emphasis on optimisation, especially for high performance applications
- However, numerous external libraries are available to expand Python's reach into the HPC community
- The lineage of scientific libraries is confusing

Python Scientific Libraries (History)

- *Numeric*, a Python array package originated in 1995 to improve Python for array operations. Reasonably complete and stable, but now obsolete
- *Numarray* was a complete rewrite of *Numeric*, but is also deprecated now (last release 2006)
- SciPy, in 2005 had a subproject to take the best of Numeric and Numarray. This was separated and called NumPy
- NumPy compatible with Python versions 2.4-2.7 and 3.1+

Numerical Python (NumPy)

- Working with arrays and matrices in a natural way
- Package contains large list of mathematical functions
- If available, LAPACK is used for linear algebra routines, enhancing performance.
- Originally part of SciPy, later separated and used by SciPy for array and matrix processing

NumPy - www.numpy.org

- NumPy code is cleaner than native Python code for numerical operations
 - Many operations work on entire matrices, requiring fewer loops
 - Many of the algorithms are mature
- Arrays stored more efficiently than in base Python (which would use a list of lists).
- Performance scales with number of elements in arrays

Python versus NumPy example

- Initialisation and summation of two vectors

Python

```
def pythonsum(n):  
    a = range(n)  
    b = range(n)  
    c = []  
  
    for i in range( len(a) ):  
        a[i] = i**2  
        b[i] = i**3  
        c.append( a[i] + b[i] )  
  
    return c
```

NumPy

```
def numpysum(n):  
    import numpy  
  
    a = numpy.arange(n)**2  
    b = numpy.arange(n)**3  
    c = a + b  
  
    return c
```

seconds

N	Python	NumPy
1000	707	171
4000	2829	274

(NumPy 1.5 Beginner's Guide, Ivan Idris, 2011)

Simple NumPy matrix multiplication example

- Create 2 1x4 float arrays
- Reshape to 2x2
- Matrix multiplication

```
import numpy as np

A = np.array( np.arange(1,5), float )
B = np.array( np.arange(5,9), float )

A = A.reshape(2,2)
B = B.reshape(2,2)

print 'A: ' + str(A)
print 'B: ' + str(B)

C = np.dot(A,B)

print 'C: ' + str(C)
```

→

A: [[1. 2.]
[3. 4.]]
B: [[5. 6.]
[7. 8.]]
C: [[19. 22.]
[43. 50.]]

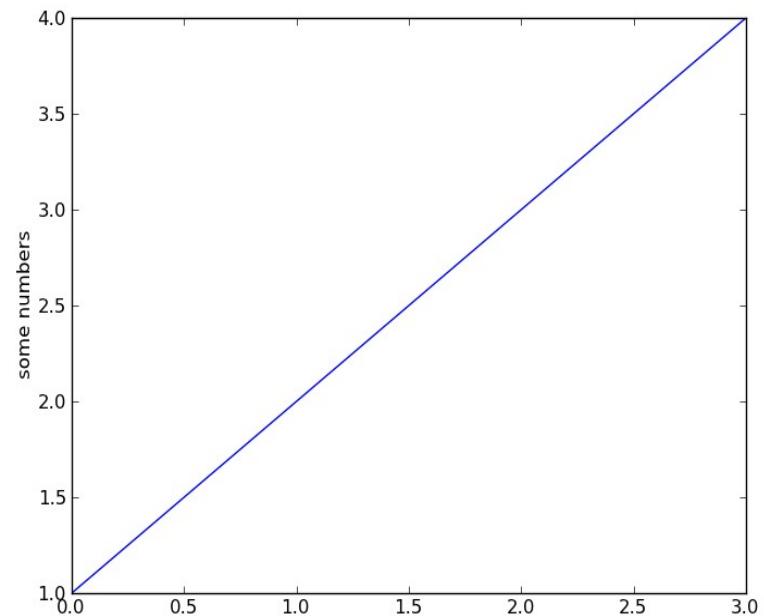
Matplotlib - matplotlib.org

- Library for making 2D plots of arrays in Python
- Origins in emulating MATLAB graphics commands, but is independent of MATLAB, and can be used in a Pythonic, object oriented framework
- Makes heavy use of NumPy for high performance

matplotlib.pyplot

- Collection of command-style functions that make matplotlib work like MATLAB
- Stateful - keeps track of current figure and plotting area

```
import matplotlib.pyplot as plt  
  
plt.plot([1,2,3,4])  
plt.ylabel('some numbers')  
plt.show()
```



Basemap toolkit - matplotlib.org/basemap

- Library for plotting 2D data on maps in Python
- Similar to MATLAB mapping toolbox, and others
- Does no plotting on its own. Transforms coordinates to one of many provided map projections. Matplotlib then plots contours, etc. in the transformed coordinates

Basemap example - just plotting a map

```
#!/usr/bin/env python

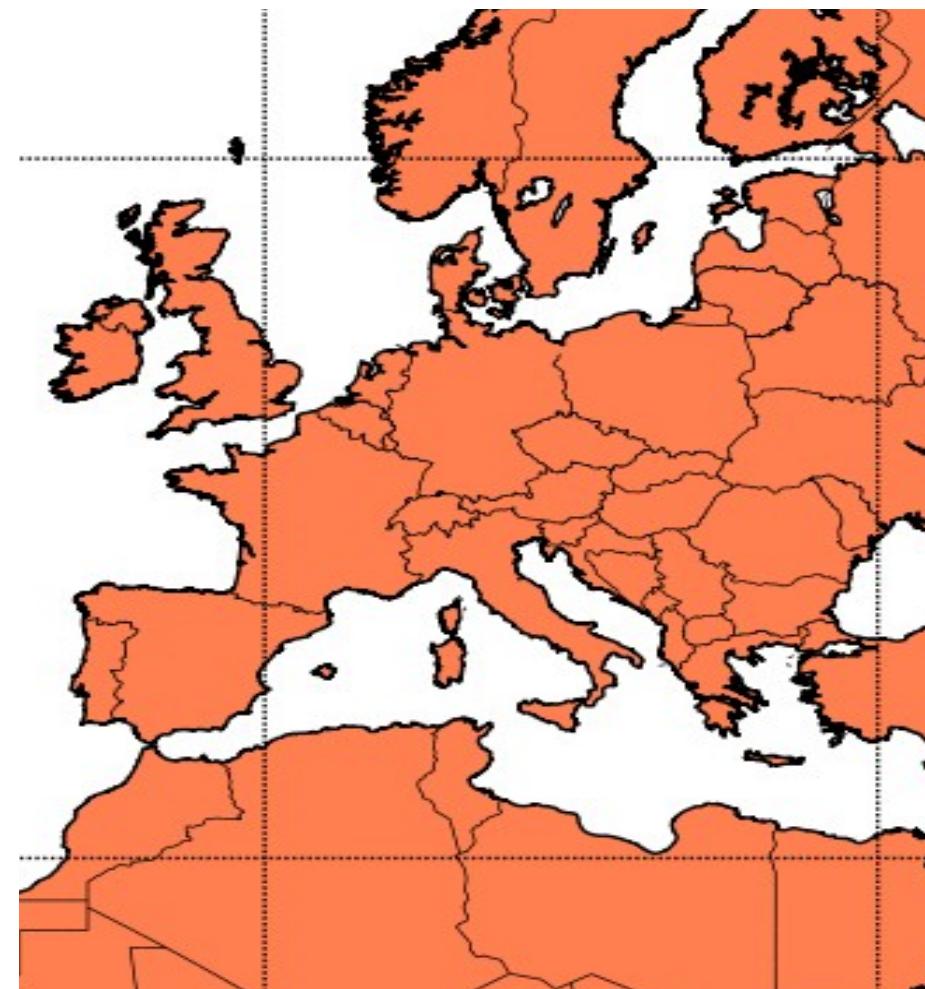
import matplotlib as mpl
mpl.use('Agg')
from mpl_toolkits.basemap import Basemap, cm
import matplotlib.pyplot as plt
import numpy as np

map = Basemap(projection='merc',
    resolution = 'h', area_thresh = 1000.0,
    llcrnrlon=-20.0, llcrnrlat=20,
    urcrnrlon=40, urcrnrlat=65)

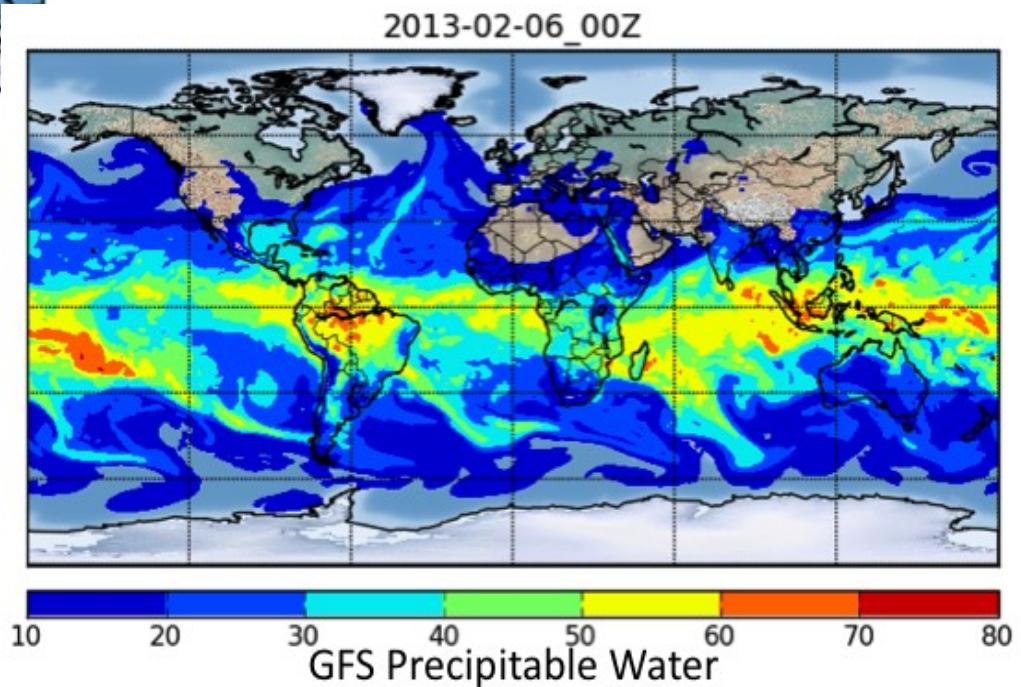
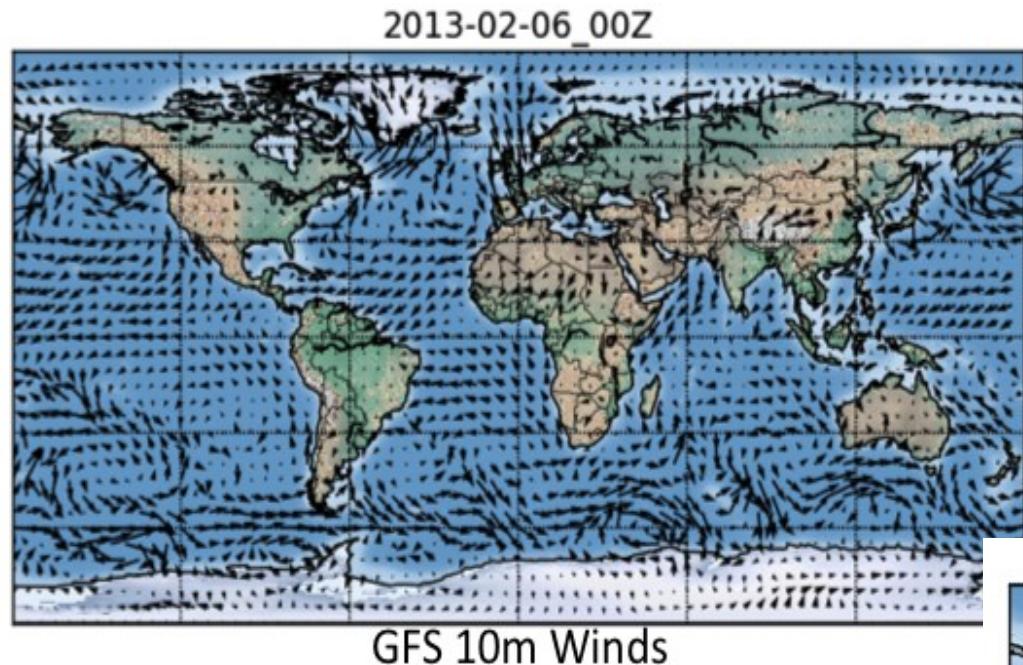
map.drawcoastlines()
map.drawcountries()
map.fillcontinents(color='coral')
map.drawmapboundary()

map.drawmeridians(np.arange(0, 360, 30))
map.drawparallels(np.arange(-90, 90, 30))

plt.savefig("out.png")
plt.close()
```



Matplotlib GFS plotting



Plotting GFS winds with matplotlib

```
#!/usr/bin/env python

import os
import sys

import gribapi

import matplotlib as mpl
mpl.use('Agg')
import matplotlib.pyplot as plt
import pylab as pl
from mpl_toolkits.basemap import Basemap, cm
import numpy as np

OUTPUT_FILENAME = 'quickGFSWind.png'

def main(argv=sys.argv):

    if len(argv) != 2:
        print 'Usage: gfsWindVis.py <GFS File Path>'
        sys.exit()
    else:
        MET_INPUT_FILE = argv[1]

    if not os.path.isfile(MET_INPUT_FILE):
        print 'Unable to find MET_INPUT_FILE: ' + MET_INPUT_FILE
        sys.exit()

    FH = open(MET_INPUT_FILE)

    # Turn on the multi-field support
    gribapi.grib_multi_support_on()
```



```

while 1:
    gid = gribapi.grib_new_from_file(FH)
    if gid is None: break

    shortName = gribapi.grib_get(gid, 'shortName')

    if shortName == '10u':
        print 'U WIND!!!'
        Ni = int( gribapi.grib_get(gid, 'Ni') )
        Nj = int( gribapi.grib_get(gid, 'Nj') )
        dataDate = gribapi.grib_get(gid, 'dataDate')
        dataTime = gribapi.grib_get(gid, 'dataTime')
        fcstTime = gribapi.grib_get(gid, 'forecastTime')
        latLL = gribapi.grib_get(gid, 'latitudeOfFirstGridPointInDegrees')
        lonLL = gribapi.grib_get(gid, 'longitudeOfFirstGridPointInDegrees')
        latUR = gribapi.grib_get(gid, 'latitudeOfLastGridPointInDegrees')
        lonUR = gribapi.grib_get(gid, 'longitudeOfLastGridPointInDegrees')

        # Store this stuff in a dictionary so it's easy to pass to
        # a function as an arg
        gridInfoDict = { \
            'LL_LAT' : latLL, \
            'LL_LON' : lonLL, \
            'UR_LAT' : latUR, \
            'UR_LON' : lonUR, \
            'NI' : Ni, \
            'NJ' : Nj \
        }

        print "Grid info: %d %d (%f/%f) (%f/%f) " % \
            (Ni, Nj, latLL, lonLL, latUR, lonUR)

        theVarGrid = gribapi.grib_get_values(gid)
        theUGrid = np.reshape(theVarGrid, (Nj, Ni) )
        theUGrid = np.flipud(theUGrid)
    # END if 10u

```



```

if shortName == '10v':
    print 'V WIND!!!!'
    theVarGrid = gribapi.grib_get_values(gid)
    theVGrid = np.reshape(theVarGrid, (Nj, Ni) )
    theVGrid = np.flipud(theVGrid)
# END if 10v

# END while 1

dataDateStr = '%08d' % dataDate
dataTimeStr = '%02d' % dataTime
fcastTimeStr = '%02d' % fcastTime
timeStampTitle = dataDateStr[0:4] + '-' + \
                 dataDateStr[4:6] + '-' + \
                 dataDateStr[6:8] + ' ' + \
                 dataTimeStr + 'Z' + '-' + fcastTimeStr
createWindFrame(gridInfoDict, theUGrid, theVGrid, \
                timeStampTitle, OUTPUT_FILENAME)

# END main()

```



```

def createWindFrame(gridInfo, theUGrid, theVGrid, timeStampTitle, \
                   fullFilename):

    # Creates a single image snapshot (currently defaults to PNG)

    #----- arguments -----
    # gridInfo:      a dictionary containing grid info
    # theUGrid, theVGrid:      2D horizontal grids of the data to be plotted
    # timeStamp:      YYYYMMDDHH timestamp associated with the data
    # fullFilename: full path and filename of image (without the extension)

    # Define the type and boundaries of this map
    m = Basemap(projection='cyl', \
                llcrnrlat=gridInfo['UR_LAT'], urcrnrlat=gridInfo['LL_LAT'], \
                llcrnrlon=gridInfo['LL_LON'], urcrnrlon=gridInfo['UR_LON'])
    QUIVER_THIN = 5

    # add in features to the map
    m.drawcoastlines()
    m.drawcountries()

    # Create matrix of lats and lons for this map
    lons, lats = m.makegrid( gridInfo['NI'], gridInfo['NJ'] )

    # Create the x,y values (in meters) for lons and lats
    x, y = m(lons, lats)

```



```

thin = QUIVER_THIN
xThin, yThin = m(lons[::thin,::thin], lats[::thin,::thin])
uThin = theUGrid[::thin,::thin]
vThin = theVGrid[::thin,::thin]
Q = m.quiver(xThin, yThin, uThin, vThin, color="blue")

'''
# Use the timestamp as a title (for now).
thePlottedTimestamp = timeStamp[0:4] + '-' + timeStamp[4:6]
thePlottedTimestamp += '-' + timeStamp[6:8] + '_'
thePlottedTimestamp += timeStamp[8:10] + 'z'
'''
plt.title(timeStampTitle)

plt.savefig( fullFilename )
plt.close()

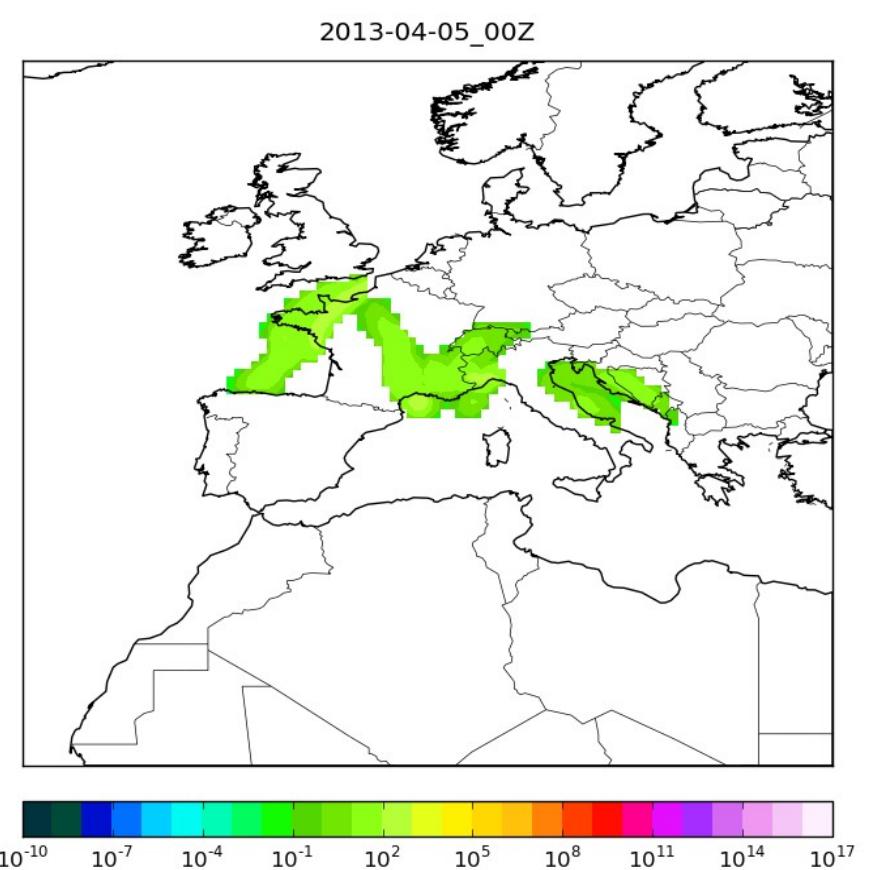
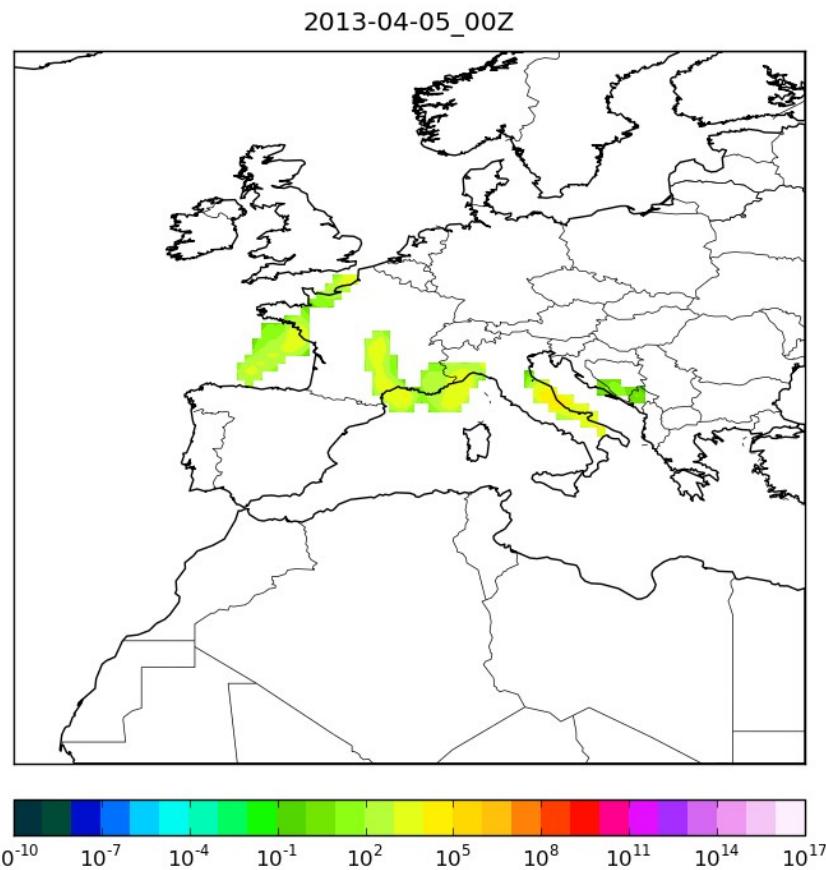
# END createFrame()

#-----
if __name__ == "__main__":
    main()

```



Plotting concentrations from dispersion models (e.g. FLEXPART)



FLEXPART output/ directory

```
dates          grid_conc_20130404140000_001
grid_conc_20130404010000_001  grid_conc_20130404150000_001
grid_conc_20130404020000_001  grid_conc_20130404160000_001
grid_conc_20130404030000_001  grid_conc_20130404170000_001
grid_conc_20130404040000_001  grid_conc_20130404180000_001
grid_conc_20130404050000_001  grid_conc_20130404190000_001
grid_conc_20130404060000_001  grid_conc_20130404200000_001
grid_conc_20130404070000_001  grid_conc_20130404210000_001
grid_conc_20130404080000_001  grid_conc_20130404220000_001
grid_conc_20130404090000_001  grid_conc_20130404230000_001
grid_conc_20130404100000_001  grid_conc_20130405000000_001
grid_conc_20130404110000_001  header
grid_conc_20130404120000_001  receptor_conc
grid_conc_20130404130000_001  trajectories.txt
```

Plotting concentration from a FLEXPART output file

- Assuming a simple, one-species, forward, non-nested run
- Use pflexible (from NILU) Python module to extract necessary data from FLEXPART output header and desired file
 - This is an evolving module that provides access to all of the variables stored in the FLEXPART output
 - Ultimate goal is to extract a 2D horizontal slice, plus projection parameters
- Use matplotlib and basemap to overlay a colour-filled contour on the region

Plotting concentration from a FLEXPART output file

- Most of the work is in the plotting of a single snapshot FLEXPART output
 - Get 2D grid via pflexible and store in 2D numpy array
 - Get grid and projection information via pflexible
 - Use Basemap to define the map region
 - Define a 2D grid over the map region, based on FLEXPART output grid specifications
 - Create a colour-filled contour plot based on FLEXPART output grid values and, mapping it to the region map
 - Add in features like colourbar, title, etc.

quickFLEXVis.py Excerpts

```
# Read the general header information common to all files
H = pf.Header(FLEXPARTOutputDir, nested=False)
.....
# Extract the grid
G = pf.read_grid(H, nspec_ret=0, date=timeStamp)
theTuple = (0, timeStamp)
theGrid = G[theTuple]
gridShape = theGrid.shape      # 4-tuple (x, y, z, p)
.....
theHorizSlice = theGrid.grid[:, :, LEVEL_NUMBER, 0]
theHorizSlice = np.transpose(theHorizSlice)
.....
m = Basemap(projection='cyl', lon_0=lon_0, lat_0=lat_0, \
    lat_ts=lat_0, \
        llcrnrlat=lat_ll, urcrnrlat=lat_ur, \
        llcrnrlon=lon_ll, urcrnrlon=lon_ur, \
        rsphere=6371200., resolution='l', area_thresh=10000)
.....
cs = m.contourf(x, y, theHorizSlice, levels,
norm=mpl.colors.LogNorm(vmin=LO_CONTOUR_VALUE,
vmax=HI_CONTOUR_VALUE), cmap=COLORMAP)
.....
plt.savefig( imageFilename )
plt.close()
```

Timeouts and retries

- Often, the data we want isn't available when we try to retrieve it, or, it may not be available in full
- For flexibility, we try to give the data several chances to come in completely
- Example - fetching of GRIB data from a remote server

Various hardwired criteria

```
##### Various criteria for success, retry, or giving up
# If the file size on the server is not at least this big, then don't retrieve
#MIN_FILESIZE_THRESHOLD_BYTES = 100000000
MIN_FILESIZE_THRESHOLD_BYTES = 50000000

MAX_PROCESS_AGE_MINUTES = 120      # Age at which to terminate process

FILESIZE_WAIT_TIME_SECONDS = 120    # How long to wait before trying filesize again
FILESIZE_WAIT_MAXTRIES = 10         # How many times to try on filesize before
                                    # giving up
FILERETRIEVE_WAIT_TIME_SECONDS = 60 # How long to wait before trying file retrieve
again
FILERETRIEVE_WAIT_MAXTRIES = 5     # How many times to try on file retrieve before
                                    # giving up
SOCKET_TIMEOUT_SECONDS = 60        # How many seconds before a socket times out

.....
# Set the timeout for any sockets created
socket.setdefaulttimeout(SOCKET_TIMEOUT_SECONDS)

.....
# Get process start time (seconds since epoch) for later use
PROCESS_START_TIME = time.time()
currTimeString = time.ctime( PROCESS_START_TIME )
print '[' + currTimeString + '] Start of driving script'
```

Code excerpts

```
# Form timestamp string for current time (will use with ungribbing)
# Format is YYYYMMDDHHmm
currentTimestampStr = startTimeObj.offsetZuluByMinutes( forecastMinute )

# At this point, I want to connect with the FTP server and find out
# how big the existing file is (or if it exists). If all is OK I
# should grab and store. If not, I should try again in a bit for some
# max amount of time. If the file doesn't appear in the expected
# amount of time, then I should either abort this process completely,
# or consider iterating to the next one.
```

Code excerpts

```
# First, just try to get the filesize. Once we have a filesize
# that makes the threshold, we assume file is present.
fileIsPresent = False
numFilesizeTries = 0
serverFileSize = 0
while not fileIsPresent and numFilesizeTries < FILESIZE_WAIT_MAXTRIES:
    theFTPSession = ftplib.FTP(FTP_SERVER_NAME)
    theFTPSession.login(FTP_SERVER_USER, FTP_SERVER_PASSWD)
    try:
        serverFileSize = theFTPSession.size(expectedFullPathname)
        if serverFileSize > MIN_FILESIZE_THRESHOLD_BYTES:
            fileIsPresent = True
        else:
            print 'File found but size does not exceed threshold'
            print '    File: ' + expectedFullPathname
            print '    Size: ' + str(serverFileSize) + ' bytes'
    except ftplib.all_errors:
        numFilesizeTries += 1
        theReason = 'FTP filesize failed: ' + expectedFullPathname
        busyWait(FILESIZE_WAIT_TIME_SECONDS, theReason)
        evaluateProcessTermination(PROCESS_START_TIME, MAX_PROCESS_AGE_MINUTES)
```

Code excerpts

```
# If file passed the previous test (we got filesize and it makes
# file size threshold), try to retrieve it.
if fileIsPresent:

    readyForNext = False
    numRetrieveTries = 0
    while not readyForNext and numRetrieveTries < FILERETRIEVE_WAIT_MAX_TRIES:
        # Retrieve the file
        print 'Retrieving: ' + expectedFullPathname

        fullLocalPathName = LOCAL_STORAGE_DIR + '/' + expectedFilename
        file = open(fullLocalPathName, 'wb')

    try:
        theFTPSession.retrbinary('RETR ' + expectedFullPathname, \
                                open(fullLocalPathName, 'wb').write)

        # Clean up - close the file and connection
        file.close()
        print 'Closing FTP connection...'
        theFTPSession.close()
        readyForNext = True

    except ftplib.all_errors:

        # Wait
        print 'Closing FTP connection...'
        theFTPSession.close()
        print 'Waiting'
        theReason = 'FTP file retrieve failed: ' + expectedFullPathname
        busyWait(FILERETRIEVE_WAIT_TIME_SECONDS, theReason)
        evaluateProcessTermination(PROCESS_START_TIME,
                                    MAX_PROCESS_AGE_MINUTES)
        numRetrieveTries += 1
```

Code Excerpts

```
def busyWait(seconds, reason):
    # Puts process to sleep for "seconds" secs, stating "reason" first

    print 'busyWait() for ' + str(seconds) + ' seconds'
    print '    reason: ' + reason
    print ''
    time.sleep(seconds)

#-----

def evaluateProcessTermination(processStartTime, processKillAge):
    # Compares current process age against designated max age, then
    # kills if we've passed the threshold.  The following code currently
    # assumes units in minutes

    # Get current time in seconds since epoch
    currentTime = time.time()

    # Get current age (seconds), then convert to minutes
    currentAge = (currentTime - processStartTime) / 60.0

    # If we've exceeded max age, exit
    if currentAge > processKillAge:
        print 'evaluateProcessTermination() - process exceeds max age
(minutes)'
        print '    processKillAge: ' + str(processKillAge)
        print '    currentAge     : ' + str(currentAge)
        print '    RIP...'
        sys.exit(1)
```

Logging

<http://code.activestate.com/recipes/577025-loggingwebmonitor-a-central-logging-server-and-mon/>

The screenshot shows the 'Logging Server Status Page' in an Opera browser window. The page displays server statistics and recent log records.

Logging Server Status Page

Logging Server Start Time	2010-02-02 06:26:39
Logging Server Up Time	0:23:11
Log Records Total	1375

Most Recent Log Records

Date	Channel	Level	Location	Message
2010-02-02 06:49:50	myapp.area2	ERROR	<module> (test.py:25)	The five boxing wizards jump quickly.
2010-02-02 06:49:50	myapp.area2	WARNING	<module> (test.py:24)	Jail zesty vixen who grabbed pay from quack.
2010-02-02 06:49:50	myapp.areas1	INFO	<module> (test.py:23)	How quickly daft jumping zebras vex.
2010-02-02 06:49:50	myapp.areas1	DEBUG	<module> (test.py:22)	Quick zephyrs blow, vexing daft Jim.

Logging Web Monitor

AS LoggingWebMonitor - a c x

code.activestate.com/recipes/577025-loggingwebmonitor-a-central-logging-server-and-mon/

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LoggingWebMonitor - a central logging server and monitor. (Python recipe)

1 LoggingWebMonitor listens for log records sent from other processes running in the same box or network. Collects and saves them concurrently in a log file. Shows a summary web page with the latest N records received.

Download Copy to clipboard Python, 409 lines

```
1 """
2 LoggingWebMonitor - a central logging server and monitor.
3
4 Listens for log records sent from other processes running
5 in the same box or network. Collects and saves them
6 concurrently in a log file. Shows a summary web page with
7 the latest N records received.
8
9 Usage:
10 - Add a SocketHandler to your application::
11
12     from logging.handlers import SocketHandler, DEFAULT_TCP_LOGGING_PORT
13     socketh = SocketHandler(servername, DEFAULT_TCP_LOGGING_PORT)
14     logging.getLogger('').addHandler(socketh)
15
16     where servername is the host name of the logging server ('localhost'
17     if run on the same box)
18
19 - Start an instance of this script (the logging server).
```

Created by Gabriel Genellina on Tue, 2 Feb 2010 (MIT)

◀ Python recipes (3997) ▶
◀ Gabriel Genellina's recipes (9) ▶

Tags

client_server debugging distributed logging remote sysadmin web

▶ Show machine tags (19)

Required Modules

os sys cpickle logging

Logging Web Monitor

AS LoggingWebMonitor - a c x

code.activestate.com/recipes/577025-loggingwebmonitor-a-central-logging-server-and-mon/

 **Steve Romanow** 2 years, 11 months ago

@Gabriel: I got it, it was a typo.

 **Don Morton** 2 years, 7 months ago

This server and client are FANTASTIC! I have 20-30 Python scripts that get launched several times a day - all of them need to work for data fetching and processing to culminate in numerical weather forecasts for Alaska. They don't all run on the same machine, and I even have plans to run some of these scripts redundantly on different machines. Each script writes its own log file, but of course if something goes wrong it's an incredibly tedious task to track it back to the particular script that failed.

With only trivial modifications to this code, I've been able to set up a logging server, and then build a Python "LoggerClient" class that each script instantiates. So, now I can refer to the single web page, or the log file, to get a feel for which scripts have run correctly and which haven't.

I had planned on doing this with low-level file locking, but of course that can be very problematic, and requires the use of a filesystem common to all processes (which, in itself, can be problematic). With less than a day of effort, this code has done everything I wanted!

 **rtmie** 2 years, 6 months ago

Great stuff, thanks for this useful script.

You may need to add a `thr_recv.join()` after the `recv.shutdown()` in the main to ensure that no sockets are left in a wait state

Logging

```
import LoggerClient

# Set up the logger client
theLogger = LoggerClient.LoggerClient('FetchAndUngrib RRNatFull')

.....
logMessage = 'Started.  START_TIME: ' + START_TIME + ', '
logMessage += 'NUM_FORECAST_HOURS: ' + str(NUM_FORECAST_HOURS)
theLogger.logInfo(logMessage)
```

Ambitions

- A multi-machine process organiser and timer
 - Trying to get out of "crontab hell"
- System for rapidly running and combining multiple model runs and configurations into a single output mechanism – e.g. Emergency response system for volcano eruptions, nuclear power plant disasters, etc.
- Seamless utilisation of cloud computing environments