

Meteorology and Python

*desperately trying to forget
technical details*

*Claude Gibert,
Europython 2011*

Background

- Meteorology - NWP – Numerical Weather Prediction
- ECMWF – European Centre for Medium-Range Weather Forecasts.
- GMAO – Global Modelling and Assimilation Office at NASA. Run a numerical forecast to calibrate satellite sensors.
- Atmospheric models running from initial conditions: the analysis – optimisation function – observations
- Forecast verification – statistics computed against the verifying analysis and observations
- Observation statistics – monitoring of observing systems

History

- It all started in 2004.
- Design a forecast verification package:
 - 1) offer an interface which doesn't require programming, but can be used in batch processing
 - 2) retrieve raw data from the archive
 - 3) compute statistics
 - 4) store statistics
 - 5) extract and display statistics (plots)

Thoughts (1)

- offer an interface which doesn't require programming, but can be used in batch processing:
 - this called for a “descriptive” language which could have been anything from XML to a “home made” syntax
 - I have always been against proprietary languages which proliferated. Good examples of scientific software were grads, matlab, IDL, pwwave. Not a good idea.
 - There was an interface for accessing the archive which was request based:

```
retrieve,  
  param = T,  
  level = 500/850,  
  type = forecast,  
  expver = 0001
```


Thoughts (2)

- retrieving raw data is difficult:
 - interface the multiple PB archive and handle its complexity
 - interface data decoding libraries (Fortran and C) to decode the data from the archive, two main formats
 - have sufficient knowledge of meteorological parameters to derive some of them from others which are in the archive.
- compute statistics: this is relatively trivial, but it is the stage preceding the computing which requires the most attention:
 - to pair and possibly aggregate data correctly
 - this means associating the right forecast with the right reference or the observations from a same observing system together.

Thoughts (3)

- store statistics. The example of the “retrieve” request shows few parameters, in fact there are probably 15 different ones for each statistic. This is a typical meteorological archive, the size of the metadata is about 20 times the size of the data.
- A SQL database was used although I have now been looking at Non-SQL databases. This is out of the scope of this presentation.

```
retrieve,  
  param = T,  
  level = 500/850,  
  type  = forecast,  
  expver = 0001
```

Thoughts (4)

- extract and display (plot) statistics. A flexible system should offer a decent choice of plots, but the user should have very good control over them:
 - choice of bars, curves etc...
 - way of grouping curves based on metadata
 - ways of organising plots
 - control over titles and legends

Decisions (1)

- this package was going to do what most people in the organisation already did in separate programs and languages (Fortran and C) for many different purposes:
 - access the archive
 - decode data
 - identify data
 - post-process data
 - pair data if applicable
 - compute
 - store / plot
- at best the code reuse was “copy-and-paste”

Choices (1)

- it made sense to develop a framework first and then build the verification package with it:
 - hopefully if users invested in the framework, a common tool would simplify maintenance and development efforts
 - the ultimate goal of developing an application would be met
 - offers a platform for new developments
- the question was: in which language should this be done?

Choices (2)

- I needed a language which would be easy to interface with C, C++ and possibly Fortran. I also needed to glue different libraries together.
- Then I learnt Python in 1 day using “Dive into Python”, and actually tried it.

Purpose of the talk

- I am going to describe how the **language** for the user was developed to be both:
 - an efficient way of conveying information to the application, let's be modest, a “fancy” argument system
 - a way of supporting the developer, by guaranteeing the validity of the input and configuration defaults.
- I would like to show how the definition of the concept of a **directive** can contribute to the creation of complex requests for a plotting system.

The language

- Using the Python interpreter for the interface was the best option to insert custom code. I went back to:

```
retrieve,  
    param = T,  
    level = 500/850,  
    type = forecast,  
    expver = 0001
```

this could easily be mapped to:

```
retrieve(  
    param = 'T',  
    level = [500,850],  
    type = 'forecast',  
    expver = '0001'  
)
```

- this is called a **Directive**

The directive

- Directives are basically Python dictionaries to which semantics are added, to help both the programmer and the user:
 - list of valid keywords (dictionary keys)

```
{
  "directive": "store",
  "keywords": {
    "name": {
    },
    "age": {
    },
    "nationality": {
    }
  }
}
```

Why dictionaries? Because they are part of the core of Python and they make it awesome.


```
observation = {
  'directive': 'obsidentifier',
  'keywords' : {
    'date': { 'type': int },
    'domain_name': {
      'alias': ['domain'],
      'default_value': ['global']
    },
    'variable': {
    },
    'level': {
      'alias': 'channel',
      'type': float,
      'optional': True
    },
    'type': {
      'validate': ['ValidateChoice', 'ob', 'im'],
      'default_value': 'ob',
      'unique': True
    },
    'kt': { 'type': int, },
    'kx': { 'type': int, }
  }
}
```

```

observation = {
'directive': 'obsidentifier',
'keywords' : {
    'date': { 'type': int },
    'domain_name': {
        'alias': ['domain'],
        'default_value': ['global']
    },
    'variable': {
    },
    'level': {
        'alias': 'channel',
        'type': float,
        'optional': True
    },
    'type': {
        'validate': ['ValidateChoice', 'ob', 'im'],
        'default_value': 'ob',
        'unique': True
    },
    'kt': { 'type': int, },
    'kx': { 'type': int, }
}
}

```

```

class Observation(Directive):

    def __init__(self,*args,**kwargs):
        super(Observation,self).__init__(*args,**kwargs)
        self.checkLanguage()

    def languageReader(self):
        return DirectiveReader()

```

```

print Observation()
...
In directive observation:
Keyword variable is missing and is required
Keyword kt is missing and is required
Keyword date is missing and is required
Keyword kx is missing and is required

```

```

print Observation(
    variable = 'omf',
    kt = [4,5],
    kx = 220,
    type = ['ob'],
    date = 2011020312,
    domain = ['europe']
)
...
observation:
    domain_name = ['europe']
    variable = ['omf']
    kt = [4, 5]
    date = [2011020312]
    kx = [220]
    type = ob
...
type = 'wrong',
...
In directive observation:
Validation error for keyword type. The value: 'wrong'
is not in the set: ob, im

```

Argument checking

- As the directive system is defined here, it can be useful to format arguments to methods, either for the lifetime of the application or only at debug stage:

```
from checkargs import checkArgs

checkArgs.register('__call__',dict(
    directive = '__call__',
    keywords = dict(
        a = dict(unique = True),
        b = dict(alias = 'd', optional = True),
        c = dict(default_value = '12', type = int))
    )
)

class MyClass(object):

    @checkArgs
    def __call__(self,*args,**kwargs):
        return kwargs

print MyClass()(a = [12],d = 10)
```

```
{'a': 12, 'c': [12], 'b': [10]}
```

Directive

- Defining semantics for a directive is like working at class level when writing code. Just the same way, inheritance and specialisation are available for directive definition:

```
surface_observation = {  
  'directive': 'surface_observation',  
  'inherit_from': 'observation',  
  'keywords' : {  
    'level': { 'default_value': 0 },  
    'station_height': { 'unique' : True }  
  }  
}
```

- Class inheritance is not required to mimic inheritance in directive definition.
- Multiple inheritance is supported.

Directive: specialisation

```
observation = {
  'directive': 'obsidentifier',
  'specialise_from': {
    'type == "im"': 'impact',
    'type == "ob"': 'rawobs',
  }
  'keywords' : {
    'type': {
      'validate': ['ValidateChoice', 'ob', 'im'],
      'default_value': 'ob',
      'unique': True
    },
    etc...
  }
}
{
  "directive": "impact",
  "keywords" : {
    "variable": {
      "default_value": "xvec"
    }
  }
}
```


Directive behaviour

- An object instance from a class inheriting from **Directive** can also inherit from other instances. This is some way of merging dictionaries with different flavours.
- `child.inherit_from(parent)`: assign to **child** all keys from **parent** which:
 - are defined in its language **and**
 - are not defined in child **or**
 - are a default value in child and not in parent
- `child.overwrite_from(parent)`: assign to **child** all keys from **parent** which:
 - are defined in its language **and**
 - are not defined in child **or**
 - are a default value in child
- recursively
- `__setitem__` is overloaded to keep track of default values and language.

Plotting

- Plotting is not simple and it seems that it has always more or less been a “semi-manual” process. Even when the graphics software provides good support (e.g. matlab, matplotlib) the user normally specifies manually:
 - the data for each curve,
 - the legends,
 - the title
- Most graphical packages do not provide sufficient support for automatic plotting, trying to figure out what to use, and are probably not sufficiently object orientated, for example:
 - xaxis, yaxis methods, plot_date etc...
 - there is always a way around but the good stuff is hardcoded

Plotting

- However, knowing the data to be plotted is normally the biggest problem. Observations have the following attributes:
 - kt
 - kx
 - level
 - domain_name
 - date
 - statistic
- How can we specify that we want n curves, each of them is one combination of kt, kx, level and we want a different domain for each plot?
- How can we specify that we want n curves, each of them is one combination of kt, kx, domain and we want a different level for each plot?

Document – Plot – Curve Model

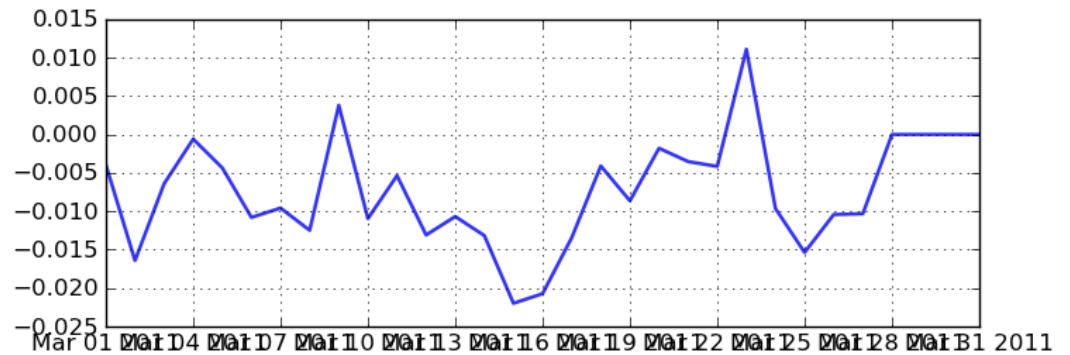
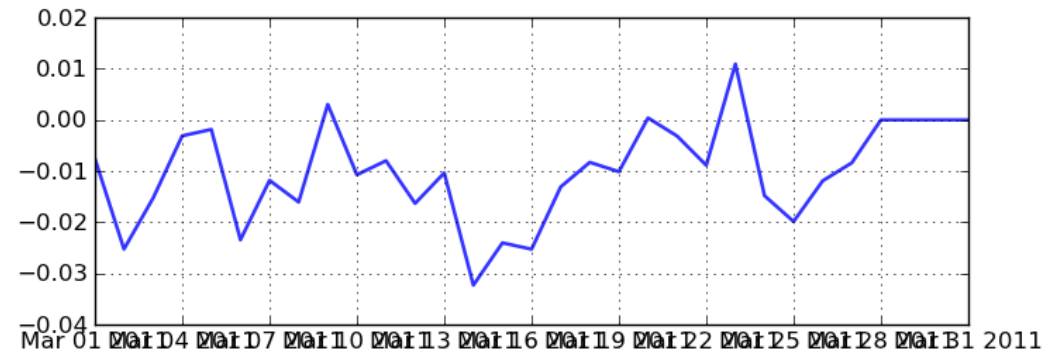
- Each directive inherits from **observation**:
 - a **document** object contains:
 - attributes related to graphics
 - attributes related to layout and output
 - attributes related to the data being plotted
 - possibly title information
 - a list of **plot** objects
 - a **plot** object contains
 - attributes related to graphics
 - attributes related to the data being plotted
 - title information
 - a list of **curve** objects
 - a **curve** object contains:
 - attributes related to graphics
 - attributes related to the data being plotted
 - legend information

```

d = obsdocument(
  plot = [
    timeseriesplot(
      date = Dates(2011030100,2011033100,24),

      curve = line(
        kx = [120,220,221,132,229,232],
        kt = [4,5,11,44]
      )
    ),
  ],
  type = 'im',
  level = 1000,
  statistic = 'rate',
  domain = ['global','n.hem'],
  layout = [1,2]
}

```

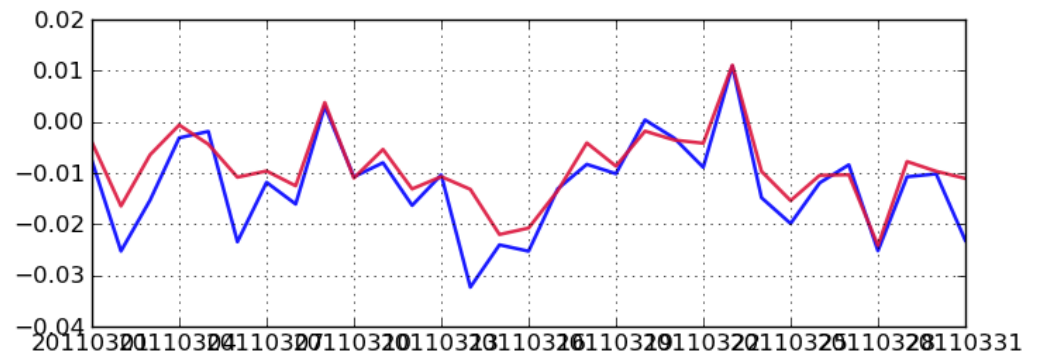



```

d = obsdocument(
  plot = [
    timeseriesplot(
      date = Dates(2011030100,2011033100,24),
      domain = ['global','n.hem'],
      curve = line(
        kx = [120,220,221,132,229,232],
        kt = [4,5,11,44]
      )
    ),
  ],
  type = 'im',
  level = 1000,
  statistic = 'rate',

  layout = [1,2]
}

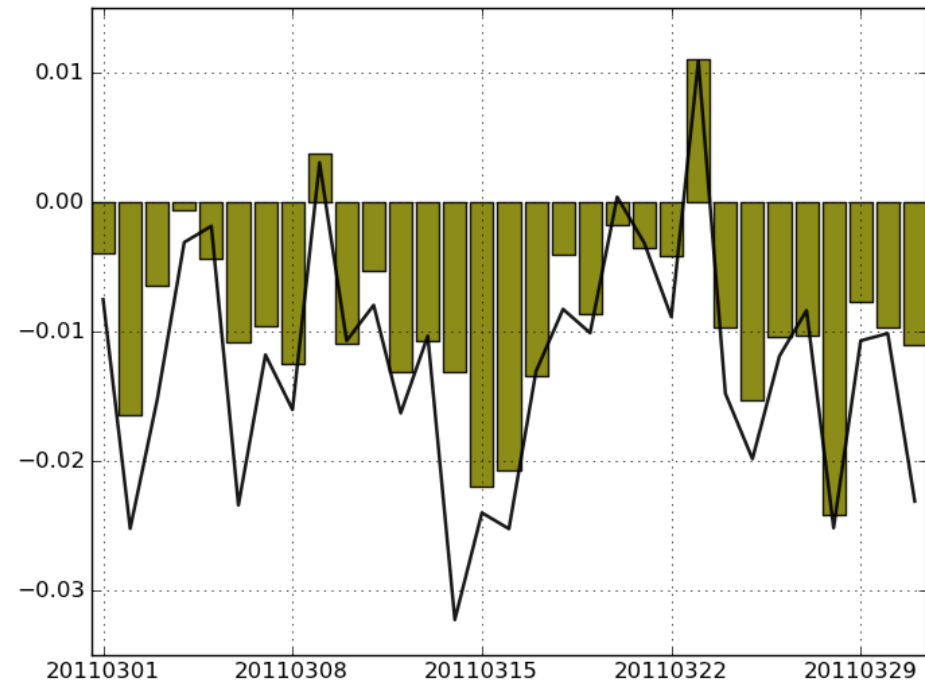
```



```

d = obsdocument(
  plot = [
    timeseriesplot(
      date = Dates(2011030100,2011033100,24),
      domain = ['s.hem','tropics'],
      kx = [120,220,221,132,229,232],
      kt = [4,5,11,44],
      curve = [
        line(
          domain = 'global',
          graphics = graphics(color = 'black')
        ),
        bar( ← became bar
          domain = 'n.hem',
          graphics = graphics(c
        ),
      ],
      yaxis = axis(
        min = -0.035,
        max = 0.015
      ),
    ),
  ],
  type = 'im',
  level = 1000,
  statistic = 'rate',
)

```

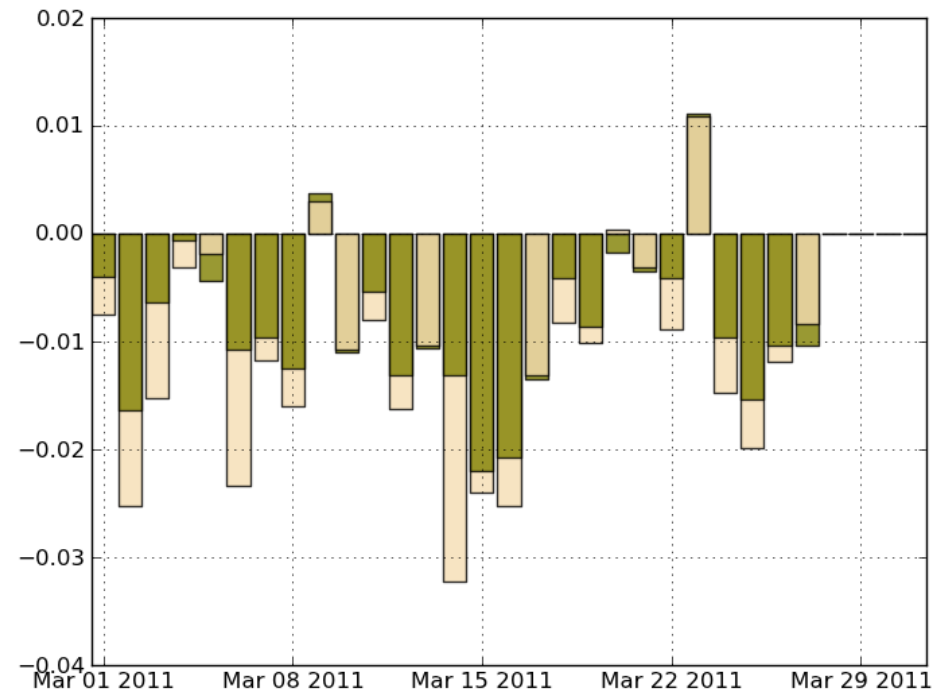


```

d = obsdocument(
  plot = [
    timeseriesplot(
      date = Dates(2011030100,2011033100,24),
      domain = ['global','n.hem'],
      curve = bar(          # <-- was line became bar
        kx = [120,220,221,132,229,232],
        kt = [4,5,11,44]
      )
    ),
  ],
  type = 'im',
  level = 1000,
  statistic = 'rate',
)

"_combinable": {
  "default_value": [
    "domain_name",
    "statistic",
    "level"
  ]
},
"_index": {
  "default_value": "date"
}

```

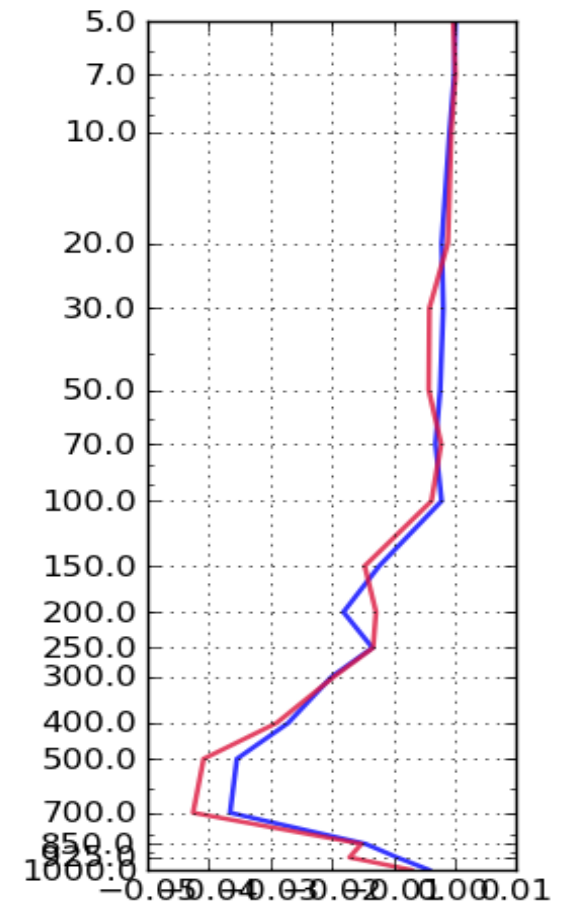


```

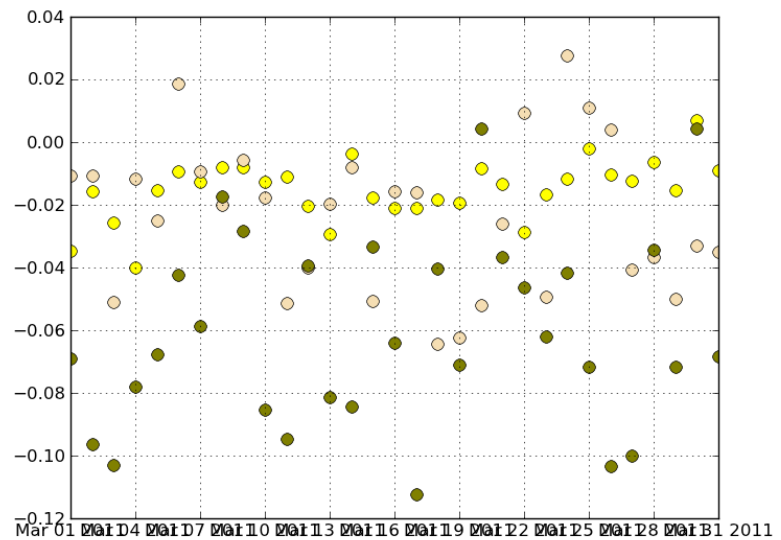
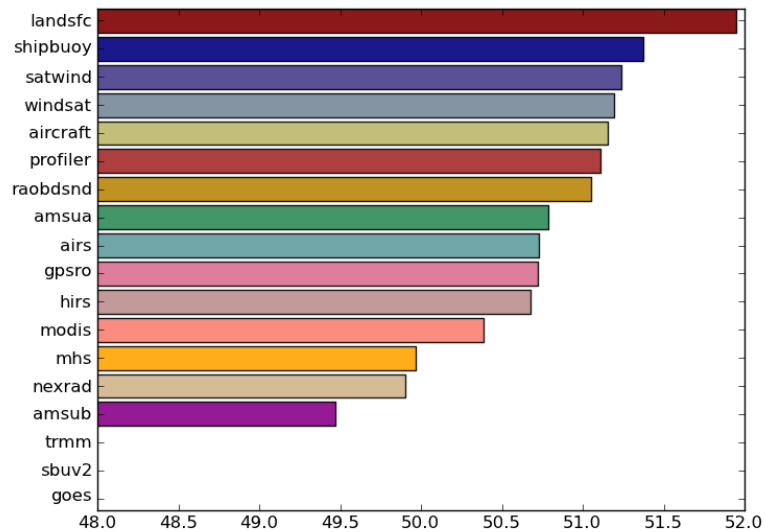
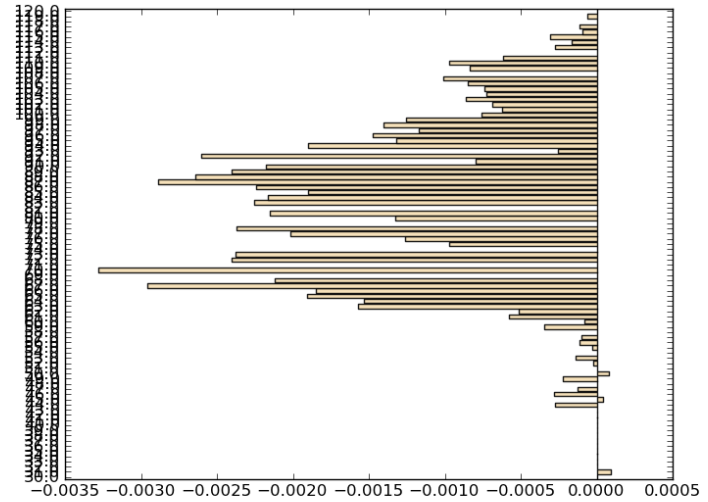
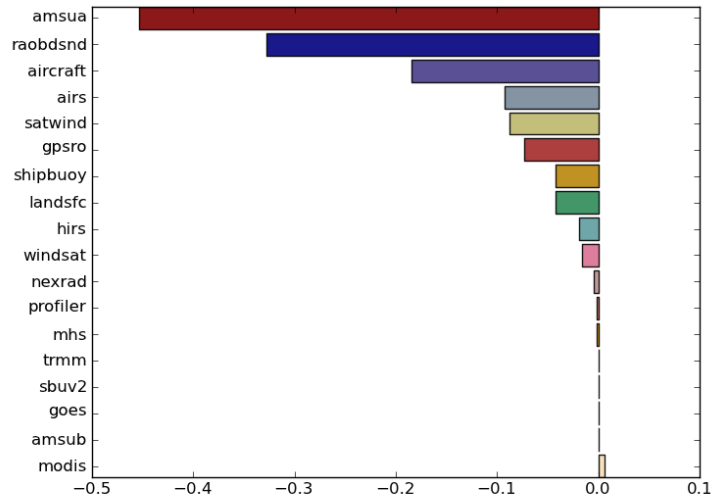
d = obsdocument(
  plot = [
    verticalxsection(
      statistic = 'rate',
      level = levels,
      domain = ['n.hem', 's.hem']
      curve = [
        line(
          kx = 220,
          kt = [4,5],
          date = Dates(2010090100,2010103100,24),
        ),
      ],
    ),
  ],
  type = 'im',
  domain = 'global',
)

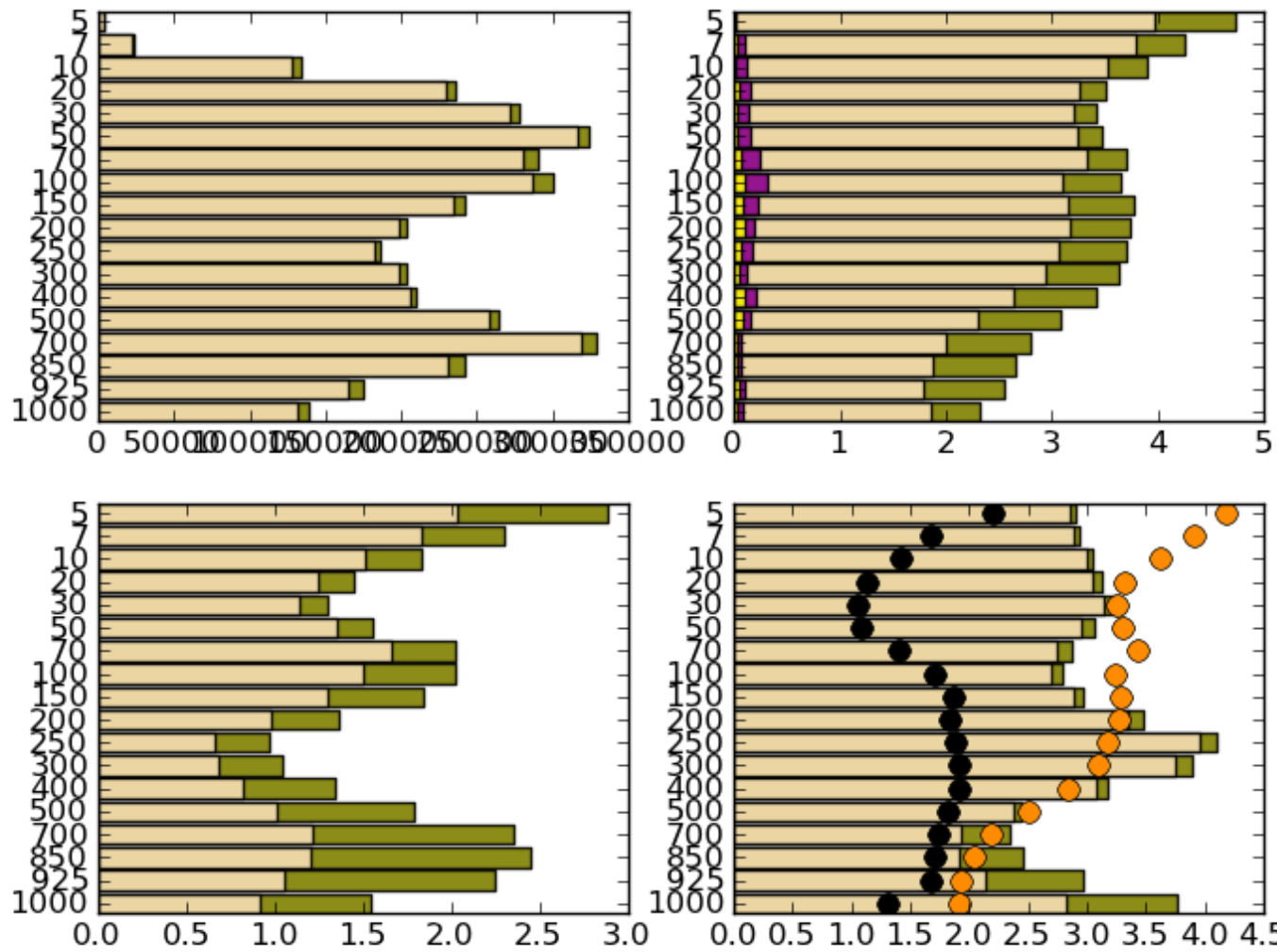
"_combinable": {
  "default_value" [
    "domain_name",
    "statistic"
  ]
},
"_index": {
  "default_value": "level"
}

```



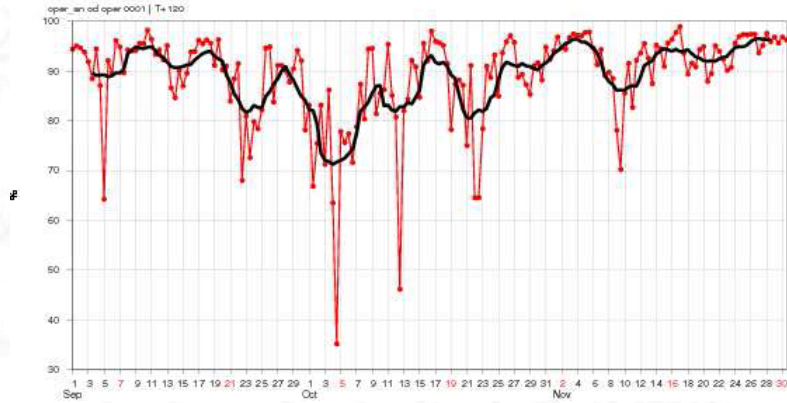
Gallery



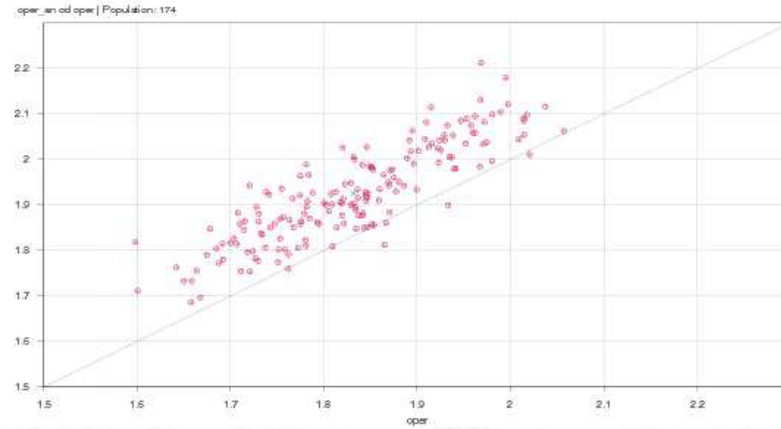


Gallery

500hPa geopotential
Correlation coefficient of forecast anomaly
Europe (lat: 35.0 to 75.0, lon: -12.5 to 42.5)

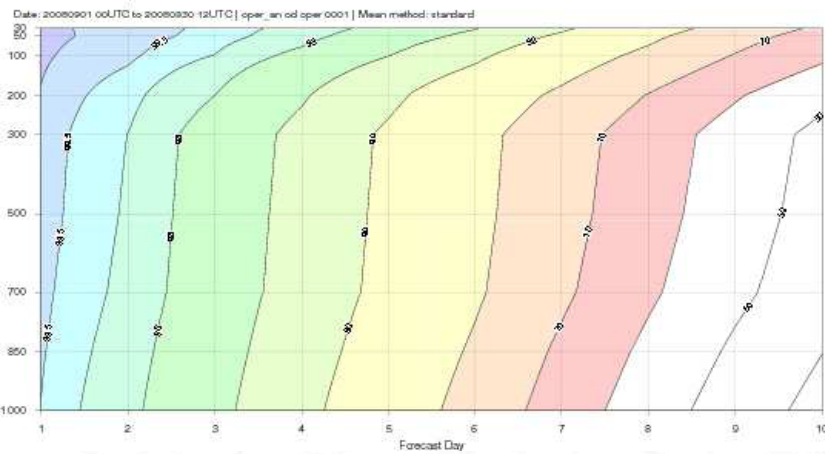


850hPa wind speed
Root mean square error of forecast
Tropics (lat: -20.0 to 20.0, lon: -180.0 to 180.0)
Date: 20100802 00UTC to 20101031 12UTC

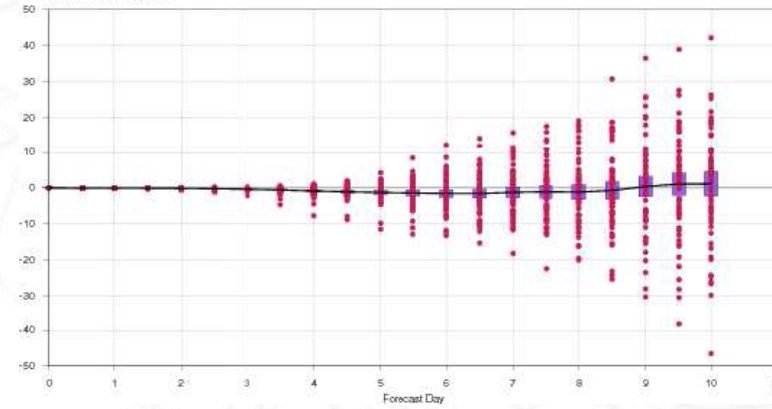


"oper" is WORSE than "results 36r4" at the 0.1 level (sign_test)
"oper" is BETTER than "results 36r4" at the 0.1 level (paired_t_test)
"oper" is BETTER than "results 36r4" at the 0.1 level (weak_t_test)

geopotential
Correlation coefficient of forecast anomaly
S Hem Extratrop (lat: -90.0 to -20.0, lon: -180.0 to 180.0)



enfo minus oper
500 hPa geopotential
Correlation coefficient of forecast anomaly
N Hem Extratrop (lat: 20.0 to 90.0, lon: -180.0 to 180.0)
Date: 20090901 00UTC to 20090930 12UTC
Confidence: 92.0 | Population: 60



Conclusion

- The directive helps the user in specifying values, a 'help' key can be added in the definition file. The syntax is quite simple and no programming is required. However, for some keywords, callables can be specified.
- The directive helps the programmer in the sense that it guarantees that the values of a dictionary are formatted according to specifications (lower case, lists etc...)

Score computation

