



EXAMPLES USING THE IMOS USER CODE LIBRARY (Python VERSION)

Version 1.0

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IMOS - eMII

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**Integrated Marine
Observing System**

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Revision History

Name	Date	Reason For Changes	Version

1. Introduction

This document intends to present how to load IMOS NetCDF data into a Python environment, and offers some suggestions about how to use the data once loaded. All the examples below will use the NETCDF4 python module (<http://code.google.com/p/netcdf4-python/>)

The examples provided in this document only represent a tiny bit of the content of most of the NetCDF files. There are usually many more variables available in a NetCDF file, and therefore many other ways to display data.

1.1 Installation of the IMOS User Code Library (Python)

The IMOS User Code Library for Python can be downloaded from:

https://github.com/aodn/imos_user_code_library/tree/master/Python/

This contains few Python functions to handle more easily IMOS NetCDF file with the NETCDF4 module.

It can be checked out using a Git client, or be downloaded as a zip file :

https://github.com/aodn/imos_user_code_library/archive/master.zip

The NETCDF4 python module needs to be installed. Please follow the description available at <http://code.google.com/p/netcdf4-python/>.

1.2 Finding an IMOS NetCDF File

In order to find a dataset you are interested in, please refer to the portal help:

http://portalhelp.aodn.org.au/Portal2_help/

A HOWTO has been written to help any user in his way to find an IMOS NetCDF file.

For users who are already familiar with IMOS facilities and datasets, IMOS NetCDF files are also directly accessible via an OPeNDAP catalog at :

<http://thredds.aodn.org.au/thredds/catalog/IMOS/catalog.html>

Once a NetCDF file has been chosen to work with (See http://portalhelp.aodn.org.au/Portal2_help/?q=node/112), the user needs to go to the '*OPeNDAP Dataset Access Form*' page. The '**Data URL**' located just above the '**Global Attributes**' field is the URL which needs to be replaced in the examples which we'll present in the following sections.

The NetCDF file can also be downloaded to the user's local machine using the HTTP Server option on the THREDDS options page. In this case, the user has to replace the 'Data URL' variable with the local address of the NetCDF file when using the function '**Dataset**' in the following examples.

2. General Features of the IMOS user code library

The first step consists of parsing a NetCDF file, whether this file is available locally or remotely on an OPeNDAP server.

Type in your Python command window:

```
from imosNetCDF import *
from netCDF4 import Dataset

aatams_URL = 'http://thredds.aodn.org.au/thredds/dodsC/IMOS/AATAMS/marine_mammal_ctd-
tag/2009_2011_ct64_Casey_Macquarie/ct64-M746-09/IMOS_AATAMS-SATTAG_TSP_20100205T043000Z_ct64-M746-
09_END-20101029T071000Z_FV00.nc'
aatams_DATA = Dataset(aatams_URL)
```

2.1 Discover Metadata

In order to see all the metadata available, type in your command window:

```
metadata = getAttNC(aatams_DATA)
```

```
u'abstract': u'CTD Satellite Relay Data Loggers are used to explore how marine mammal
behaviour relates to their oceanic environment. Loggers developed at the University of St
Andrews Sea Mammal Research Unit transmit data in near real time via the Argo satellite system',
u'acknowledgment': u'Any users of IMOS data are required to clearly acknowledge the source of
the material in the format: "Data was sourced from the Integrated Marine Observing System
(IMOS) - IMOS is supported by the Australian Government through the National Collaborative
Research Infrastructure Strategy (NCRIS) and the Super Science Initiative (SSI)",
u'author': u'Besnard, Laurent',
u'author_email': u'laurent.besnard@utas.edu.au',
u'body_code': u'11449',
u'cdm_data_type': u'Trajectory',
...
```

You can attach to the python variable 'title_str' the corresponding NetCDF metadata attribute:

```
title_str = metadata['title']
```

```
u' Temperature, Salinity and Depth profiles in near real time'
```

2.2 Discover Variables

In order to list all the variables available in each NetCDF file, type:

```
aatams_DATA.variables
```

```
OrderedDict([(u'TIME', <netCDF4.Variable object at 0x4561bd0>), (u'LATITUDE',
<netCDF4.Variable object at 0x4561cd0>), (u'LONGITUDE', <netCDF4.Variable object at
```

```
0x4561d50>), (u'TEMP', <netCDF4.Variable object at 0x4561dd0>), (u'PRES',  
<netCDF4.Variable object at 0x4561e50>), (u'PSAL', <netCDF4.Variable object at 0x4561ed0>),  
(u'parentIndex', <netCDF4.Variable object at 0x4561f50>), (u'TIME_quality_control',  
<netCDF4.Variable object at 0x549f050>), (u'LATITUDE_quality_control', <netCDF4.Variable  
object at 0x549f0d0>), (u'LONGITUDE_quality_control', <netCDF4.Variable object at  
0x549f150>), (u'TEMP_quality_control', <netCDF4.Variable object at 0x549f1d0>),  
(u'PRES_quality_control', <netCDF4.Variable object at 0x549f250>), (u'PSAL_quality_control',  
<netCDF4.Variable object at 0x549f2d0>))
```

To access the Temperature variable :

- data values

```
aatams_DATA.variables['TEMP']
```

- names of corresponding dimension variables

```
aatams_DATA.variables['TEMP'].dimensions
```

```
(u'obs',)
```

3. Dataset examples – Using the NetCDF Parser for Plotting

3.1 AATAMS – Animal Tagging and Monitoring

The Australian Animal Tagging And Monitoring System (AATAMS) is a coordinated marine animal tagging project. CTD Satellite Relay Data Loggers are used to explore how marine mammal behaviour relates to their oceanic environment.

NetCDF files can be found at :

http://thredds.aodn.org.au/thredds/catalog/IMOS/AATAMS/marine_mammal_ctd-tag/catalog.html

In the example below, we demonstrate how to use the netCDF4 module to plot all the animal's dives as a single profile time-series of temperature, measured by CTD tag.

```
from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

## AATAMS - Animal Tagging and Monitoring
aatams_URL = 'http://thredds.aodn.org.au/thredds/dodsC/IMOS/AATAMS/marine_mammal_ctd-
tag/2009_2011_ct64_Casey_Macquarie/ct64-M746-09/IMOS_AATAMS-SATTAG_TSP_20100205T043000Z_ct64-M746-
09_END-20101029T071000Z_FV00.nc';
aatams_DATA = Dataset(aatams_URL)
metadata = getAttNC(aatams_DATA)

nProfiles = len(aatams_DATA.dimensions['profiles']) # the number of profiles undertaken by the seal
parentIndex = aatams_DATA.variables['parentIndex'][:,] #for each obs which profile it is linked to

# loading of the variable objects
TEMP = aatams_DATA.variables['TEMP']
PRES = aatams_DATA.variables['PRES']
PSAL = aatams_DATA.variables['PSAL']
TIME = aatams_DATA.variables['TIME']

# creation of a 2 dimension array for temperature, pressure and salinity
psalData = aatams_DATA.variables['PSAL'][:,]
tempData = aatams_DATA.variables['TEMP'][:,]
presData = aatams_DATA.variables['PRES'][:,]

# we want to know the maximum number of observations (or depth level) per profile
# for all the profile. This number 'maxObsProfile' will be used to create a 2d
# array for Temperature salinity and pressure.
maxObsProfile = 0.
for profileNumber in range(1,nProfiles):
    indexVar = where(parentIndex == profileNumber)
```

```

    if size(indexVar) > maxObsProfile:
        maxObsProfile = size(indexVar)

# we recreate those variables to have a 2d array
TEMP_DATA_reshaped = numpy.empty((nProfiles,maxObsProfile,))
PSAL_DATA_reshaped = numpy.empty((nProfiles,maxObsProfile,))
PRES_DATA_reshaped = numpy.empty((nProfiles,maxObsProfile,))

for profileNumber in range(nProfiles):
    indexVar = where(parentIndex == profileNumber)
    TEMP_DATA_reshaped[profileNumber,0:size(indexVar)] = tempData[indexVar]
    PSAL_DATA_reshaped[profileNumber][range(0,size(indexVar))] = psalData[indexVar]
    PRES_DATA_reshaped[profileNumber][range(0,size(indexVar))] = presData[indexVar]

# we load the latitude and longitude values for all the profiles
latProfile = numpy.array(aatams_DATA.variables['LATITUDE'][:])
lonProfile = numpy.array(aatams_DATA.variables['LONGITUDE'][:])

#longitude in the original dataset goes from -180 to +180
#For a nicer plot, we change the values to the [0 360] range
lonProfile[lonProfile < 0 ] = lonProfile[lonProfile < 0 ] +360

# we convert the time values into a python time object
timeData = convertTime(TIME) # one value per profile

# creation of a profile variable array
size = ones((1,maxObsProfile),'float')
#observation = range(nProfiles)
profIndex = array(range(nProfiles))
profIndex = profIndex.reshape(nProfiles,1)
prof_2D = profIndex * size

## PLOT
#plot all the profiles as a timeseries
figure1 = figure(num=None, figsize=(15, 10), dpi=80, facecolor='w', edgecolor='k')
subplot(311)
pcolor(prof_2D, -PRES_DATA_reshaped, TEMP_DATA_reshaped)
cbar = colorbar()
cbar.ax.set_ylabel(TEMP.long_name + ' in ' + TEMP.units)
title(metadata['species_name'] + ' - released in ' + metadata['release_site'] + ' \n animal
reference number : ' + metadata['unique_reference_code'])
xlabel('Profile Index')
ylabel(PRES.long_name + ' in negative ' + PRES.units)

from matplotlib.dates import MONTHLY, DateFormatter, rrulewrapper, RRuleLocator
rule = rrulewrapper(MONTHLY, bymonthday=1, interval=1)
formatter = DateFormatter('%d/%m/%y')
loc = RRuleLocator(rule)

#plot the LON timeseries
ax3 = subplot(234)
plot(timeData,lonProfile)
ax3.xaxis.set_major_locator(loc)

```

```

ax3.xaxis.set_major_formatter(formatter)
labels = ax3.get_xticklabels()
setp(labels, rotation=30, fontsize=10)
xlabel(TIME.long_name + ' in ' + 'dd/mm/yy' )
ylabel(aatams_DATA.variables['LONGITUDE'].long_name + ' in ' +
aatams_DATA.variables['LONGITUDE'].units)

#plot the LAT timeseries
ax4 = subplot(235)
plot(timeData,latProfile)
ax4.xaxis.set_major_locator(loc)
ax4.xaxis.set_major_formatter(formatter)
labels = ax4.get_xticklabels()
setp(labels, rotation=30, fontsize=10)
xlabel(TIME.long_name + ' in ' + 'dd/mm/yy' )
ylabel(aatams_DATA.variables['LATITUDE'].long_name + ' in ' +
aatams_DATA.variables['LATITUDE'].units)

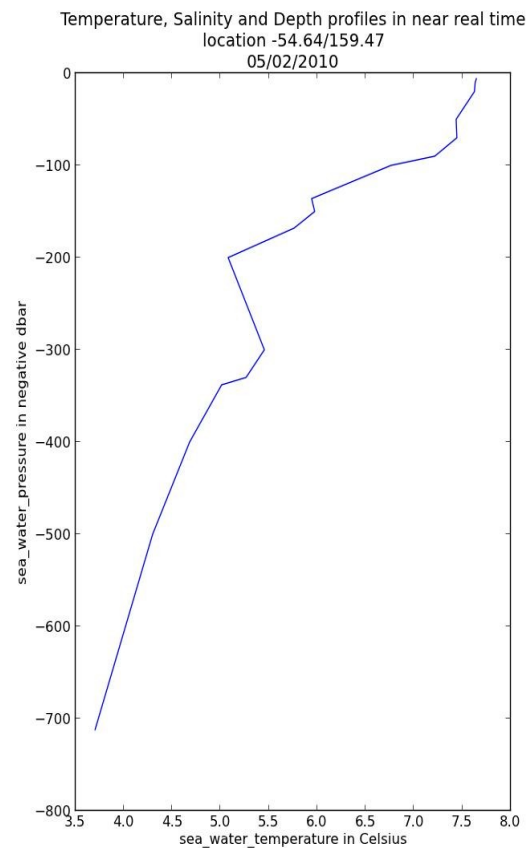
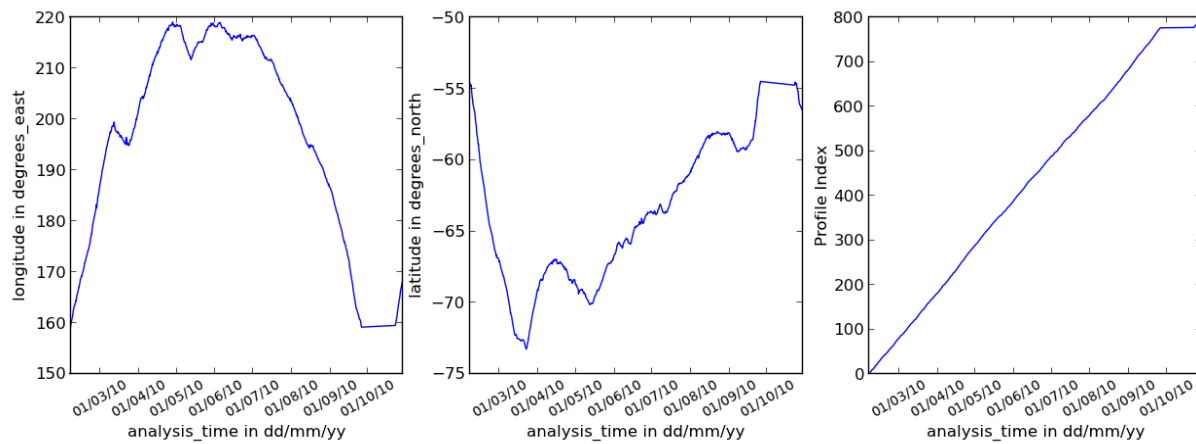
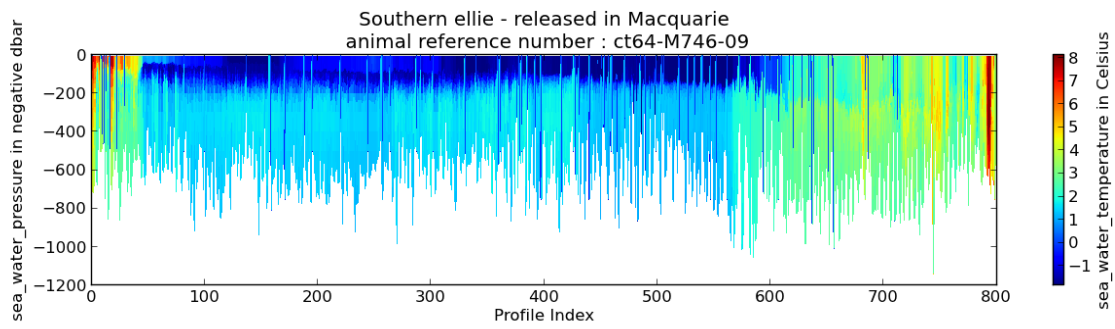
#plot the profile index with time values
ax5 = subplot(236)
plot(timeData,profIndex)
ax5.xaxis.set_major_locator(loc)
ax5.xaxis.set_major_formatter(formatter)
labels = ax5.get_xticklabels()
setp(labels, rotation=30, fontsize=10)
xlabel(TIME.long_name + ' in ' + 'dd/mm/yy' )
ylabel('Profile Index')

#plot of a single profile
profileToPlot = 1# this is arbitrary. We can plot all profiles from 1 to nProfiles, modify
profileToPlot if desired
figure2 = figure(num=None, figsize=(7, 10), dpi=80, facecolor='w', edgecolor='k')
plot (TEMP_DATA_reshaped[profileToPlot,:],-PRES_DATA_reshaped[profileToPlot,:])
title(metadata['title'] + '\nlocation ' + "%0.2f" % latProfile[profileToPlot] + '/' + "%0.2f" %
lonProfile[profileToPlot] + '\n' + timeData[profileToPlot].strftime('%d/%m/%Y'))
xlabel(TEMP.long_name + ' in ' + TEMP.units)
ylabel(PRES.long_name + ' in negative ' + PRES.units)
plt.show()

```

Variables to modify :

- aatams_URL : the opendap url of the chosen file
- ProfileToPlot : the profile number to plot.



3.2 ABOS – Deep Water Mooring

3.2.1 Southern Ocean Time-series

The Southern Ocean Time Series (SOTS) sub-facility provides high temporal resolution observations in sub-Antarctic waters. Observations are broad and include measurements of physical, chemical and biogeochemical parameters from multiple deep-water moorings.

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/catalog/IMOS/ABOS/SOTS/catalog.html>

In the example below, the netCDF4 module is used to extract temperature data from a Pulse mooring instrument and then produce a temperature time series plot .

```

from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### ABOS
abos_URL = 'http://thredds.aodn.org.au/thredds/dodsC/IMOS/ABOS/SOTS/Pulse/IMOS_ABOS-
SOTS_20110803T000000Z_PULSE_FV01_PULSE-8-2011_END-20120719T000000Z_C-20121009T214808Z.nc'
abos_DATA = Dataset(abos_URL)

tempDataStructure = abos_DATA.variables['TEMP_85_1']
TIME = abos_DATA.variables['TIME']

tempData = tempDataStructure[:]
timeData = convertTime(TIME) # one value per profile

metadata = getAttNC(abos_DATA)
abstract = metadata['abstract']

figure1 = figure( figsize=(10, 10), dpi=80, facecolor='w', edgecolor='k')
ax = subplot(111)
# we have to find the index for no nan values
indexNoNan = where( ~numpy.isnan(tempData))
plot(timeData[indexNoNan],tempData[indexNoNan])

xlabel(TIME.long_name + ' in ' + 'dd/mm/yy' )
ylabel(tempDataStructure.standard_name + ' in ' + tempDataStructure.units)
title(metadata['title'] + '\nat ' + "%0.2f" %tempDataStructure.sensor_depth + ' m depth' )

# time ticks
from matplotlib.dates import MONTHLY, DateFormatter, rrulewrapper, RRuleLocator
rule = rrulewrapper(MONTHLY, bymonthday=1, interval=1)
formatter = DateFormatter('%d/%m/%y')

```

```
loc = RRuleLocator(rule)
ax.xaxis.set_major_locator(loc)
ax.xaxis.set_major_formatter(formatter)
labels = ax.get_xticklabels()
setp(labels, rotation=30, fontsize=10)
show()
```

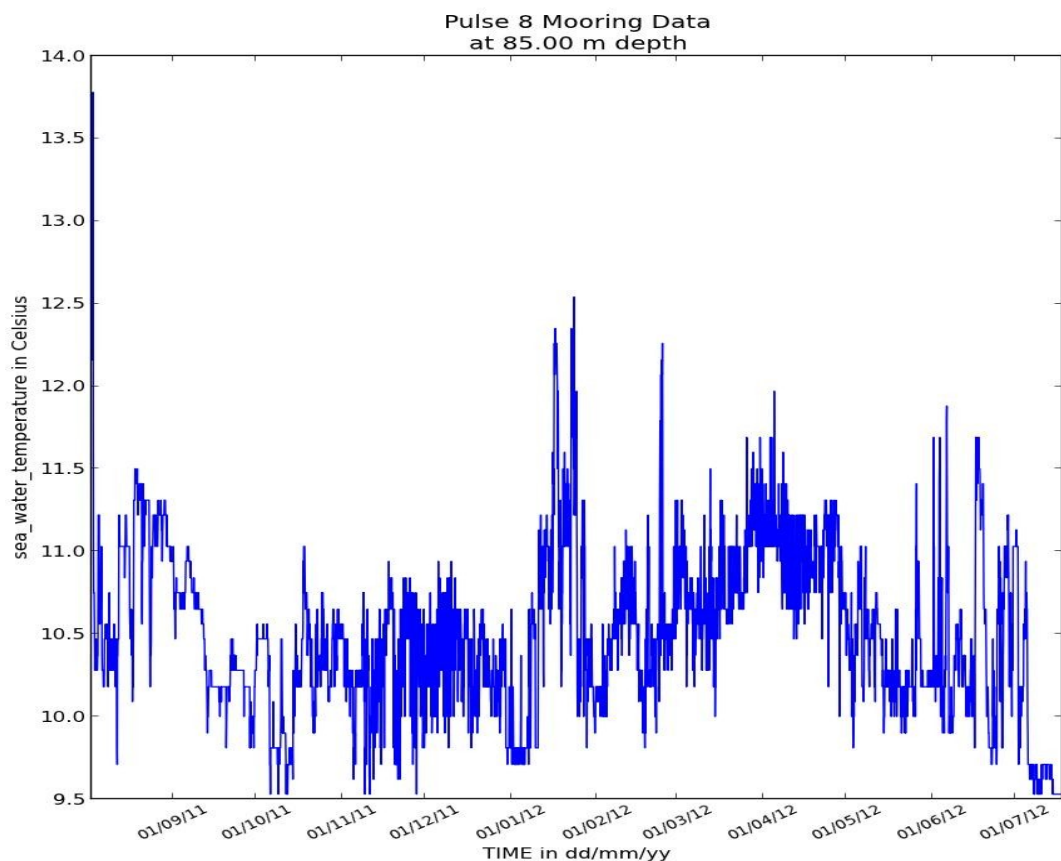
It is possible to see the abstract by typing in the Python command window

```
abstract
```

```
u"The Pulse 6 mooring was deployed from September 2009 to March 2010 at Lat -46.3224, Lon 140.6776. Moored instruments are deployed by the IMOS Australian Bluewater Observing System (ABOS) Southern Ocean Time Series sub-facility for time-series observations of physical, biological, and chemical properties, in the Sub-Antarctic Zone southwest of Tasmania, with yearly servicing. The Southern Ocean Time Series (SOTS) Sub-Facility is responsible for the deployment of Pulse moorings. These time-series observations are crucial to resolving ecosystem processes that affect carbon cycling, ocean productivity and marine responses to climate variability and change, ocean acidification and other stresses.'
```

Variables to modify :

- abos_URL : the opendap url of the chosen file
- tempDataStructure : we arbitrarily chose the variable called TEMP_85_1, but many more are available.



3.3 ACORN – Ocean Radar

The Australian Coastal Ocean Radar Network (ACORN) facility comprises a coordinated network of HF radars delivering real-time, non-quality controlled and delayed-mode, quality controlled surface current data into a national archive.

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/catalog/IMOS/ACORN/catalog.html>

Monthly aggregated files are also available in the following folders:

- monthly gridded 1h-avg-current-map QC
- monthly gridded 1h-avg-current-map non-QC

In the example below, we demonstrate how to use the netCDF4 module to plot velocity data for one time value only in a latitude / longitude grid.

```
from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### ACORN
acorn_URL = 'http://thredds.aodn.org.au/thredds/dodsC/IMOS/ACORN/monthly_gridded_1h-avg-current-map_non-
QC/TURQ/2012/IMOS_ACORN_V_20121001T000000Z_TURQ_FV00_monthly-1-hour-avg_END-20121029T180000Z_C-
20121030T160000Z.nc.gz'
acorn_DATA = Dataset(acorn_URL)
metadata = getAttNC(acorn_DATA)

speedData = acorn_DATA.variables['SPEED']
latData = acorn_DATA.variables['LATITUDE']
lonData = acorn_DATA.variables['LONGITUDE']
timeData = acorn_DATA.variables['TIME']

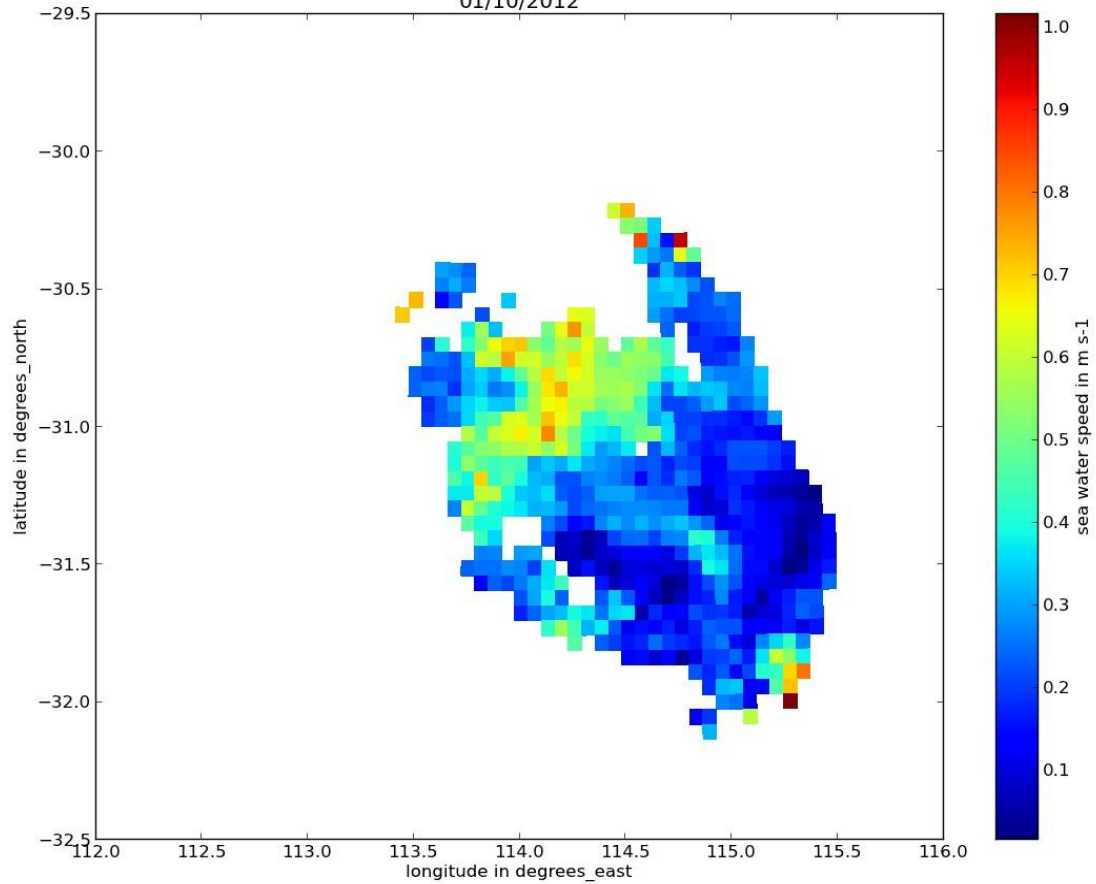
# Only one time value is being plotted. modify timeIndex if desired (value between 1 and length(timeData))
timeIndex = 5
figure1 = figure( figsize=(13, 10), dpi=80, facecolor='w', edgecolor='k')
pcolor(lonData[:,],latData[:,], speedData[timeIndex,:,:])
cbar = colorbar()
cbar.ax.set_ylabel(speedData.long_name + ' in ' + speedData.units)

title(metadata['title'] + '\n' + convertTime(timeData)[timeIndex].strftime('%d/%m/%Y'))
xlabel(lonData.long_name + ' in ' + lonData.units)
ylabel(latData.long_name + ' in ' + latData.units)
show()
```

Variables to modify :

- `acorn_URL` : the opendap url of the chosen file
- `timeIndex` : the time index number to plot

IMOS ACORN Turquoise Coast site (WA) (TURQ), monthly aggregation of one hour averaged current data
01/10/2012



3.4 ANFOG – Ocean Gliders

The Australian National Facility for Ocean Gliders (ANFOG), with IMOS/NCRIS funding, deploys a fleet of eight gliders around Australia.

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/catalog/IMOS/ANFOG/seaglider/catalog.html>

In the example below, we demonstrate how to use the netCDF4 module to plot salinity data as well as depth data in a same graph. Only the data points with a Quality Control flag greater than 1 (which means 'good data', please refers to IMOS NetCDF User Manual for a description of the Quality Control, available at http://imos.org.au/facility_manuals.html)

```

from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### ANFOG
anfog_URL =
'http://thredds.aodn.org.au/thredds/dodsC/IMOS/ANFOG/seaglider/SOTS20110420/IMOS_ANFOG_BCEOSTUV_2011042
0T111022Z_SG517_FV01_timeseries_END-20110420T140511Z.nc'
anfog_DATA = Dataset(anfog_URL)
metadata = getAttNC(anfog_DATA)

PSAL = anfog_DATA.variables['PSAL']
DEPTH = anfog_DATA.variables['DEPTH']
PSAL_qcFlag = anfog_DATA.variables['PSAL_quality_control']

qcLevel = 1 # we use the quality control flags to only select the good_data
index_qcLevel = where( PSAL_qcFlag[:] == qcLevel)

psalData = PSAL[index_qcLevel]
timeData = convertTime(anfog_DATA.variables['TIME'])[index_qcLevel]
depthData = DEPTH[index_qcLevel]

figure1 = figure( figsize=(13, 10), dpi=80, facecolor='w', edgecolor='k')

ax1 = figure1.add_subplot(111)
ax1.plot(timeData,psalData, 'b-')
ax1.set_xlabel('time (s)')
# Make the y-axis label and tick labels match the line color.
ax1.set_ylabel(PSAL.standard_name + ' in ' + PSAL.units, color='b')
for tl in ax1.get_yticklabels():
    tl.set_color('b')

ax2 = ax1.twinx()
ax2.plot(timeData,depthData, 'r.')
ax2.set_ylabel(DEPTH.standard_name + ' in ' + DEPTH.units, color='r')

```

```

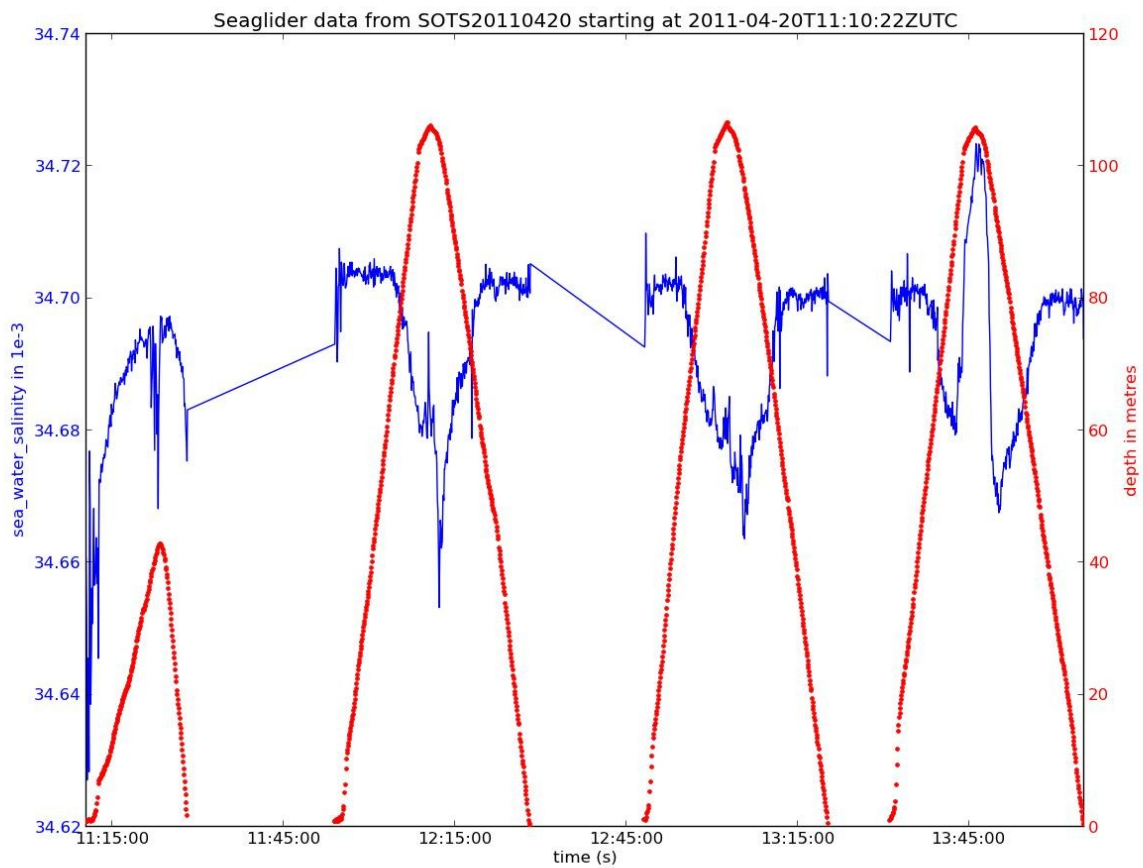
for tl in ax2.get_yticklabels():
    tl.set_color('r')

xlabel(anfog_DATA.variables['TIME'].standard_name)
title(metadata['title'] + ' starting at ' + metadata['time_coverage_start'] + 'UTC')
plt.show()

```

Variables to modify :

- `anfog_URL` : the opendap url of the chosen file
- `qcLevel` : quality control value (varies from 0 to 9)



3.5 ANMN – National Mooring Network

The Australian National Mooring Network Facility is a series of national reference stations and regional moorings designed to monitor particular oceanographic phenomena in Australian coastal ocean waters.

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/catalog/IMOS/ANMN/catalog.html>

In the example below, we demonstrate how to use the netCDF4 module to plot the U current variable measured with an ADCP instrument (in Western Australia).

```
from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### ANMN
anmn_URL = 'http://thredds.aodn.org.au/thredds/dodsC/IMOS/ANMN/WA/WATR50/Velosity/IMOS_ANMN-
WA_VATPE_20120516T040000Z_WATR50_FV01_WATR50-1205-Workhorse-ADCP-498_END-20121204T021500Z_C-
20121207T023956Z.nc'
anmn_DATA = Dataset(anmn_URL)
metadata = getAttNC(anmn_DATA)

UCUR = anmn_DATA.variables['UCUR']
DEPTH = anmn_DATA.variables['HEIGHT_ABOVE_SENSOR']

uCurrentData = UCUR[:]
timeData = convertTime(anmn_DATA.variables['TIME'])
depthData = DEPTH[:]

# creation of a observation/profile variable because pcolor can't handle a time object in the x axis
sizer = ones((1,len(depthData)), 'float')
profIndex = array(range(len(timeData)))
profIndex = profIndex.reshape(len(timeData),1)
prof_2D = profIndex * sizer

[depthData_mesh,prof_2D_mesh] = meshgrid(depthData,profIndex)

figure1 = figure( figsize=(13, 18), dpi=80, facecolor='w', edgecolor='k')
ax1 = subplot(211)

pcolor(prof_2D_mesh , depthData_mesh , uCurrentData[:,0,0])
clim(UCUR.valid_min, UCUR.valid_max)
cbar = colorbar()
cbar.ax.set_ylabel(UCUR.long_name + ' in ' + UCUR.units)
```



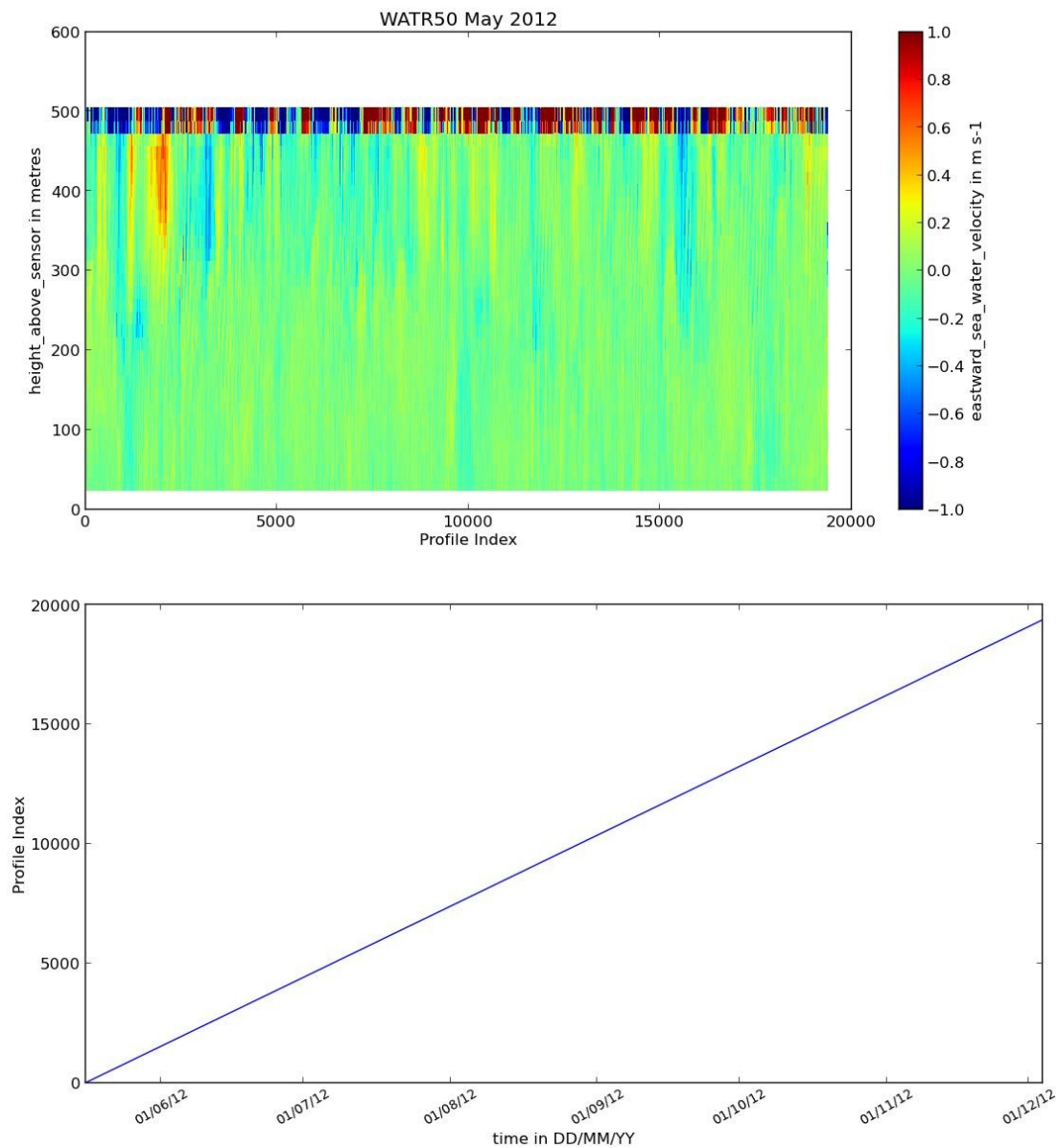
```
title(metadata['title'])
xlabel('Profile Index')
ylabel(DEPTH.long_name + ' in ' + DEPTH.units)

# plot profile index with time
ax2 = subplot(212)
plot(timeData, profIndex)
ylabel('Profile Index')
xlabel(anmn_DATA.variables['TIME'].long_name + ' in DD/MM/YY')

from matplotlib.dates import MONTHLY, DateFormatter, rrulewrapper, RRuleLocator
rule = rrulewrapper(MONTHLY, bymonthday=1, interval=1)
formatter = DateFormatter('%d/%m/%y')
loc = RRuleLocator(rule)

ax2.xaxis.set_major_locator(loc)
ax2.xaxis.set_major_formatter(formatter)
labels = ax2.get_xticklabels()
setp(labels, rotation=30, fontsize=10)

plt.show()
```



Variables to modify :

- `anmn_URL` : the opendap url of the chosen file. (The example URL used here may not work if the file has been replaced by a newer version. A currently available file can be selected as described in section 1.2 above.)

3.6 AUV – Autonomous Underwater Vehicle

The IMOS Autonomous Underwater Vehicle (AUV) Facility operates an ocean going AUV called Sirius capable of undertaking high resolution, geo-referenced survey work.

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/catalog/IMOS/AUV/catalog.html>

In the example below, the netCDF4 module is used to extract depth, temperature, and time data and then produce a multiple time-series plot showing the variation of water temperature with depth and time during the robot's dive.

```

from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### AUV
auv_URL =
'http://thredds.aodn.org.au/thredds/dodsC/IMOS/AUV/GBR201102/r20110301_012810_station1195_09_transect/hydro_netcdf/IMOS_AUV_ST_20110301T012815Z_SIRIUS_FV00.nc'
auv_DATA = Dataset(auv_URL)
metadata = getAttNC(anmn_DATA)

tempData = auv_DATA.variables['TEMP']
timeData = convertTime(auv_DATA.variables['TIME'])
depthData = auv_DATA.variables['DEPTH']

averageLat = auv_DATA.variables['LATITUDE'][:].mean()
averageLon = auv_DATA.variables['LONGITUDE'][:].mean()

figure1 = figure( figsize=(10, 7), dpi=80, facecolor='w', edgecolor='k')

ax1 = figure1.add_subplot(111)
ax1.plot(timeData,tempData[:], 'b-')
ax1.set_xlabel('time (s)')
# Make the y-axis label and tick labels match the line color.
ax1.set_ylabel(tempData.standard_name + ' in ' + tempData.units, color='b')
for tl in ax1.get_yticklabels():
    tl.set_color('b')

ax2 = ax1.twinx()
ax2.plot(timeData,depthData[:], 'r.')
ax2.set_ylabel(depthData.standard_name + ' in ' + depthData.units, color='r')
for tl in ax2.get_yticklabels():
    tl.set_color('r')

```

```

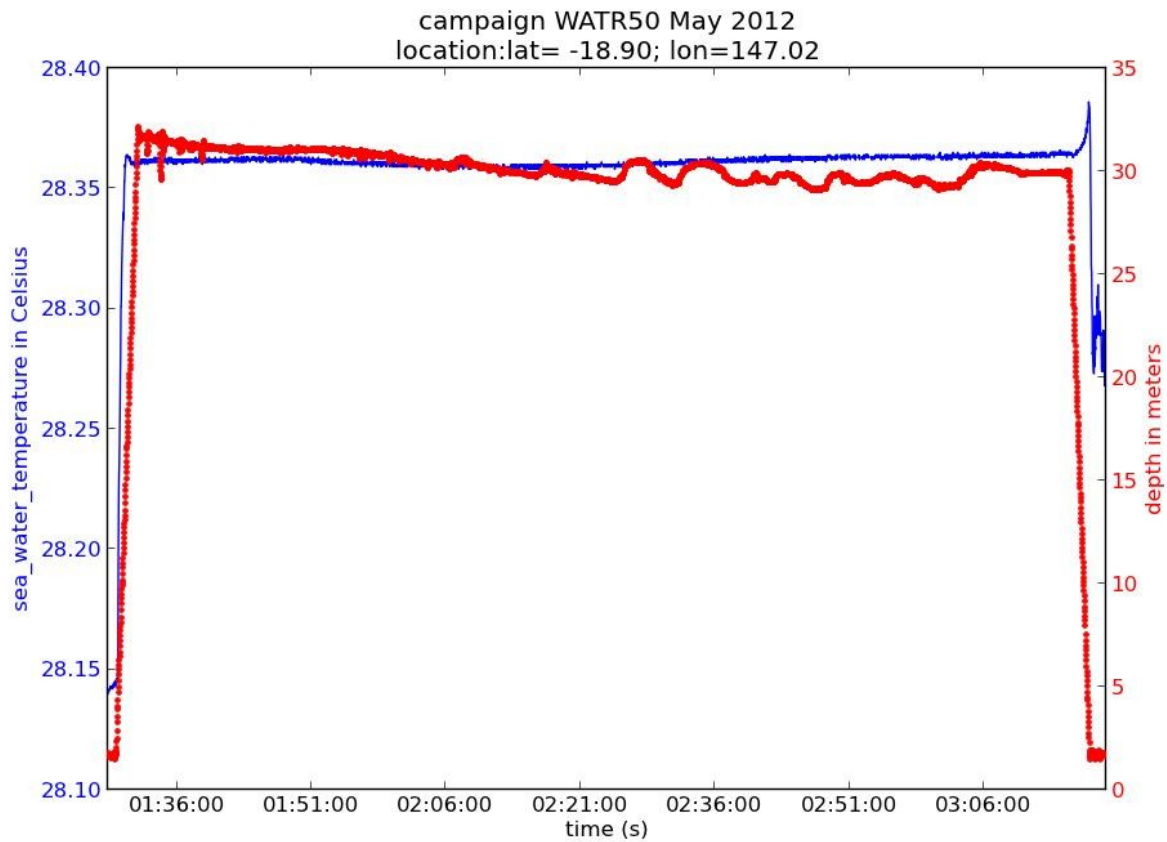
xlabel(auv_DATA.variables['TIME'].standard_name)

title('campaign ' + metadata['title'] + '\nlocation:lat= ' + "%0.2f" % averageLat + '; lon=' + "%0.2f" % averageLon )
plt.show()

```

Variables to modify :

- `auv_URL` : the opendap url of the chosen file



3.7 Argo – Argo Floats Program

Argo floats have revolutionised our understanding of the broad scale structure of the oceans to 2000 m depth. In the past 10 years more high resolution hydrographic profiles have been provided by Argo floats than from the rest of the observing system put together. Each Argo float is identified by a unique identification number called a WMO ID.

NetCDF files can be found at :

http://thredds.aodn.org.au/thredds/catalog/IMOS/Argo/aggregated_datasets/catalog.html

In the examples below, we demonstrate how to use the netCDF4 module to plot Argo data from an aggregated file (One file per year per basin : Atlantic, Indian, Pacific North, Pacific South).

```

from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### Argo
argo_URL =
'http://thredds.aodn.org.au/thredds/dodsC/IMOS/Argo/aggregated_datasets/south_pacific/IMOS_Argo_TPS-
20020101T000000_FV01_yearly-aggregation-South_Pacific_C-20121102T220000Z.nc'
argo_DATA = Dataset(argo_URL)
metadata = getAttNC(argo_DATA)

nProfData = len(argo_DATA.dimensions['N_PROF']) #Number of profiles contained in the file.
nLevelData = len(argo_DATA.dimensions['N_LEVELS']) #Maximum number of pressure levels contained in a profile.

# we list all the argo floats number in the variable 'argoFloatNumber' and
#chose one value
argoFloatNumber = unique(argo_DATA.variables['PLATFORM_NUMBER'][:])

argoFloatNumberChosen = 5900106 # we randomly chose one float number
# we load the data for this float

argoFloatProfilesIndexes = argo_DATA.variables['PLATFORM_NUMBER'][:] == argoFloatNumberChosen
tempData = argo_DATA.variables['TEMP_ADJUSTED'][argoFloatProfilesIndexes]
psalData = argo_DATA.variables['PSAL_ADJUSTED'][argoFloatProfilesIndexes]
presData = argo_DATA.variables['PRES_ADJUSTED'][argoFloatProfilesIndexes]
latProfile = argo_DATA.variables['LATITUDE'][argoFloatProfilesIndexes]
lonProfile = argo_DATA.variables['LONGITUDE'][argoFloatProfilesIndexes]
timeProfile = convertTime(argo_DATA.variables['JULD'])[argoFloatProfilesIndexes]

# creation of a profile variable array
nProfForFloat = sum(argoFloatProfilesIndexes == True)
sizer = ones((1,nLevelData),'float')
profIndex = array(range(nProfForFloat))
profIndex = profIndex.reshape(nProfForFloat,1)

```

```

prof_2D = profIndex * sizar

figure1 = figure(num=None, figsize=(15, 10), dpi=80, facecolor='w', edgecolor='k')
subplot(311)
pcolor(prof_2D, -presData, tempData)
cbar = colorbar()
cbar.ax.set_ylabel(argo_DATA.variables['TEMP_ADJUSTED'].long_name + '\n in ' +
argo_DATA.variables['TEMP_ADJUSTED'].units)
xlabel('Profile Index')
ylabel(argo_DATA.variables['PRES_ADJUSTED'].long_name + ' in negative ' +
argo_DATA.variables['PRES_ADJUSTED'].units)

title(metadata['description'] + '\nArgo Float Number : ' + "%0.0f" % argoFloatNumberChosen )

from matplotlib.dates import MONTHLY, DateFormatter, rrulewrapper, RRuleLocator
rule = rrulewrapper(MONTHLY, bymonthday=1, interval=1)
formatter = DateFormatter('%d/%m/%y')
loc = RRuleLocator(rule)

#plot the LON timeseries
ax3 = subplot(234)
plot(timeProfile, lonProfile)
ax3.xaxis.set_major_locator(loc)
ax3.xaxis.set_major_formatter(formatter)
labels = ax3.get_xticklabels()
setp(labels, rotation=30, fontsize=10)
xlabel(argo_DATA.variables['JULD'].long_name + ' in ' + 'dd/mm/yy' )
ylabel(argo_DATA.variables['LONGITUDE'].long_name + ' in ' + argo_DATA.variables['LONGITUDE'].units)

#plot the LAT timeseries
ax4 = subplot(235)
plot(timeProfile, latProfile)
ax4.xaxis.set_major_locator(loc)
ax4.xaxis.set_major_formatter(formatter)
labels = ax4.get_xticklabels()
setp(labels, rotation=30, fontsize=10)
xlabel(argo_DATA.variables['JULD'].long_name + ' in ' + 'dd/mm/yy' )
ylabel(argo_DATA.variables['LATITUDE'].long_name + ' in ' + argo_DATA.variables['LATITUDE'].units)

#plot the profile index with time values
ax5 = subplot(236)
plot(timeProfile, profIndex)
ax5.xaxis.set_major_locator(loc)
ax5.xaxis.set_major_formatter(formatter)
labels = ax5.get_xticklabels()
setp(labels, rotation=30, fontsize=10)
xlabel(argo_DATA.variables['JULD'].long_name + ' in ' + 'dd/mm/yy' )
ylabel('Profile Index')

#plot of a single profile

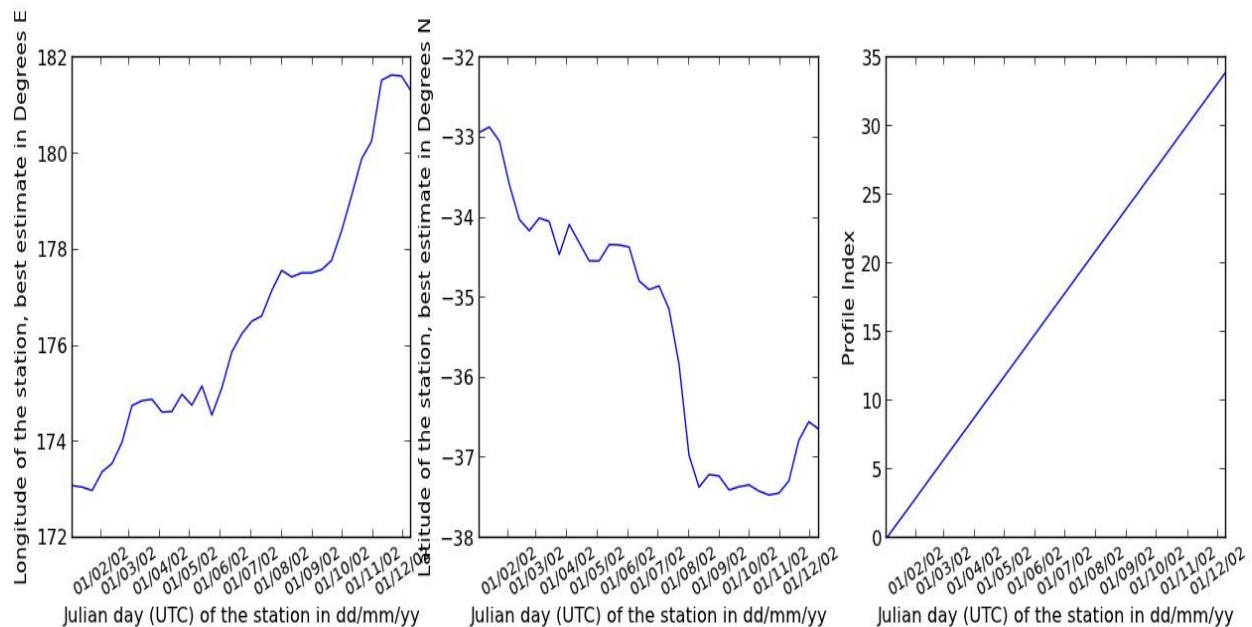
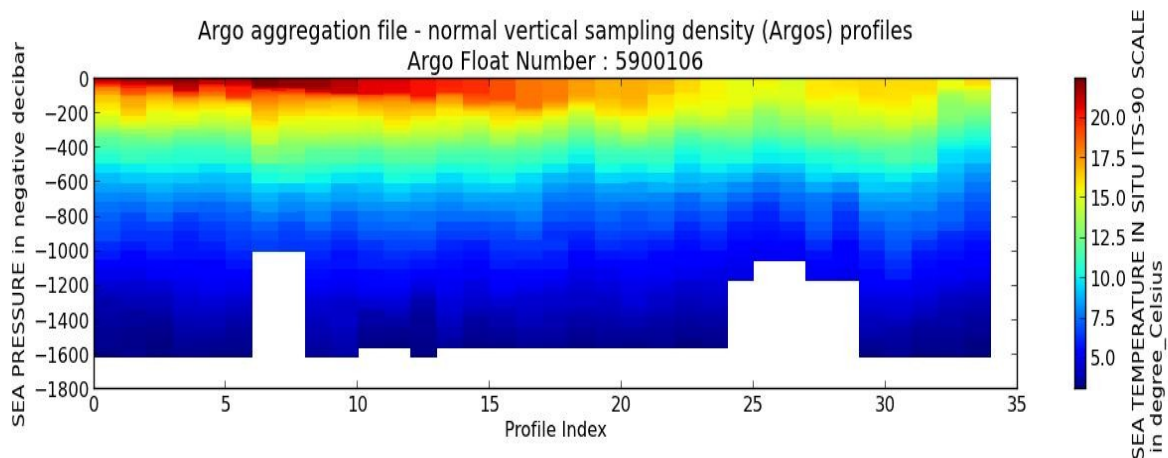
```

```

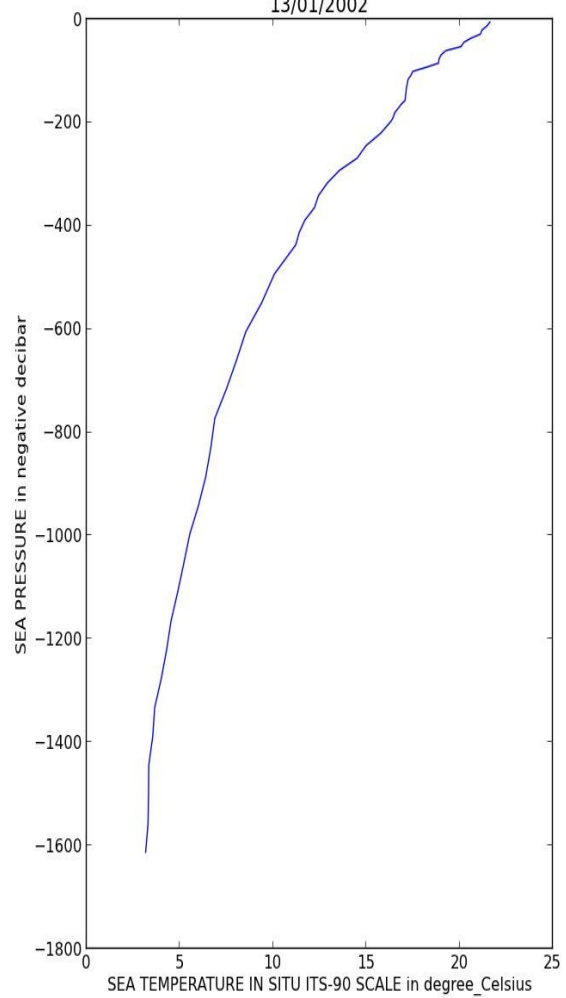
profileToPlot = 1# this is arbitrary. We can plot all profiles from 1 to nProfiles, modify profileToPlot if desired
figure2 = figure(num=None, figsize=(7, 10), dpi=80, facecolor='w', edgecolor='k')
plot (tempData[profileToPlot,:],-presData[profileToPlot,:])
title(metadata['description'] + '\nlocation ' + "%0.2f" % latProfile[profileToPlot] + '/' + "%0.2f" %
lonProfile[profileToPlot] + '\n' + timeProfile[profileToPlot].strftime('%d/%m/%Y'))
xlabel(argo_DATA.variables['TEMP_ADJUSTED'].long_name + ' in ' + argo_DATA.variables['TEMP_ADJUSTED'].units)
ylabel(argo_DATA.variables['PRES_ADJUSTED'].long_name + ' in negative ' +
argo_DATA.variables['PRES_ADJUSTED'].units)

plt.show()

```



Argo aggregation file - normal vertical sampling density (Argos) profil
location -32.86/173.06
13/01/2002



3.8 FAIMMS – Wireless Sensor Networks

The IMOS Facility for Intelligent Monitoring of Marine Systems is a sensor network established in the Great Barrier Reef off the coast of Queensland, Australia. A 'sensor network' is an array of small, wirelessly interconnected sensors that collectively stream sensor data to a central data aggregation point. Sensor networks can be used to provide spatially dense bio-physical measurements in real-time.

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/catalog/IMOS/FAIMMS/catalog.html>

In the example below, we demonstrate how to use the netCDF4 module to plot a temperature time-series. Only data points which have a flag value equal to 1 are used (which means 'good data', please refers to IMOS NetCDF User Manual for a description of the Quality Control, available at http://imos.org.au/facility_manuals.html).

```
from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### FAIMMS
FAIMMS_URL =
'http://thredds.aodn.org.au/thredds/dodsC/IMOS/FAIMMS/Myrmidon_Reef/Sensor_Float_1/water_temperature/sea_wat
er_temperature@5.0m_channel_114/2012/QAQC/IMOS_FAIMMS_T_20121201T000000Z_FV01_END-
20130101T000000Z_C-20130426T102459Z.nc'
faimms_DATA = Dataset(FAIMMS_URL)
metadata = getAttNC(faimms_DATA)

qcLevel = 1

TEMP = faimms_DATA.variables['TEMP'][:]
TEMP_qcFlag = faimms_DATA.variables['TEMP_quality_control']
index_qcLevel = where( TEMP_qcFlag[:,0,0] == qcLevel)

timeData = convertTime(faimms_DATA.variables['TIME'])(index_qcLevel)
tempData = TEMP[index_qcLevel[0][:],0,0]

figure1 = figure(num=None, figsize=(15, 10), dpi=80, facecolor='w', edgecolor='k')
ax1 = subplot(111)
plot (timeData,tempData[:,])

title(metadata['title'] + '\n' + "%0.2f" % faimms_DATA.variables['TEMP'].sensor_depth + ' m depth' + '\nlocation:lat='
+ "%0.2f" % faimms_DATA.variables['LATITUDE'][:] + '; lon=' + "%0.2f" % faimms_DATA.variables['LONGITUDE'][:] )
xlabel( faimms_DATA.variables['TIME'].long_name)
ylabel( faimms_DATA.variables['TEMP'].standard_name + ' in ' + faimms_DATA.variables['TEMP'].units)

from matplotlib.dates import DAILY, DateFormatter, rrulewrapper, RRuleLocator
```

```

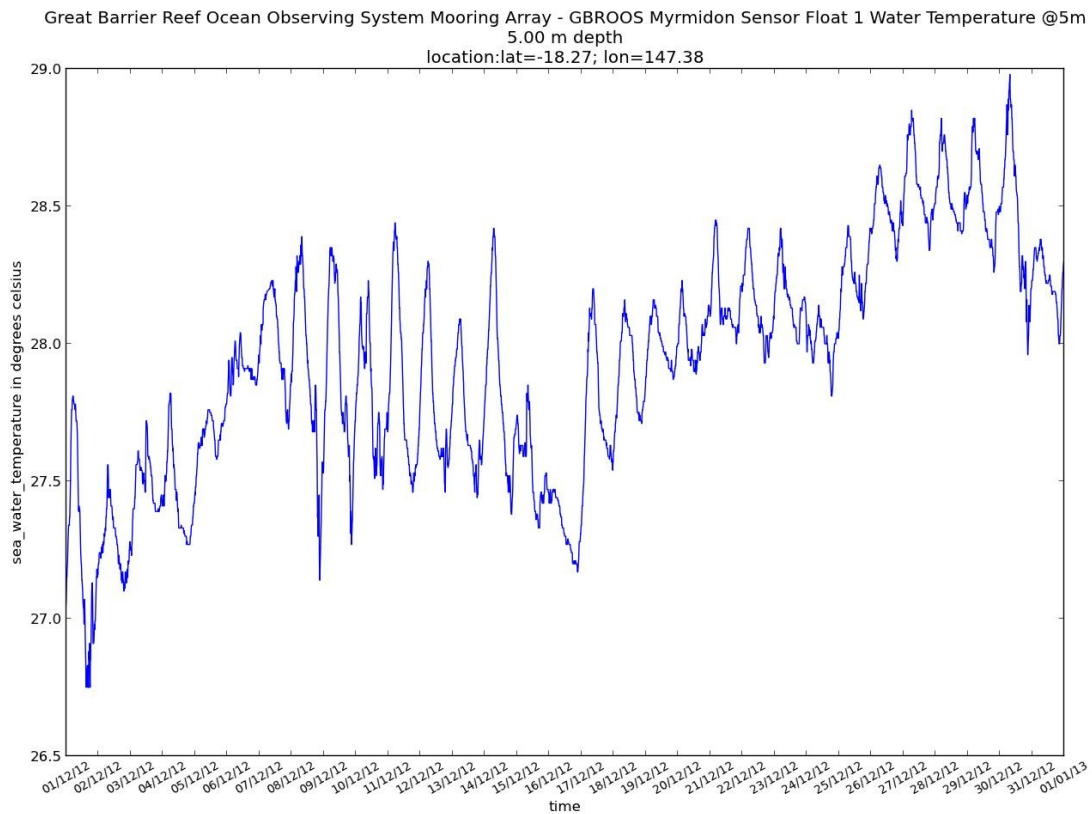
rule = rrulewrapper(DAILY, interval=1)
formatter = DateFormatter('%d/%m/%y')
loc = RRuleLocator(rule)
ax1.xaxis.set_major_locator(loc)
ax1.xaxis.set_major_formatter(formatter)
labels = ax1.get_xticklabels()
setp(labels, rotation=30, fontsize=10)

plt.show()

```

Variables to modify :

- `faimms_URL` : the opendap url of the chosen file
- `qcLevel` : quality control value (varies from 0 to 9)



3.9 SRS – Satellite Remote Sensing

3.9.1 Bio-Optical database – Pigment data

The bio-optical data base underpins the assessment of ocean colour products in the Australian region (e.g. chlorophyll a concentrations, phytoplankton species composition and primary production).

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/catalog/IMOS/SRS/BioOptical/catalog.html>

In the example below, we demonstrate how to use the netCDF4 module to plot a Chlorophyll-a profile (High Performance Liquid Chromatography of pigments in discrete sea-water samples)

```
from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### BioOptic pigment
srs_pigment_URL = 'http://thredds.aodn.org.au/thredds/dodsC/IMOS/SRS/BioOptical/1997_cruise-
FR1097/pigment/IMOS_SRS-OC-BODBAW_X_19971201T052600Z_FR1097-pigment_END-19971207T220700Z_C-
20121129T120000Z.nc'
srs_pigment = Dataset(srs_pigment_URL)
metadata = getAttNC(srs_pigment)

nProfiles = len(srs_pigment.dimensions['profile'])
# we choose the first profile
ProfileToPlot = 9 # this is arbitrary. We can plot all profiles from 0 to nProfiles
nObsProfile = srs_pigment.variables['rowSize'][ProfileToPlot] #number of observations for ProfileToPlot
timeProfile = convertTime(srs_pigment.variables['TIME'][ProfileToPlot])
latProfile = srs_pigment.variables['LATITUDE'][ProfileToPlot]
lonProfile = srs_pigment.variables['LONGITUDE'][ProfileToPlot]

# we look for the observations indexes related to the choosen profile
indexObservationStart = sum( srs_pigment.variables['rowSize'][range(0,ProfileToPlot)])
indexObservationEnd = sum(srs_pigment.variables['rowSize'][range(0,ProfileToPlot+1)])
indexObservation = range(indexObservationStart,indexObservationEnd )

cphl_aData = srs_pigment.variables['CPHL_a'][indexObservation] # for ProfileToPlot
depthData = srs_pigment.variables['DEPTH'][indexObservation]

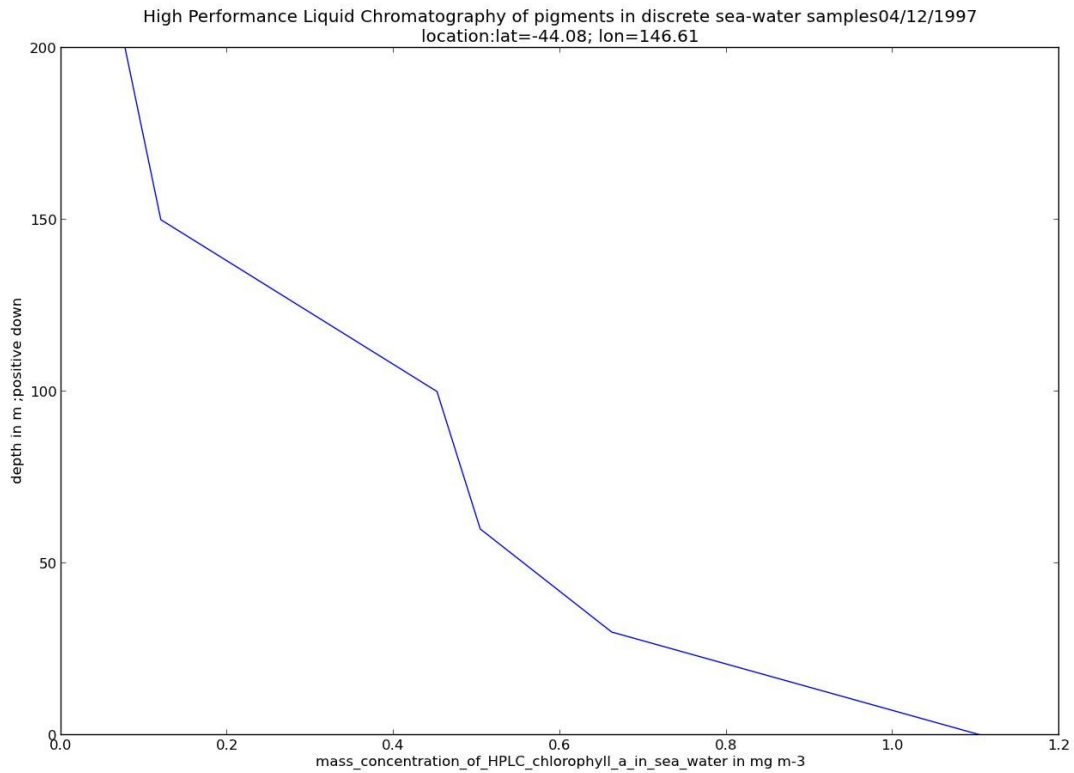
figure1 = figure(num=None, figsize=(15, 10), dpi=80, facecolor='w', edgecolor='k')
plot (cphl_aData,depthData)

title(metadata['source'] + timeProfile.strftime('%d/%m/%Y') + '\nlocation:lat=' + "%0.2f" % latProfile + '; lon=' +
"%0.2f" %lonProfile )
xlabel(srs_pigment.variables['CPHL_a'].long_name + ' in ' + srs_pigment.variables['CPHL_a'].units)
```

```
ylabel( srs_pigment.variables['DEPTH'].long_name + ' in ' + srs_pigment.variables['DEPTH'].units + ';positive ' +  
srs_pigment.variables['DEPTH'].positive )  
  
plt.show()
```

Variables to modify :

- `srs_URL` : the opendap url of the chosen file
- `ProfileToPlot` : the profile number to plot



3.9.2 Bio-Optical database – Absorption data

The bio-optical data base underpins the assessment of ocean colour products in the Australian region (e.g. chlorophyll a concentrations, phytoplankton species composition and primary production).

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/catalog/IMOS/SRS/BioOptical/catalog.html>

In the example below, we demonstrate how to use the netCDF4 module to plot (1) the variation of Absorption coefficients of CDOM (gilvin) in discrete sea-water samples at different wavelengths and (2) the variation of absorption coefficients of CDOM at different wavelengths and different depths.

```
from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### BioOptic absorption
srs_absorption_URL = 'http://thredds.aodn.org.au/thredds/dodsC/IMOS/SRS/BioOptical/1997_cruise-
FR1097/absorption/IMOS_SRS-OC-BODBAW_X_19971201T052600Z_FR1097-absorption-CDOM_END-
19971207T180500Z_C-20121129T130000Z.nc'
srs_absorption = Dataset(srs_absorption_URL)
metadata = getAttNC(srs_absorption)

nProfiles = len(srs_absorption.dimensions['profile'])
# we choose the first profile
ProfileToPlot = 9 # this is arbitrary. We can plot all profiles from 0 to nProfiles
nObsProfile = srs_absorption.variables['rowSize'][ProfileToPlot] #number of observations for ProfileToPlot
timeProfile = convertTime(srs_absorption.variables['TIME'][ProfileToPlot])
latProfile = srs_absorption.variables['LATITUDE'][ProfileToPlot]
lonProfile = srs_absorption.variables['LONGITUDE'][ProfileToPlot]

# we look for the observations indexes related to the choosen profile
indexObservationStart = sum( srs_absorption.variables['rowSize'][range(0,ProfileToPlot)])
indexObservationEnd = sum(srs_absorption.variables['rowSize'][range(0,ProfileToPlot+1)])
indexObservation = range(indexObservationStart,indexObservationEnd )

agData = srs_absorption.variables['ag'][indexObservation,:]
wavelengthData = srs_absorption.variables['wavelength']
depthData = srs_absorption.variables['DEPTH'][indexObservation]
[wavelengthData_mesh,depthData_mesh] = meshgrid(wavelengthData,depthData)

figure1 = figure(num=None, figsize=(15, 10), dpi=80, facecolor='w', edgecolor='k')
pcolor(wavelengthData_mesh , depthData_mesh , agData)
pcolor(profile_2D, -presData, tempData)
cbar = colorbar()
cbar.ax.set_ylabel(srs_absorption.variables['ag'].long_name + '\n in ' + srs_absorption.variables['ag'].units)
```

```

title(metadata['source'])
xlabel( srs_absorption.variables['wavelength'].long_name + ' in: ' + srs_absorption.variables['wavelength'].units)
ylabel(srs_absorption.variables['DEPTH'].long_name + ' in ' + srs_absorption.variables['DEPTH'].units + '; positive
'+srs_absorption.variables['DEPTH'].positive )

##
nDepth = len(depthData)
figure2 = figure(num=None, figsize=(15, 10), dpi=80, facecolor='w', edgecolor='k')

labels = []
for iplot in range(shape(agData)[0]):
    plot(wavelengthData[:,agData[iplot,:]],agData[iplot,:],'x')
    labels.append(r'Depth = %i m' % depthData[iplot])

plt.legend(labels,loc='upper right')

unitsMainVar=char(srs_absorption.variables.ag.units)
ylabel(srs_absorption.variables['ag'].long_name + ' in: ' + srs_absorption.variables['ag'].units)
xlabel( srs_absorption.variables['wavelength'].long_name + ' in: ' + srs_absorption.variables['wavelength'].units)

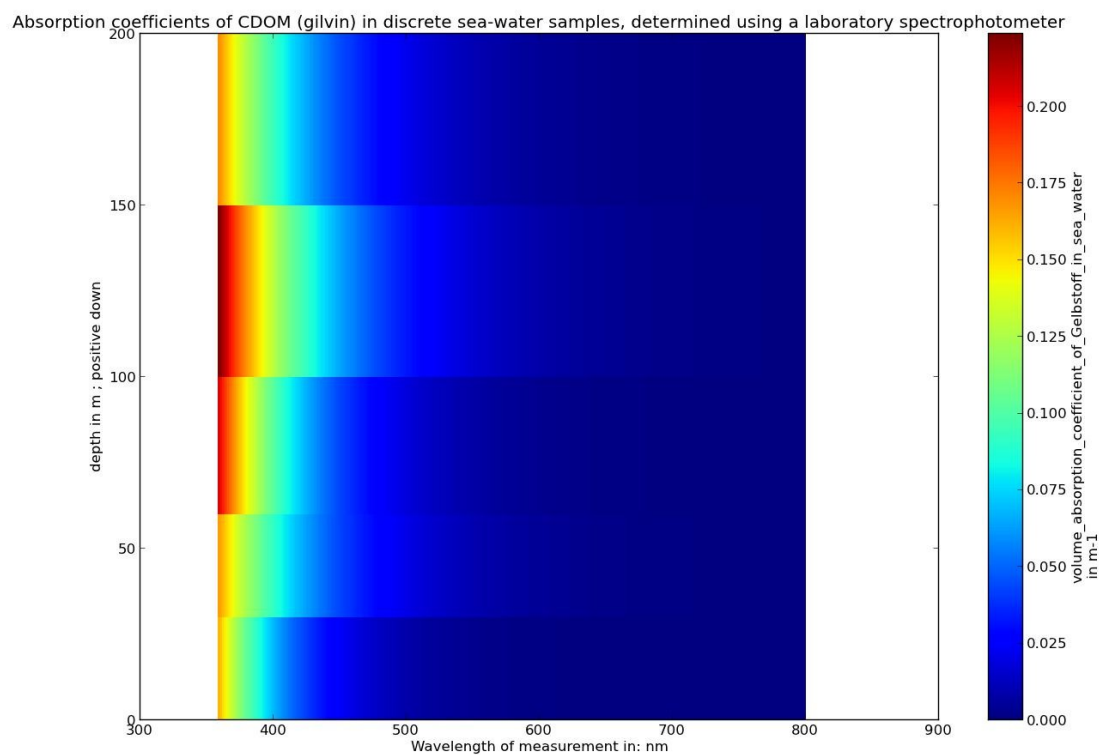
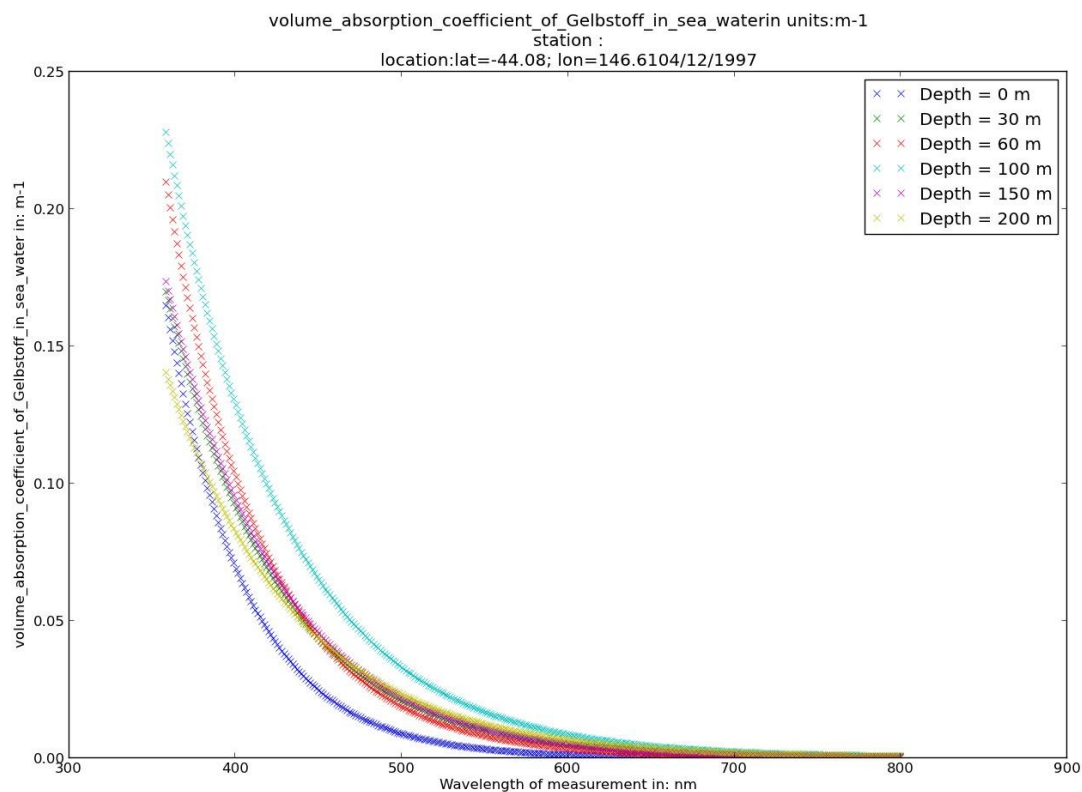
title(srs_absorption.variables['ag'].long_name + 'in units:' + srs_absorption.variables['ag'].units + '\nstation :'+
'\nlocation:lat=' + "%0.2f" % latProfile + '; lon=' + "%0.2f" %lonProfile + timeProfile.strftime('%d/%m/%Y') )

plt.show()

```

Variables to modify :

- srs_URL : the opendap url of the chosen file
- ProfileToPlot : the profile number to plot



3.9.3 GHR SST – L3P mosaic

Please refer to the SRS product Help page : http://portalhelp.aodn.org.au/Portal2_help/?q=node/149

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/dodsC/IMOS/SRS/GHR SST-SSTsubskin/>

In the example below, we demonstrate how to use the netCDF4 module to plot the Sea Surface Temperature from a gridded data product.

```
from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### GHR SST - L3P mosaic
srs_L3P_URL = 'http://thredds.aodn.org.au/thredds/dodsC/IMOS/SRS/GHR SST-SSTsubskin/2013/20130315-ABOM-
L3P_GHR SST-SSTsubskin-AVHRR_MOSAIC_01km-AO_DAAC-v01-fv01_0.nc'
srs_L3P_DATA = Dataset(srs_L3P_URL)
metadata = getAttNC(srs_L3P_DATA)

step = 20 # we take one point out of 'step'. Only to make it faster to plot
sst = srs_L3P_DATA.variables['sea_surface_temperature'][:,0::step,::step]
lat = srs_L3P_DATA.variables['lat'][:,::step]
lon = srs_L3P_DATA.variables['lon'][:,::step]
[lon_mesh,lat_mesh] = meshgrid(lon,lat)

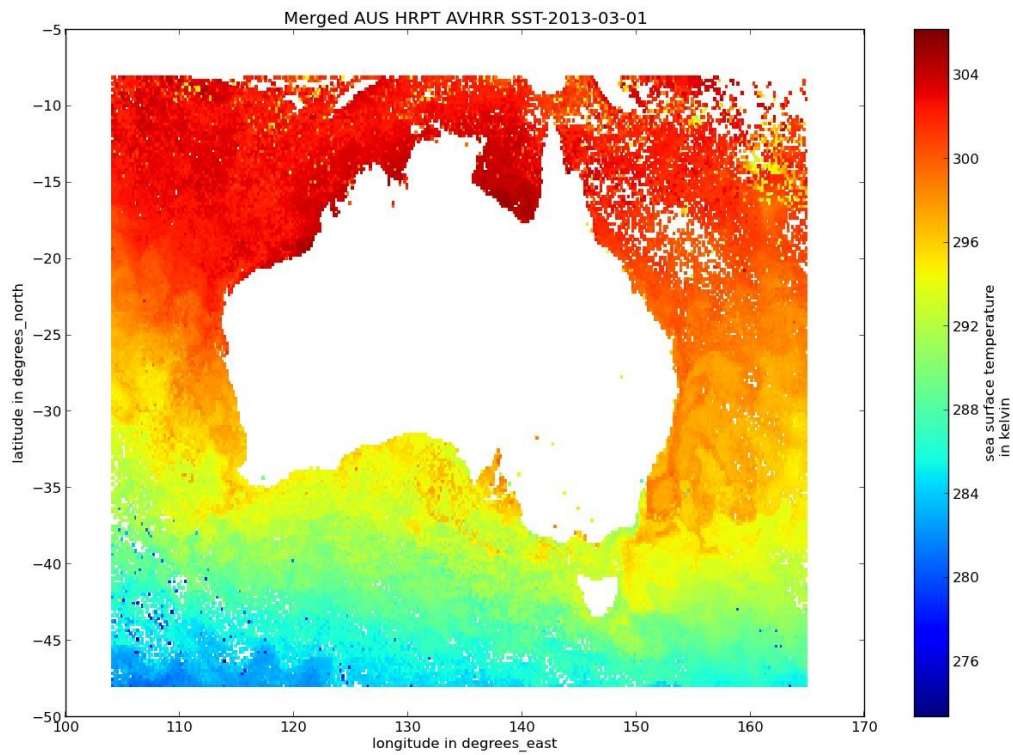
figure1 = figure(num=None, figsize=(15, 10), dpi=80, facecolor='w', edgecolor='k')
pcolor(lon_mesh,lat_mesh ,sst)

title( metadata['title'] + '-' + metadata['start_date'] )
xlabel(srs_L3P_DATA.variables['lon'].long_name + ' in ' + srs_L3P_DATA.variables['lon'].units)
ylabel(srs_L3P_DATA.variables['lat'].long_name + ' in ' + srs_L3P_DATA.variables['lat'].units)

cbar = colorbar()
cbar.ax.set_ylabel(srs_L3P_DATA.variables['sea_surface_temperature'].long_name + '\n in ' +
srs_L3P_DATA.variables['sea_surface_temperature'].units)
plt.show()
```


Variables to modify :

- `srs_URL` : the opendap url of the chosen file
- `step` : a number to lower the resolution. This helps to reduce memory issues.



3.9.4 GHR SST – L3S – multi swath, multi sensor, one day

Please refer to the SRS product Help page : http://portalhelp.aodn.org.au/Portal2_help/?q=node/149

NetCDF files can be found at :

<http://thredds.aodn.org.au/thredds/dodsC/IMOS/SRS/SRS-SST/L3S-01day/>

In the example below, we demonstrate how to use the netCDF4 module to plot the Sea Surface Temperature from a gridded data product.

```

from netCDF4 import Dataset
from datetime import datetime, timedelta
from pylab import *
import numpy
import matplotlib.pyplot as plt
from imosNetCDF import *

##### GHR SST - L3S mosaic
srs_L3S_URL = 'http://thredds.aodn.org.au/thredds/dodsC/IMOS/SRS/SRS-SST/L3S-
01day/L3S_1d_night/2013/20130401152000-ABOM-L3S_GHR SST-SSTskin-AVHRR_D-1d_night-v02.0-fv01.0.nc.gz'
srs_L3S_DATA = Dataset(srs_L3S_URL)
metadata = getAttNC(srs_L3S_DATA)

step = 5 # we take one point out of 'step'. Only to make it faster to plot
sst = srs_L3S_DATA.variables['sea_surface_temperature'][:,::step,::step]
lat = srs_L3S_DATA.variables['lat'][:,::step]
lon = srs_L3S_DATA.variables['lon'][:,::step]

if sum(lon<0) > 0:
    lon[lon<0] = lon[lon<0]+360

[lon_mesh,lat_mesh] = meshgrid(lon,lat)

figure1 = figure(num=None, figsize=(15, 10), dpi=80, facecolor='w', edgecolor='k')
pcolor(lon_mesh,lat_mesh ,sst)
title( metadata['title'] + '-' + metadata['start_time'] )
xlabel(srs_L3S_DATA.variables['lon'].long_name + ' in ' + srs_L3S_DATA.variables['lon'].units)
ylabel(srs_L3S_DATA.variables['lat'].long_name + ' in ' + srs_L3S_DATA.variables['lat'].units)

cbar = colorbar()
cbar.ax.set_ylabel(srs_L3S_DATA.variables['sea_surface_temperature'].long_name + '\n in ' +
srs_L3S_DATA.variables['sea_surface_temperature'].units)
plt.show()

```

Variables to modify :

- `srs_URL` : the opendap url of the chosen file
- `step` : a number to lower the resolution. This helps to reduce memory issues.

