

METEOROLOGICAL SATELLITE : SENSING PRINCIPLES, DATA COLLECTION AND ANALYSIS

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March - 2022

Advance Training in Meteorological Instrumentation & Information System Batch No.IX

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CERTIFICATE

This is to certified that the project report title "Meteorological satellite: sensing princi-

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Place: Garoua

Date:

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ACKNOWLEDGEMENT

I am grateful to Mr. TCHAMABO Urbain, Forecaster - MEng Degree in Meteorology, who has guided me on this project with his immense knowledge and experience.

I am grateful to officers and staffs of training section for the support extended throughout the training period.

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INTRODUCTION

Weather monitoring and forecasting was one of the first civilian applications of satellite remote sensing, dating back to the first true weather satellite, TIROS-1 launched in 1960 by the United States. Several other weather satellites were launched over the next five years, in near-polar orbits, providing repetitive coverage of global weather patterns. In 1966, National Aeronautics and Space Administration (NASA) launched the geostationary Applications Technology Satellite (ATS-1) which provided hemispheric images of the Earth's surface and cloud cover every half hour. For the first time, the development and movement of weather systems could be routinely monitored. Today, several countries (India, China, Russia, Japan) operate weather, or meteorological satellites to monitor weather conditions around the globe. Generally speaking, these satellites use sensors which have fairly coarse spatial resolution (when compared to systems for observing land) and provide large areal coverage. Their temporal resolutions are generally quite high, providing frequent observations of the Earth's surface, atmospheric moisture, and cloud cover, which allows for near continuous monitoring of global weather conditions, and hence forecasting.

The current project report Meteorological satellite: Sensing principles, data collection and analysis is structured as outlined below.

The first chapter is a brief overview of processes and steps involve in satellite remote sensing.

The second chapter is a case study, we choose GOES-R series satellites, which cover East Africa countries and is used by meteorologists for weather monitoring and forecasting.

The third chapter deals about different ways to access GOES-R series data and visualise a single image band.

The fourth chapter covers satellite data analysis, we will process three hours satellite data from GLDAS and compare with data obtained from direct observation at aviation meteorolgical stations.

The fith chapter presents the results and interpretation that we can derive from our analysis.

Satellite data are large amount of data, so analysis requires computer hardware with good capabilities, softwares and programming skills.

For completion of the current work, we use the following tools:

Computer harward: HP Probook, Intel R Core i5, RAM 16 GB.

Operating systems: Ubuntu 20.04.4

Processing: GRASS GISS 7.8 with bash script for worklow automation.

Visualisation: python 3.8 and QGIS 16 LTR.

BASIC PRINCIPLES OF SATELLITE REMOTE SENSING

1.1 Definition

Remote sensing is the science (and to some extent, art) of acquiring information about the Earth's surface without actually being in contact with it. This is done by sensing and recording reflected or emitted energy and processing, analysing, and applying that information.

In much of remote sensing, the process involves an interaction between incident radiation and the targets of interest. This is exemplified by the use of imaging systems where the following seven steps are involved. However that remote sensing also involves the sensing of emitted energy and the use of non-imaging sensors.

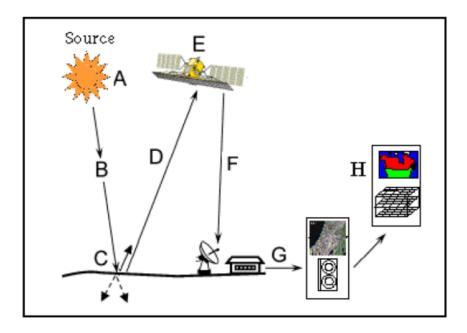


Figure 1.1: Remote sensing process

Step 1-Energy Source or Illumination (A): the first requirement for remote sensing is to have an energy source which illuminates or provides electromagnetic energy to the target of interest.

Step 2-Radiation and the Atmosphere (B): as the energy travels from its source to the

target, it will come in contact with and interact with the atmosphere it passes through. This interaction may take place a second time as the energy travels from the target to the sensor.

- **Step 3-Interaction with the Target (C)**: once the energy makes its way to the target through the atmosphere, it interacts with the target depending on the properties of both the target and the radiation.
- Step 4-Recording of Energy by the Sensor (D): after the energy has been scattered by, or emitted from the target, we require a sensor (remote not in contact with the target) to collect and record the electromagnetic radiation.
- Step 5-Transmission, Reception, and Processing (E): the energy recorded by the sensor has to be transmitted, often in electronic form, to a receiving and processing station where the data are processed into an image (hardcopy and/or digital).
- Step 6-Interpretation and Analysis (F) the processed image is interpreted, visually and/or digitally or electronically, to extract information about the target which was illuminated.
- **Step 7-Application (G)** the final element of the remote sensing process is achieved when we apply the information we have been able to extract from the imagery about the target in order to better understand it, reveal some new information, or assist in solving a particular problem.

1.2 Electromagnetic Spectrum

Solar radiation can be divided into different wavelengths call Electromagnetic spectrum as depicted by the figure below.

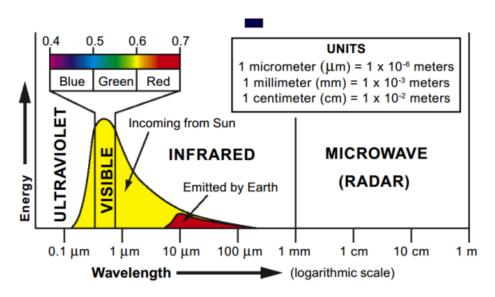


Figure 1.2: Electromagnetic spectrum

Most Multispectral satellite can acquire data in visible and infrared region (including the thermal infrared region). Nowdays there is an increasing demand for data in microwaves region, because the microwave energy can penetrate through the cloud.

1.3 Source of radiation

We have two major source of radiation: the sun and the earth.

The **sun** is the mayor source of radiation with highest energy in visible, near infrared and short-wave infrared region.

The earth surface can also emit energy usually in thermal Infrared and microwaves region.

1.4 Interaction in the atmosphere

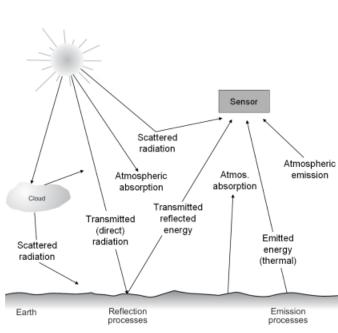


Figure 1.3: radiance at the sensor

Before radiation used for remote sensing reaches the Earth's surface it has to travel through some distance of the Earth's atmosphere. Different processes occur due to the presence of particles and gases in the atmo-Some get scattered, some get absorbed, some managed to transmit through the atmosphere and get absorbed by the earth, and some get **reflected**. This reflected energy is captured by the sensor on board a satellite. The earth equally emits energy, the amount of energy emitted is directly proportional to the heat level of the part of the earth that emit This emitted energy can also be recorded by sensors on board a satellite if that sensor has capabilities to record energy in thermal infrared region. A sensor on board a satellite can record **re-mitted** or **reflected** radiation depending on the capabilities of the sensors.

CASE STUDY: GOES-R Series

2.1 GOES project current status

The Geostationary Operational Environmental Satellite Program (GOES) is a joint effort of NASA and the National Oceanic and Atmospheric Administration (NOAA). The GOES system currently consists of GOES-13, operating as GOES-East, in the eastern part of the constellation at 75 degrees west longitude and GOES-15, operating as GOES-West, at 135 degrees west longitude. The GOES-R series will maintain the two-satellite system implemented by the current GOES series. However, the locations of the operational GOES-R satellites will be 75 degrees west longitude and 137 degrees west longitude. The latter is a shift in order to eliminate conflicts with other satellite systems. The GOES-R series operational lifetime extends through December 2036. These spacecraft help meteorologists observe and predict local weather events, including thunderstorms, tornadoes, fog, hurricanes, flash floods and other severe weather. In addition, GOES observations have proven helpful in monitoring dust storms, volcanic eruptions and forest fires. The benefits that directly enhance the quality of human life and protection of Earth's environment include:

- Supporting the search-and-rescue satellite aided system (SARSAT).
- Contributing to the development of worldwide environmental warning services and enhancements of basic environmental services.
- Improving the capability for forecasting and providing real-time warning of solar disturbances.
- Providing data that may be used to extend knowledge and understanding of the atmosphere and its processes.

The next series of GOES satellites includes GOES-R, S, T and U.

2.2 GOES-R Series Instruments

GOES-R Series has six instruments on board:

Earth-pointing:

- Advanced Baseline Imager (ABI) the primary instrument for imaging Earth's weather, oceans and environment.
- Geostationary Lightning Mapper (GLM) a single-channel, near-infrared optical transient detector that can identify momentary changes in an optical scene, indicating the presence of lightning.

Sun-pointing:

- Extreme Ultraviolet and X-ray Irradiance Sensors (EXIS) monitors solar irradiance in the upper atmosphere using two primary sensors: the Extreme Ultraviolet Sensor (EUVS) and the X-Ray Sensor (XRS).
- Solar Ultraviolet Imager (SUVI) a telescope that monitors the sun in the extreme ultraviolet wavelength range, detecting solar flares and solar eruptions, and compiling full disk solar images.

In-situ:

- Magnetometer (MAG) measures the space environment magnetic field that controls charged particle dynamics in the outer region of the magnetosphere
- Space Environment In-Situ Suite (SEISS) monitors proton, electron, and heavy ion fluxes in the magnetosphere using four sensors: the Energetic Heavy Ion Sensor (EHIS), the High and Low Magnetospheric Particle Sensors (MPS-HI and MPS-LO), and the Solar and Galactic Proton Sensor (SGPS).

Only two sensors are used for atmospheric purpose, Advanced Baseline Imager (ABI) and Geostationary Lightning Mapper (GLM).

2.3 Advanced Baseline Imager (ABI)

ABI is a multi-channel passive imaging radiometer that images Earth's weather, oceans and environment with 16 spectral bands (2 visible, 4 near-infrared, and 10 infrared channels).

| ABI Band | Central Wavelength(µm) | Туре | Nickname | Best Spatial Resolution |
|----------|------------------------|---------------|--------------------------|-------------------------|
| 1 | 0.47 | Visible | Blue | 1 |
| 2 | 0.64 | Visible | Red | 0.5 |
| 3 | 0.86 | Near-Infrared | Veggie | 1 |
| 4 | 1.37 | Near-Infrared | Cirrus | 2 |
| 5 | 1.6 | Near-Infrared | Snow/Ice | 1 |
| 6 | 2.2 | Near-Infrared | Cloud particle size | 2 |
| 7 | 3.9 | Infrared | Shortwave window | 2 |
| 8 | 6.2 | Infrared | Upper-level water vapor | 2 |
| 9 | 6.9 | Infrared | Midlevel water vapor | 2 |
| 10 | 7.3 | Infrared | Lower-level water vapor | 2 |
| 11 | 8.4 | Infrared | Cloud-top phase | 2 |
| 12 | 9.6 | Infrared | Ozone | 2 |
| 13 | 10.3 | Infrared | "Clean" longwave window | 2 |
| 14 | 11.2 | Infrared | Longwave window | 2 |
| 15 | 12.3 | Infrared | "Dirty" longwave window | 2 |
| 16 | 13.3 | Infrared | CO ₂ longwave | 2 |

Schmit, T. J., S. S. Lindstrom, J. J. Gerth, M. M. Gunshor, 2018. Applications of the 16 spectral bands on the Advanced Baseline Imager (ABI)

J. Operational Meteor., 6 (4), 33-46.

Figure 2.1: ABI optical components

2.4 ABI modes of operation

ABI has four modes of operation.

Full Disk: Hemispheric Coverage of 83° local zenith angle, temporal resolution of 5-15 minutes, and spatial resolution of 0.5 to 2km.

Mesoscale: Provides coverage over a 1000x1000km box with a temporal resolution of 30 seconds, and spatial resolution of 0.5 to 2km.

Continental US/Pacific US: The CONUS and PACUS scans are performed every five minutes, providing coverage of the 5000km (east/west) and 3000km (north/south) rectangle over the continental United States (GOES-16) or the Pacific Ocean, including Hawaii (GOES-17). The spatial resolution is 0.5 to 2km.

Flex Modes: The flex modes provide a full disk scan every 10 minutes (mode 6) or every 15 minutes (mode 3), a CONUS/PACUS every five minutes, and two mesoscale domains every 60 seconds (or one sub-region every 30 seconds).

2.5 GOES-R Data Products

GOES-R Data products have three levels.

Level 0 (L0): are observation data received directly from the 6 satellite instruments. The data is not meaningful to most users prior to processing by the ground system.

Level 1b (L1b): are calibrated and, where applicable, geographically corrected, L0 data. This means that the data has been processed so that its values are in standard units of physical quantities. For ABI, the L1b product is Radiances. This is useful for users who require radiance units, instead of reflectance/brightness temperature (Kelvin) units. All of the instruments have L1b products available except GLM, which is only distributed as an L2+ product.

Level 2+ Products: contain environmental physical qualities, such as cloud top height or land surface temperature. Aside from the GLM Lightning Detection Product, the data source for these products is the ABI L1b data.

2.6 ABI meteorological product

The mission-critical ABI product is Cloud and Moisture Imagery (CMI), which utilizes all 16 ABI spectral bands, and is used to generate an array of products aiding forecasters in monitoring and predicting weather hazards. The following is a list ABI meteorological and space weather data products that are available to the user community.

- Aerosol detection (including smoke and dust)
- Aerosol optical depth (AOD)
- Aerosol particle size
- Clear sky masks
- Cloud layers/heights
- Cloud and moisture imagery
- Cloud optical depth
- Cloud particle size distribution

- Cloud top height
- Cloud top phase
- Cloud top pressure
- Cloud top temperature
- Derived motion winds
- Derived stability indices
- Downward shortwave radiation: surface
- Fire/hot spot characterization
- Hurricane intensity estimation
- Land surface albedo
- Land surface bidirectional reflectance factor
- Land surface temperature (skin)
- Legacy vertical moisture profile
- Legacy vertical temperature profile
- Radiances
- Rainfall rate/QPE
- Reflected shortwave radiation: TOA
- Sea and lake ice: age
- Sea and lake ice: concentration
- Sea and lake ice: motion
- Sea surface temperature (skin)
- Snow cover
- Total precipitable water
- Volcanic ash: detection and height

GOES-R SERIES DATA COLLECTION AND VISUALISATION

3.1 Acces GOES-R series data files

This section describes few approaches to access, download and visualise data from GOES-R series satellites.

3.1.1 Access data files from NOAA CLASS

- The NOAA Comprehensive Large Array-data Stewardship System (CLASS) repository is the official site for accessing all available GOES-R Series Products.
- A similar but easier website to navigate is NCEI's Archive Archive Information Request System (AIRS).

3.1.2 Access data files from Amazon, Microsoft, OCC

- Amazon Web Service (AWS) ABI L1b and L2+, GLM L2+, and SUVI L1b products are available in AWS S3 Buckets. These open datasets can be accessed by the public from AWS for free.
- Microsoft Azure Two GOES-16 ABI Full Disk products are stored in a Azure blob container.
 The products currently available are L1b Radiances and L2+ MCMI, which can be accessed from Azure for free.
- Open Commons Consortium (OCC) Stores a 100 TB rolling archive of GOES-16 data (~8 months), the products stored are ABI L1b and ABI L2+ CMI and MCMI. OCC recommends using the AWS CLI or the python boto library, to access the data.

3.1.3 Access data files from Google Cloud

- Google Cloud ABI L1b and L2+, GLM L2+, and SUVI L1b products are available in two different buckets for GOES-16 and GOES-17.
- Google Earth Engine (GEE) A cloud-based platform for geospatial analysis. This service runs through Google Cloud, and pulls GOES-R datasets from the Google Cloud buckets.

3.1.4 Use Python to retrieve data from AWS

Python is the most powerful programming language using in data science and geospatial data analysis. The following sample scripts demonstrate how to retrieve GOES-R files from AWS using Python.

```
1 from goespy.Downloader import ABI_Downloader
2 ### to use ABI_Downloader, you need 7 arguments:
3 import datetime as dt
4 ### Getting the current date (in UTC coordinate)
5 utcDateTime = dt.datetime.utcnow()
6 ## current year
7 year = utcDateTime.strftime("%Y")
8 # current month
9 month = utcDateTime.strftime("%m")
10 ## current day
11 day = utcDateTime.strftime("%d")
12 ## current hour in UTC
hour = utcDateTime.strftime("%H")
14 ##Choose a channel from your preference (can be C01-C16)
15 channel = ["CO1"]
16 ## In GOES satellite they have 9 products
## 3 are L1b-Rad(M,C,F)
18 ## 3 are L2-CMIP(M,C,F)
## 3 are L2-MCMIP(M,C,F)
20 ### In your case we will get the CMIPF, F means FullDisk (all the projection by
      the satellite)
product = 'L1b-L2-CMIPF'
22 ## The Bucket is the variable contains the name of dataset server from goes on
     AWS
Bucket = 'noaa-goes16' ## in the future on AWS they will have goes17.
24 ## Now we will call the function ABI_Downloader:
Abi = ABI_Downloader('noaa-goes16', year, month, day, hour, product, channel)
```

After all the dataset is downloaded, they are in your home directory with that structure:

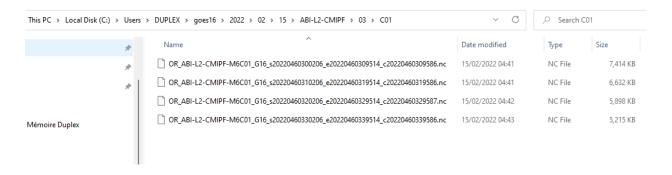


Figure 3.1: L1b-L2-CMIPF product downloaded

3.2 GOES-R Series metadata discovery

3.2.1 GOES-R Series data format

GOES-R Series product files use the netCDF-4 format. Network Common Data Form (NetCDF) is a set of software libraries and machine-independent data formats that support the creation, access, and sharing of array-oriented scientific data. It is also a community standard for sharing scientific data.

3.2.2 Read GOES-R Series data metadata

Metadata provides information about the distinct items, such as: means of creation, purpose of the data, time and date of creation, creator or author of data, placement on a network (electronic form) where the data was created, what standards used etc. The main purpose of metadata is to facilitate in the discovery of relevant information, more often classified as resource discovery. Metadata also helps organize electronic resources, provide digital identification, and helps support archiving and preservation of the resource. Before any manipulation of data, it is vitaly important to read metadata. Now let us see how to read the metadata of this satellite data file in command line using gdal tools.

```
Cespatial) DIVIL IS DATA ANALYSIS spalinfo OR ABI-11b-RadWI-MSCO2_Gids_20071931811268_c20171931811326_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c20171931811366_c2017193
```

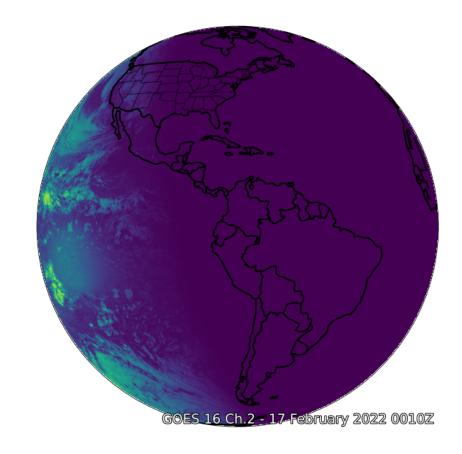
Figure 3.2: GOES-R Series metadata discovery

3.3 Visualise a single band ABI Channel image

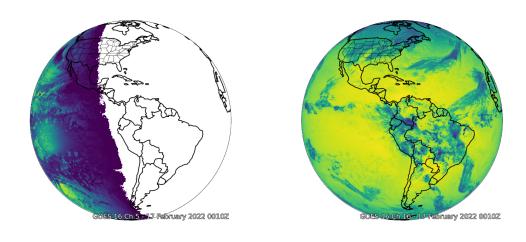
ABI imager has 16 channels, the code below show how to plot a single band ABI channel image.

```
def open_dataset(date, channel, idx, region):
      Open and return a netCDF Dataset object for a given date, channel, and
     image index
      of GOES-16 data from THREDDS test server.
      cat = TDSCatalog('https://thredds.ucar.edu/thredds/catalog/satellite/goes/
     east/products/'
                        f'CloudAndMoistureImagery/{region}/Channel{channel:02d}/{
     date: %Y%m%d}/catalog.xml')
      ds = cat.datasets[idx]
      ds = ds.remote_access(use_xarray=True)
9
      return ds
  def plot_GOES16_channel(date, idx, channel, region):
11
12
      Get and plot a GOES 16 data band from the ABI.
13
14
      ds = open_dataset(date, channel, idx, region)
      dat = ds.metpy.parse_cf('Sectorized_CMI')
16
      proj = dat.metpy.cartopy_crs
17
      x = dat['x']
```

```
y = dat['y']
19
      fig = plt.figure(figsize=(10, 10))
20
      ax = fig.add_subplot(1, 1, 1, projection=proj)
21
      ax.add_feature(cfeature.COASTLINE, linewidth=2)
22
      ax.add_feature(cfeature.STATES, linestyle=':', edgecolor='black')
      ax.add_feature(cfeature.BORDERS, linewidth=2, edgecolor='black')
24
      for im in ax.images:
          im.remove()
      im = ax.imshow(dat, extent=(x.min(), x.max(), y.min(), y.max()), origin='
     upper')
      timestamp = datetime.strptime(ds.start_date_time, '%Y%j%H%M%S')
28
      add_timestamp(ax, time=timestamp, high_contrast=True,
29
                     pretext=f'GOES 16 Ch.{channel} - ',
                     time_format='%d %B %Y %H%MZ', y=0.01,
31
                     fontsize=18)
32
      display(fig)
33
      plt.savefig("goes_fulldisk_C1.png")
34
      plt.close()
  channel_list = {u'1 - Blue Band 0.47 \u03BCm': 1,
                   u'2 - Red Band 0.64 \u03BCm': 2,
37
                   u'3 - Veggie Band 0.86 \u03BCm': 3,
                   u'4 - Cirrus Band 1.37 \u03BCm': 4,
39
                   u'5 - Snow/Ice Band 1.6 \u03BCm': 5,
40
                   u'6 - Cloud Particle Size Band 2.2 \u03BCm': 6,
                  u'7 - Shortwave Window Band 3.9 \u03BCm': 7,
42
                  u'8 - Upper-Level Tropo. WV Band 6.2 \u03BCm': 8,
43
                  u'9 - Mid-Level Tropo. WV Band 6.9 \u03BCm': 9,
44
                   u'10 - Low-Level WV Band 7.3 \u03BCm': 10,
45
                   u'11 - Cloud-Top Phase Band 8.4 \u03BCm': 11,
                  u'12 - Ozone Band 9.6 \u03BCm': 12,
47
                  u'13 - Clean IR Longwave Band 10.3 \u03BCm': 13,
                  u'14 - IR Longwave Band 11.2 \u03BCm': 14,
                   u'15 - Dirty Longwave Band 12.3 \u03BCm': 15,
                   u'16 - CO2 Longwave IR 13.3 \u03BCm': 16}
  region = Select(
52
      options = ['Mesoscale - 1', 'Mesoscale - 2', 'CONUS', 'PuertoRico', 'FullDisk'],
53
      description='Region:',)
  channel = Dropdown(options=channel_list,value=16,description='Channel:',)
57
58 interact(plot_GOES16_channel, date=fixed(date), idx=fixed(-2),
           channel=channel, region=region)
```



(a) ABI channel 1 image full disk



(b) ABI channel 5 image full disk

(c) ABI channel 16 image full disk

Figure 3.3: ABI images full disk

SATELLITE DATA ANALYSIS

The main objective of this analysis is to process satellite data and compare with data obtained from aviation meteorolgical stations. Here we will use the three hour data from GLDAS.

4.1 Aviation meteorological stations in cameroon

Table 4.1: Aviation meteorological stations



Figure 4.1: aviation meteorological stations

| NAME | OMM CODE | ICAO CODE | LATITUDE | LONGITUDE |
|-----------|----------|-----------|-----------|------------|
| Yaounde | 64950 | FKYS | 03°50'N | 011°31E |
| Douala | 64910 | FKKD | 04°00'N | 009°44'E |
| Garoua | 64860 | FKKR | 09°20N | 013°23'E |
| Maroua | 64851 | FKKL | 10°27'N | 14°15'E |
| Bafoussam | 64894 | FKKU | 05°32'05N | 010°21'15E |
| Bertoua | 64930 | FKKO | 04°36'N | 013°44'E |
| Bamenda | 64892 | FKKV | 06°03'N | 010°07'E |

4.2 GLDAS data

The goal of the Global Land Data Assimilation System (GLDAS) is to ingest satellite- and ground-based observational data products, using advanced land surface modeling and data assimilation techniques, in order to generate optimal fields of land surface states and fluxes (Rodell et al., 2004a). GLDAS data are available to download from this link. Detailed documentation about GLDAS 2.1 product is available here

4.3 Specific requirements

The requirements for the metereorological data at aviation station are:

| PARAMETER | SYMBOLS | UNIT |
|-------------------|---------|----------------------|
| Wind Speed | Ws | m/s |
| Air Temperature | Tair | $^{\circ}\mathrm{C}$ |
| Pressure | P | hpa |
| Relative Humidity | Rh | % |

Table 4.2: meteorological data required

The units at which GLDAS provide air temperature (K) and Pressure(Pa). Further GLDAS provide specific humidity. To match requirements the following conversion steps have to be performed:

- **Step 1**: Convert GLDAS air temperature from Kelvin to Deg C.
- **Step 2**: Convert the unit of GLDAS pressure from Pa to Milli bar (Mb).
- **Step 3**: Convert specific humidity to relative humidity following the description here.
- **Step 4**: Extract the Ws, Tair, P and Rh corresponding at our aviation meteorological stations at observation time: 00:00, 03:00, 06:00, 09:00, 12:00, 15:00, 18:00 and 21:00 UTC.
- **Step 5**: Compare the results with data collected at the stations.

4.4 Download GLDAS data

GLDAS documentation project provide python script to download data. We run the script below to download data from 01 to 31 January 2021.

```
## In the Ubuntu session.

## Type in the following command to install the python package "gldas"

pip3 install gldas
```

```
## press enter
## Below code will download the GLDAS data between the dates provided in the command.

gldas_download -s 2021-01-01 -e 2021-01-31 --product GLDAS_Noah_v21_025 -- username nina.younkap --password ***** /mnt/path/to/folder

## Note that the last argument in above command is path to a folder where the downloaded files will be stored.
```

4.5 Processing single data file

4.5.1 Data file structure

Let us now see how to process a single NetCDF file (.nc4) dowloaded from GLDAS using the previous step and perform the required conversions. For example let us consider the GLDAS data representing 06:00 hours on 31 January 2021. The GLDAS file name follows a particular structure. The file name is GLDAS_NOAH025_3H.A20210131.0600.021.nc4

- GLDAS_NOAH025_3H represents the name, spatial resolution (025 means 0.25 degree resolution) and temporal resolution (3H means three hour data)
- A20210131 represents the date of acquisition
- 0600 represents the time (in this case 06:00 GMT) the parameters are computed for.
- **021** represents the version of GLDAS data, in this case 2.1.
- nc4 represents the extension and format of the data, in this case netCDF4.

4.5.2 Metadata discovery

Now let us see how to read the metadata of this file in command line using gdal tools

```
## change directory to where you have downloaded the GLDAS data using the below
cd /mnt/d/mi_is_project_data/2021/31"
## Extract metadata of the 'GLDAS_NOAH025_3H.A20210131.0600.021.nc4' in the
file "metadata.txt"

gdalinfo GLDAS_NOAH025_3H.A20210131.0600.021.nc4 > metadata.txt
## press enter
## Read metadata using vim editor
vim metadata.txt
```

In the metadata, under Subdatasets: all the parameters provided as subdatasets are listed. These subdataset names are used in the further steps to process individual parameters. For example, netCDF: "GLDAS_NOAH025_3H.A20210131.0600.021.nc4": Tair_f_inst is the name of air temperature grid at 06:00 on 31 January 2021 and we will use this name to extract/process this single grid. The codes of the parameters are listed in the GLDAS user manual (Table 3.1).

4.5.3 Single data extraction

```
So for our project we need the following subdatasets from a single GLDAS file:

NETCDF: "GLDAS_NOAH025_3H.A20210131.0600.021.nc4": Qair_f_inst - Specific humidity (Kg/Kg)

NETCDF: "GLDAS_NOAH025_3H.A20210131.0600.021.nc4": Psurf_f_inst - Surface Pressure (Pa)

NETCDF: "GLDAS_NOAH025_3H.A20210131.0600.021.nc4": Tair_f_inst - Air Temperature (K)

NETCDF: "GLDAS_NOAH025_3H.A20210131.0600.021.nc4": Wind_f_inst - Wind speed (m/s)

NETCDF: "GLDAS_NOAH025_3H.A20210131.0600.021.nc4": SWdown_f_inst - Downward short-wave radiation flux (W m-2)
```

Now let us extract each of the above subdatasets and convert them into tif file. For this we will use a gdal command called gdal_translate which is a global conversion tool for all kind of raster formats. Detailed documentation on gdal_translate is given here.

```
1 ## Convert specific humidity to tif - below command
2 gdal_translate NETCDF: "GLDAS_NOAHO25_3H.A20210131.0600.021.nc4":Qair_f_inst
     GLDAS_NOAH025_3H_A20210131_0600_Qair.tif
3 ## Convert Surface Pressure to tif - below command
4 gdal_translate NETCDF: "GLDAS_NOAH025_3H.A20210131.0600.021.nc4":Psurf_f_inst
     GLDAS_NOAH025_3H_A20210131_0600_Psurf.tif
5 ## Convert air temperature to tif - below command
6 gdal_translate NETCDF: "GLDAS_NOAHO25_3H.A20210131.0600.021.nc4": Tair_f_inst
     GLDAS_NOAH025_3H_A20210131_0600_Tair.tif
7 ## Convert Wind speed to tif - below command
8 gdal_translate NETCDF: "GLDAS_NOAH025_3H.A20210131.0600.021.nc4": Wind_f_inst
     GLDAS_NOAH025_3H_A20210131_0600_Wind.tif
9 ## Convert Short wave downward radiation to tif - below command
10 gdal_translate NETCDF: GLDAS_NOAH025_3H.A20210131.0600.021.nc4": SWdown_f_tavg
     GLDAS_NOAH025_3H_A20210131_0600_SWdown.tif
## Press enter
```

4.5.4 Data visualisation in QGIS

Now that we have the required parameters of 31 January 2021 in .tif format, let us open the air temperature map in QGIS and see how it looks!

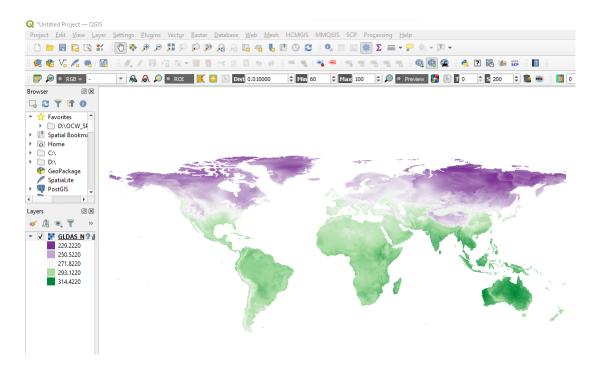


Figure 4.2: GLDAS Tair data displayed in QGIS

Let us also zoom to Cameroon boundaries and have a look at the Tair map. Remember the resolution of data is 0.25 degrees.

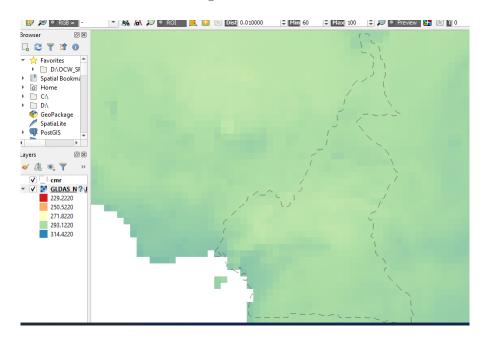


Figure 4.3: GLDAS Tair data displayed in QGIS and zoomed to Cameroon

4.5.5 Unit data conversion

Now let us see how to do unit conversion of all the four parameters required for aviation. One additional step is to clip the unit converted maps to cameroon boundaries as we are only interested in that region not global.

For further steps let us move to **GRASS GIS** as it is easier to do spatial and temporal analysis with GRASS library. We will create a new location and mapset for processing GLDAS data for Cameroon boundaries.

Import the cameroon vector file, aviation meteorological station file, and others file required for our project into grass location.

```
## IMPORT VECTOR DATA: Boundaries of Cameroon an Aviation meteorological
     station in cameroon
2 ## Navigate (cd) to the folder Cmr_Base_Layers
3 cd /mnt/d/mi_is_project_data/Cmr_Base_Layers
4 # Import 'Cameroon' boundary shapefile into a vector in Grass GIS
5 v.import in=cmr.shp out=cmr
6 # Import 'Aviation meteorological staitions' points shapefile into a vector in
     Grass GIS
v.import in=aero_met_station_cmr.shp out=met_station
8 # Import 'FKKR TMA2' boundary shapefile into a vector in Grass GIS
9 v.import in=Fkkr_tma2.shp out=fkkr_tma2
10 # Import 'FKKN TMA' boundary shapefile into a vector in Grass GIS
v.import in=Fkkn_tma.shp out=fkkn_tma
12 # Import 'FKKL CONTROL ZONE' boundary shapefile into a vector in Grass GIS
v.import in=Fkkl_ctr.shp out=fkkl_ctr
14 # set the computational region to Cameroon Boundaries and set the computational
      resolution to 0.25 degrees
15 g.region vector=cmr res=0.25 -a
```

To automate workflows, the following bash script process all the four required parameters (Tair, Rh, Wind speed and Psurf), clipped on cameroon boundaries and converted to the required units from .nc4 files in a day.

```
#!/bin/bash
 2 ## This script process a single day GLDAS data and do all the required
           conversions needed for our project.
 3 ## GENERAL ##
 4 if [ -z "$GISBASE" ] ; then
             echo "You must be in GRASS GIS to run this program." >&2
 7 fi
 8 # Set a environment to enable overwrite by default
 9 export GRASS_OVERWRITE=1
11 # Navigate to the folder containing single day .nc4 files
12 #e.g INDAT="/mnt/d/mi_is_project_data/Cmr_Base_Layers/2021/01" for 01 January
           2021
13 #Here we work on the 31 January 2021 data
14 INDAT="/mnt/d/mi_is_project_data/2021/031"
15 cd ${INDAT}
16 # set the computational region to Urmia Lake basin and set the computational
           resolution to 0.25 degrees
17 g.region vector=cmr res=0.25 -a
18 # set the mask to cameroon boundaries
20 # For loop to process all the .nc files in one go
for i in `ls GLDAS*.nc4`; do
             dt = echo ${i}|cut -d. -f2-3
             # Convert specific humidity to tif - below command
23
             gdal_translate NETCDF:"${i}":Qair_f_inst GLDAS_NOAH025_3H_${dt}_Qair.tif
24
             # Convert Surface Pressure to tif - below command
             gdal_translate NETCDF: "${i}": Psurf_f_inst GLDAS_NOAH025_3H_${dt}_Psurf.tif
26
             # Convert air temperature to tif - below command
             \verb|gdal_translate| \verb|NETCDF:|| \$ \{i\} || : \verb|Tair_f_inst| | GLDAS_NOAH025_3H_$ \{dt\}_Tair.tif| | GLDAS_NOAH025_3H_$ (dt)_Tair.tif| | GLDAS_
28
             # Convert Wind speed to tif - below command
29
             gdal_translate NETCDF:"${i}":Wind_f_inst GLDAS_NOAH025_3H_${dt}_Wind.tif
             # Convert Short wave downward radiation to tif - below command
31
             gdal_translate NETCDF: "${i}": SWdown_f_tavg GLDAS_NOAH025_3H_${dt}_SWdown.
32
           tif
             # Import to GRASS
33
             # Import and clip specific humidity
             r.import in=GLDAS_NOAH025_3H_${dt}_Qair.tif out=GLDAS_NOAH025_3H_${dt}_Qair
35
             # Import and clip Surface Pressure
36
             r.import in=GLDAS_NOAHO25_3H_${dt}_Psurf.tif out=GLDAS_NOAHO25_3H_${dt}
37
            _Psurf -o
             # Import and clip Tair
```

```
r.import in=GLDAS_NOAH025_3H_${dt}_Tair.tif out=GLDAS_NOAH025_3H_${dt}_Tair
      -0
      # Import and clip Wind speed
40
      r.import in=GLDAS_NOAH025_3H_${dt}_Wind.tif out=GLDAS_NOAH025_3H_${dt}_Wind
41
      # Import and clip Short wave radiation
42
      r.import in=GLDAS_NOAHO25_3H_${dt}_SWdown.tif out=GLDAS_NOAHO25_3H_${dt}
43
     SWdown -o
      # Unit conversion
44
      # Air temperature from Kelvin to degree celsius
45
      r.mapcalc "GLDAS_NOAHO25_3H_${dt}_Tair_final = GLDAS_NOAHO25_3H_${dt}_Tair
46
     - 273.15"
      # Short wave radiation (no conversion required)
      r.mapcalc "GLDAS_NOAH025_3H_${dt}_SWdown_final = GLDAS_NOAH025_3H_${dt}
48
     _SWdown"
      # Wind speed (no conversion required)
49
      r.mapcalc "GLDAS_NOAH025_3H_${dt}_Wind_final = GLDAS_NOAH025_3H_${dt}_Wind"
50
      ## Specific humidity re saved
51
      r.mapcalc "GLDAS_NOAH025_3H_${dt}_Qair_final = GLDAS_NOAH025_3H_${dt}_Qair"
      ## Pressure convert from pa to mb
53
      r.mapcalc "GLDAS_NOAHO25_3H_${dt}_Psurf_final = GLDAS_NOAHO25_3H_${dt}
     Psurf / 100"
      ## Humidity according to the url: https://earthscience.stackexchange.com/
     questions/2360/how-do-i-convert-specific-humidity-to-relative-humidity
      r.mapcalc "es = 6.112 * exp((17.67 * GLDAS_NOAH025_3H_${dt}_Tair_final) / (
56
     GLDAS_NOAH025_3H_${dt}_Tair_final + 243.5))"
      r.mapcalc "e = (GLDAS_NOAH025_3H_${dt}_Qair_final * GLDAS_NOAH025_3H_${dt}
57
     _Psurf_final) / (0.378 * GLDAS_NOAH025_3H_${dt}_Qair_final + 0.622)"
      r.mapcalc "GLDAS_NOAH025_3H_{4t}_Rh = (e / es) * 100"
      # Final Relative humidity in %
59
      r.mapcalc "GLDAS_NOAH025_3H_${dt}_Rh_final = float(if(GLDAS_NOAH025_3H_${dt}
     }_Rh > 100, 100, if(GLDAS_NOAH025_3H_${dt}_Rh < 0, 0, GLDAS_NOAH025_3H_${dt}}
     _Rh)))"
62 done
```

We save the script above in the file myscript.sh. The following code show how to run this file on Linux Operating System.

```
# Run the following command to install dos2unix
sudo apt-get install dos2unix
# Below command removes the trailing spaces from your script
dos2unix myscript.sh
# Run the following command to run the above saved script file
sh myscript.sh
# press enter
```

RESULTS

This chapter deals about the outcome of our analysis and how to share the results with the community.

5.1 Extract chart command line tools.

5.1.1 Visualise temperature chart.

The script below show how to extract temperature chart from GRASS GISS using command line. The same code work for others meteorological parameters (Wind, Surface Pressure, Relative Humidity), we just need to change few parameters in the script.

```
1 # Visualize data
2 #Set date and time on 31 January 2021 at 03:00 UTC
3 dt="20210131.0300"
4 #Set date and time on 31 January 2021 at 03:00 UTC
5 #dt = "20210131.1500"
6 # Open a monitor
7 d.mon wx0
8 #set the coordinate grid
9 d.grid 3
10 # Display a raster map
d.rast GLDAS_NOAH025_3H_A${dt}_Tair_final
12 # Display a vector map
d.vect map=cmr type=boundary
14 #Add text
15 #d.text text="31 JAN 2021 0300Z" color=black bgcolor=white size=3
16 d.text text="31 JAN 2021 1500Z" color=black bgcolor=white size=3
17 # Add raster legend
18 # Add raster legend
19 d.legend -t -s -b raster=GLDAS_NOAHO25_3H_A${dt}_Tair_final title=TEMPERATURE
     title_fontsize=20 font=sans fontsize=18
20 # Add North arrow
21 d.northarrow style=1b text_color=black
22 #clear the monitor
23 #d.erase -f
24 # press enter
```

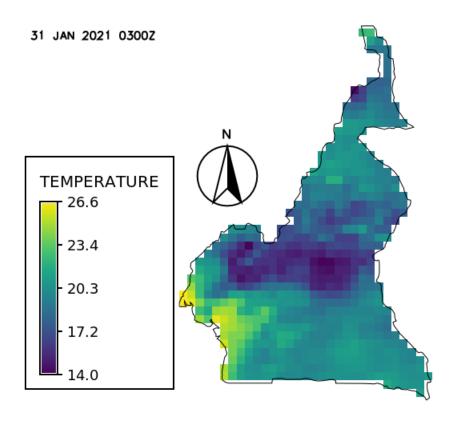


Figure 5.1: Temperature at 03:00 UTC

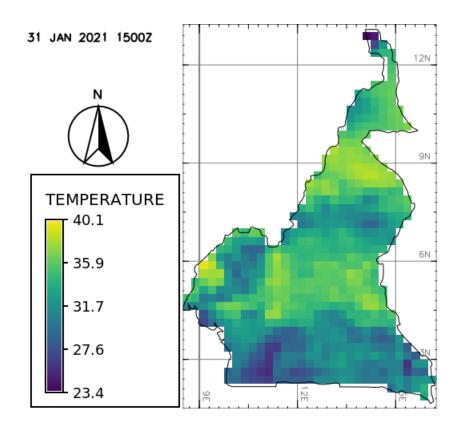


Figure 5.2: Temperature at 15:00 UTC

5.1.2 Visualise Relative Humidity Chart

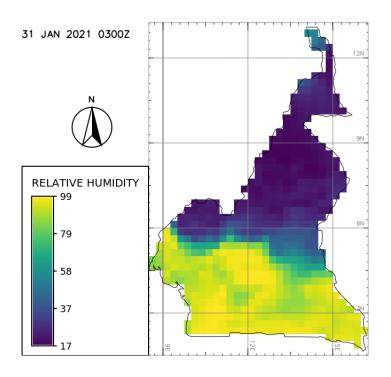


Figure 5.3: Relative Humidity at 03:00 UTC

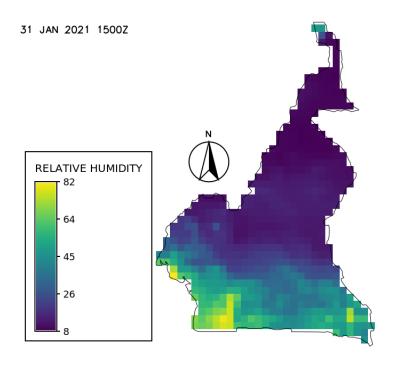


Figure 5.4: Relative Humidity at 15:00 UTC

5.1.3 Visualise Surface Pressure Chart

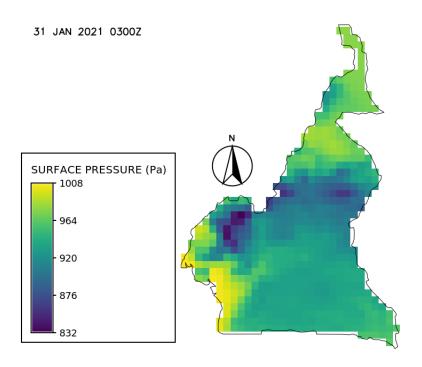


Figure 5.5: Surface Pressure at 03:00 UTC

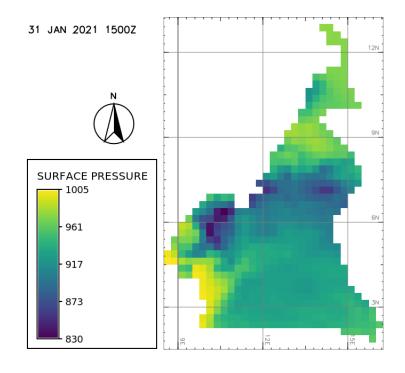


Figure 5.6: Surface Pressure at $15:00~\mathrm{UTC}$

5.1.4 Visualise Wind Speed Chart

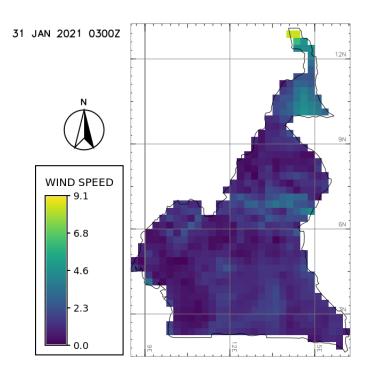


Figure 5.7: Wind Speed at 03:00 UTC

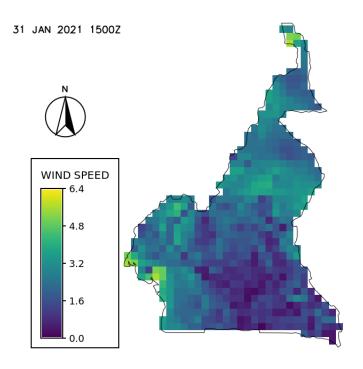


Figure 5.8: Wind Speed at 15:00 UTC

5.2 Extract data

Let us see how to extract raster values at particular observation points given as a vector layer. Here we will use aviation meteorological observation points in cameroon, to extract meterological parameters Tair, Ws, P and Rh at observation time: 00:00,03:00,06:00,09:00,12:00,15:00,18:00 and 21:00.

```
1 #!/bin/bash
_{2} ## This script process a single day GLDAS data and do all the required
     conversions needed for PySEBAL
3 ## GENERAL ##
4 if [ -z "$GISBASE" ] ; then
      echo "You must be in GRASS GIS to run this program." >&2
      exit 1
7 fi
8 # Set a environment to enable overwrite by default
9 export GRASS_OVERWRITE=1
10 # Navigate to the folder containing single day .nc4 files
11 INDAT="/mnt/d/mi_is_project_data/2021/031"
12 cd ${INDAT}
13 # set the computational region to Urmia Lake basin and set the computational
     resolution to 0.25 degrees
14 g.region vector=cmr res=0.25 -a
15 r.mask vect=cmr
17 # For loop to process all the .nc files in one go
  for i in `ls GLDAS*.nc4`; do
      dt = echo  ${i}|cut -d. -f2-3
      # Extract Air Temperature values 8 aviation meteorological station at time
     = dt and save the value in .csv file
     r.what -n map=GLDAS_NOAH025_3H_${dt}_Tair_final points=met_station
     comma output=a_temp_${dt}Z.csv
      # Extract Relative Humidity values 8 aviation meteorological station at
     time = dt and save the value in .csv file
     r.what -n map=GLDAS_NOAH025_3H_${dt}_Rh_final points=met_station sep=comma
24
      output=a_rh_${dt}Z.csv
      # Extract Surface Pressure values 8 aviation meteorological station at time
25
      = dt and save the value in .csv file
      r.what -n map=GLDAS_NOAH025_3H_${dt}_Psurf_final points=met_station sep=
     comma output=a_p_${dt}Z.csv
      # Extract Surface wind values 8 aviation meteorological station at time =
27
     dt and save the value in .csv file
      r.what -n map=GLDAS_NOAH025_3H_${dt}_Wind_final points=met_station
     comma output=a_ws_${dt}Z.csv
29 done
```

We save the script abive in the file $_{\rm extraction.sh}$, then run the bash script on Ubuntu using the command $_{\rm sh}$ extract.sh.

Let see how these file look in the directory

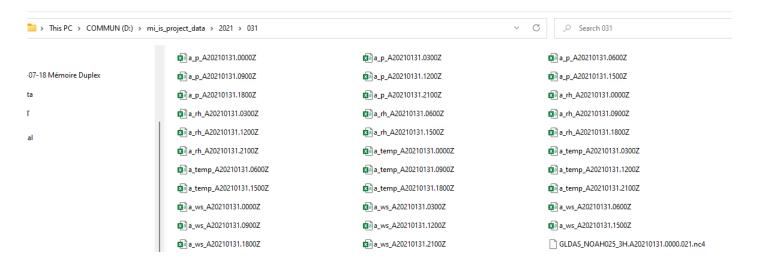


Figure 5.9: File in directory

Table 5.1: Station Temperature extracted

| Time | | | | | | | | |
|-------|----------|--------|---------|--------|------------|-----------|---------|---------|
| Time | Zaottide | Douala | Gaiolia | Maroua | ₹gaoundere | Palousand | Bertona | Samenda |
| 00:00 | 20.53 | NA | 22.69 | 21.88 | 17.82 | 16.16 | 19.95 | 16.23 |
| 03:00 | 20.89 | NA | 20.52 | 20.06 | 17.20 | 15.30 | 18.65 | 15.74 |
| 06:00 | 20.56 | NA | 20.26 | 19.83 | 17.15 | 15.12 | 18.81 | 15.93 |
| 09:00 | 28.19 | NA | 29.55 | 27.78 | 28.93 | 28.41 | 28.84 | 29.24 |
| 12:00 | 32.79 | NA | 35.92 | 32.96 | 32.38 | 31.97 | 37.50 | 33.15 |
| 15:00 | 28.80 | NA | 37.19 | 34.56 | 31.21 | 31.29 | 34.05 | 32.20 |
| 18:00 | 24.12 | NA | 29.08 | 25.97 | 22.35 | 21.54 | 25.18 | 20.65 |
| 21:00 | 22.51 | NA | 25.92 | 24.92 | 18.47 | 18.26 | 21.30 | 17.94 |

Table 5.2: Station Relative Humidity extracted

| Time | | | ve Humidit | y | | | | |
|-------|-----------------|---------|------------|--------|------------|------------|----------|---------|
| Time | Laoui de | Dollala | Gaiolia | Maroua | ₹gaoindere | Paloussall | Bertolia | Samenda |
| 00:00 | 94.54 | NA | 16.14 | 20.95 | 21.98 | 54.31 | 58.72 | 32.07 |
| 03:00 | 90.14 | NA | 18.46 | 23.37 | 20.91 | 40.82 | 72.48 | 29.25 |
| 06:00 | 90.39 | NA | 19.23 | 23.40 | 23.16 | 33.04 | 91.95 | 26.58 |
| 09:00 | 54.54 | NA | 11.91 | 13.46 | 13.44 | 14.43 | 50.12 | 13.34 |
| 12:00 | 33.52 | NA | 8.54 | 9.56 | 12.56 | 14.23 | 21.82 | 12.27 |
| 15:00 | 45.02 | NA | 8.18 | 9.07 | 13.46 | 15.93 | 23.29 | 13.37 |
| 18:00 | 60.96 | NA | 12.36 | 14.20 | 20.06 | 32.72 | 38.15 | 28.70 |
| 21:00 | 79.47 | NA | 14.99 | 14.70 | 25.07 | 54.81 | 54.48 | 34.51 |

Table 5.3: Station Surface Pressure extracted

| Time | | | | | | | | |
|-------|-----------------|---------|----------|----------|----------|-----------|----------|----------|
| Time | Laguride | Dottala | Garoua | Maigua | | Palousand | Bertotia | Banenda |
| 00:00 | 930.2783 | NA | 979.7773 | 962.4352 | 889.6512 | 871.1833 | 933.2983 | 873.8692 |
| 03:00 | 929.7397 | NA | 979.4847 | 961.8867 | 888.5327 | 870.0627 | 932.4217 | 872.4927 |
| 06:00 | 930.5676 | NA | 981.7227 | 963.8837 | 890.0356 | 871.6027 | 933.8506 | 873.8066 |
| 09:00 | 931.4058 | NA | 982.3198 | 964.6767 | 890.5017 | 872.1998 | 934.3908 | 874.4767 |
| 12:00 | 928.5176 | NA | 979.0986 | 962.0646 | 888.1417 | 869.7606 | 931.1577 | 871.7536 |
| 15:00 | 925.8957 | NA | 975.8347 | 959.1037 | 886.0237 | 867.2417 | 928.1847 | 869.2897 |
| 18:00 | 929.0854 | NA | 978.3014 | 961.4464 | 888.9214 | 870.0524 | 930.8405 | 872.8555 |
| 21:00 | 930.7338 | NA | 979.9328 | 962.7498 | 890.4208 | 871.8887 | 932.9238 | 874.6868 |

5.3 Comparison between values extracted from satellite and those obseved at met station

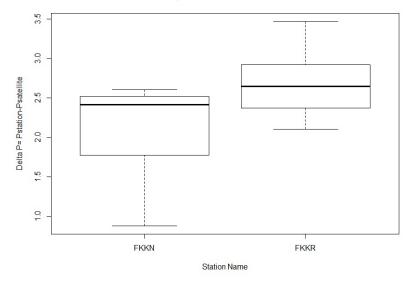
Table 5.4: Values extracted from satellite and observed at Garoua station

| Time | | Garoua meteorological station | | | | | | | | |
|-------|----------------|-------------------------------|---------------|------------|-------------|---------|--|--|--|--|
| | Tail satellite | Tait station | Ril satellite | Rh station | 2 satellite | Station | | | | |
| | Zair | Zarr | RI | Rit | Q Ž | Q / | | | | |
| 00:00 | 22.69231567 | 24 | 16.14377 | 31 | 979.7773 | 982.6 | | | | |
| 03:00 | 20.52959595 | 20.2 | 18.45571 | 39 | 979.4847 | 982.5 | | | | |
| 06:00 | 20.26201782 | 19.5 | 19.2371 | 43 | 981.7227 | 983.9 | | | | |
| 09:00 | 29.55059204 | 29 | 11.90664 | 25 | 982.3198 | 984.9 | | | | |
| 12:00 | 35.92055664 | 34 | 8.537706 | 15 | 979.0986 | 981.8 | | | | |
| 15:00 | 37.19307861 | 36 | 8.181383 | 17 | 975.8347 | 979.3 | | | | |
| 18:00 | 29.08526001 | 30.5 | 12.35657 | 27 | 978.3014 | 980.4 | | | | |
| 21:00 | 25.92067871 | 25.9 | 14.98635 | 31 | 979.9328 | 982.5 | | | | |

Table 5.5: Values extracted from satellite and observed at Ngaoundere station

| Time | | Ngaoundere meteorological station | | | | | | | |
|-------|----------------|-----------------------------------|--------------|------------|------------|---------|--|--|--|
| Time | Tail satellite | Tail station | Rh satellite | Rh station | o satelite | station | | | |
| | Tail | Tait | Str. | RIL | 2 % | 9 3 | | | |
| 00:00 | 17.82232056 | 13.8 | 21.97756 | 31 | 889.6512 | 892.1 | | | |
| 03:00 | 17.19960938 | 12.7 | 20.90649 | 30 | 888.5327 | 891.1 | | | |
| 06:00 | 17.15200195 | 11.5 | 23.15677 | 38 | 890.0356 | 892.0 | | | |
| 09:00 | 28.93059692 | 22.4 | 13.43818 | 19 | 890.5017 | 893.1 | | | |
| 12:00 | 32.38057861 | 26.4 | 12.56398 | 13 | 888.1417 | 890.6 | | | |
| 15:00 | 31.21306763 | 28.6 | 13.45516 | 12 | 886.0237 | 888.4 | | | |
| 18:00 | 22.35527954 | 20.8 | 20.06234 | 24 | 888.9214 | 889.8 | | | |
| 21:00 | 18.47069702 | 16.5 | 5.06882 | 27 | 890.4208 | 892.0 | | | |

Box an Whiskers Surface pressure difference at FKKR and FKKN 31/01/22



On the whole the satelitte over estimates the temperature, but at Garoua the difference is less than 1°C compared to Ngaoundere where we can have up to -6°C, that might be explained by the uneven landshape over Ngaoundere.

Box an Whiskers Surface pressure difference at FKKR and FKKN 31/01/22

Regardless to the time, the temperature is over estimated by the satellite, nevertheless the difference is high from 6 to 12 AM, when the sun radiation is highest.

Table 5.6: Temperature summary at FKKR

| Min. | 1st Qu. | Median | Mean | 3rd Qu. | 3rd Qu. |
|---------|---------|---------|---------|---------|---------|
| -1.9206 | -0.8698 | -0.4401 | -0.2568 | 0.3114 | 1.4147 |

Temperature difference at FKKR on 31/01/22 in °C

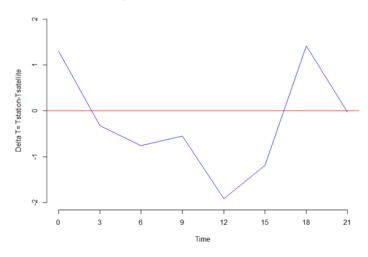
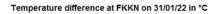
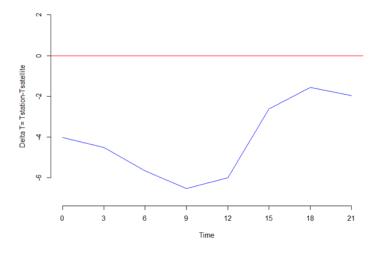


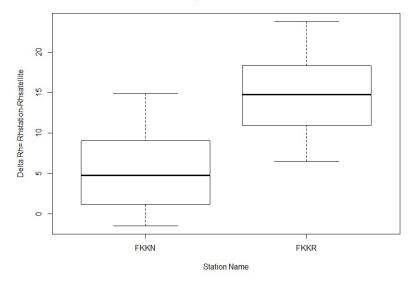
Table 5.7: Temperature summary at FKKN

| Min. | 1st Qu. | Median | Mean | 3rd Qu. | 3rd Qu. |
|--------|---------|--------|--------|---------|---------|
| -6.531 | -5.734 | -4.261 | -4.103 | -2.452 | -1.555 |



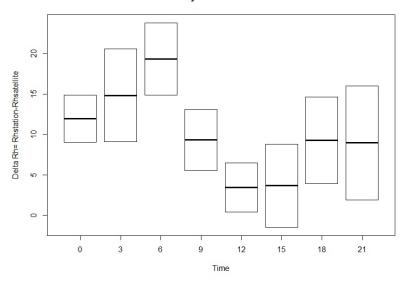


Box an Whiskers Rhumidity difference at FKKR and FKKN 31/01/22



On the whole the Relative humidity is under estimated by the satellite

Box an Whiskers Rhumidity difference at FKKR and FKKN 31/01/22



The difference of Relative humidity is more than 10% from 0 to 6 AM and less than 10% from 09:00 to 21:00 UTC

Table 5.8: Relative humidity summary at FKKR

| Min. | 1st Qu. | Median | Mean | 3rd Qu. | 3rd Qu. |
|-------|---------|--------|--------|---------|---------|
| 6.462 | 12.025 | 14.750 | 14.774 | 17.146 | 23.763 |

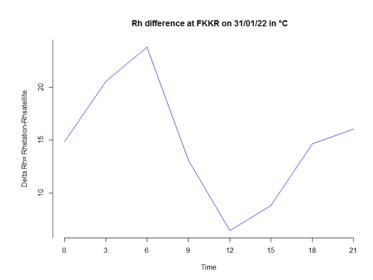
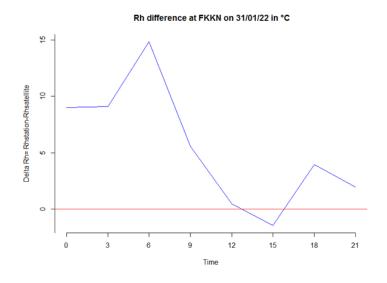
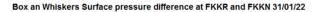
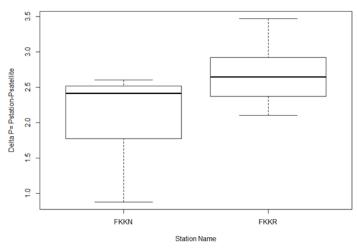


Table 5.9: Relative humidity summary at FKKN

| Min. | 1st Qu. | Median | Mean | 3rd Qu. | Qu. |
|--------|---------|--------|-------|---------|--------|
| -1.455 | 1.557 | 4.750 | 5.421 | 9.040 | 14.843 |







On the whole the surface pressure is under estimated by the satellite, however the gap is greater in Garoua than in Ngaoundere

Box an Whiskers Surface pressure difference at FKKR and FKKN 31/01/22

The pressure is under estimated, the difference of pressure has two minimum (at 6 and 18 AM) and two maximum (at 3 and 15 AM), this refers to the daily cycle of the pressure

Table 5.10: Surface pressure humidity summary at FKKR

| I | Min. | 1st Qu. | Median | Mean | 3rd Qu. | 3rd Qu. |
|---|-------|---------|--------|-------|---------|---------|
| 2 | 2.099 | 2.470 | 2.641 | 2.679 | 2.871 | 3.465 |

Surface pressure difference at FKKR on 31/01/22 in mb

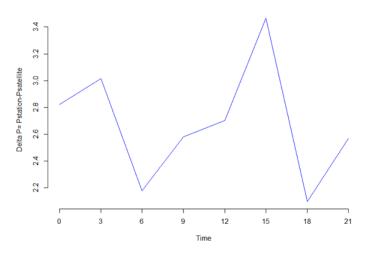
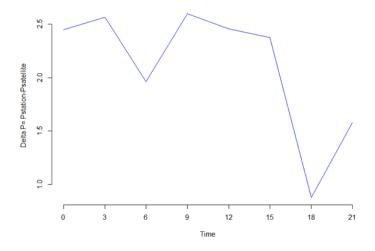


Table 5.11: Surface pressure summary at FKKN

| Min. | 1st Qu. | Median | Mean | 3rd Qu. | 3rd Qu. |
|---------|---------|--------|--------|---------|---------|
| 0.87865 | 1.8681 | 2.4125 | 2.1089 | 2.4855 | 2.5983 |





CONCLUSION AND FUTURE ENHANCEMENT

This project enable us to successfully perform spatio-temporal analysis of satellite data through the following operations:

- Extraction of single meteorological parameters (temperature, relative humidity, surface pressure and wind) required for avaition safety and efficiency;
- Conversion of temperature from Kelvin to Deg C and surface pressure from Pa to hpa;
- Conversion from specific humidity to relative humidity;
- Computation and visualisation on thematic chart of meteorological parameters to the extent of Cameroon boundaries;
- Extraction of meteorological parameters corresponding to our ground observation points;
- Comparison between data extracted from satellite remote sensing and those directly obtained in situ at ground observation points.

We found that on the whole the satellite **over estimates** temperature and **under estimates** pressure and relative humidity.

But due to the time allow for this project the comparison between satellite data and in situ data was limited to two stations (Ngaoundere and Garoua) and on a single day (31 January 2021).

These space and time limitations can not enable us to draws a global picture representative for the whole cameroon boundaries. So, as enhancement for this work, we are planing to:

- Extend our study space on the 19 meteorological stations (8 aviation and 11 agro) that compose the cameroon meteorological network;
- Extend our study time on the last 5, 10 and 30 years;
- Extend our study to others meteorological parameters required for agriculture (AETI, Net radiation, precipitation, NDVI, transpiration, PET, etc.).

Given the fact that the installation, operation and maintenance of a conventional or automatic weather station is expensive and unaffordable for many farmers. We aim to provide farmers with free weather data derived from publicly available satellites (Goes-R, Landsat, EumetSat, Modis etc.) to fufill their meteorological needs and support their business operations.

REFERENCES

Bhattacharyya Debnath, Ranjan Rahul, A. Farkhod, Alisherov, and Choi Minkyu. Biometric authentication: A review. *International*, page 16, sep 2009.