



**SCIKIT-data access**  
DATA INTERFACES FOR PYTHON

# Skdaccess: The **Scikit Data Access** Python Package

## Quick Start Guide

v1.9.17 for Python 3.6

<https://pypi.python.org/pypi/scikit-dataaccess>

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## 1 Overview

The Scikit Data Access package simplifies the handling of scientific data sets in Python. It provides a common interface across all data sets, based on a data fetcher and iterator pattern, as illustrated in the Figure below.



This paradigm places the requirements for parsing and interpreting the data inside of the data fetcher, which returns a data wrapper that provides a uniform method for accessing the data. In particular, the data wrapper implements an iterator which returns the next segment of data when requested by another function or by the user.

## Advantages of Scikit Data Access

- Import scientific data from various sources through one easy Python API.
- Use iterator patterns for each data source (configurable data generators + functions to get next data chunk).
- Skip parser programming and file format handling.
- Enjoy a common namespace for all data and unleash the power of data fusion.
- Handle data distribution in different modes: (1) local download, (2) caching of accessed data, or (3) online stream access.
- Easily pull data on cloud servers through Python scripts and facilitate large-scale parallel processing.

- Build on an extensible platform: Adding access to a new data source only requires addition of its “DataFetcher.py”.
- Open source (MIT License).

## 2 Supported Data Sets

The package introduces a common namespace and currently supports the following data sets:

Name-space	Data structure	Original Source	Data Size	Description
skdaccess. astro. kepler	Dictionary of Data Frames	Mikulski Archive for Space Telescopes ( <a href="ftp://archive.stsci.edu/pub/kepler/lightcurves/">ftp://archive.stsci.edu/pub/kepler/lightcurves/</a> )	≈ 1TB	Light curves for stars imaged by the <i>Kepler</i> Space Telescope ( <a href="https://keplerscience.arc.nasa.gov/">https://keplerscience.arc.nasa.gov/</a> ). This data set uses a cache data fetcher.
skdaccess. astro. spectra	Dictionary of Data Frames	Sloan Digital Sky Survey Science Archive Server ( <a href="https://data.sdss.org/sas/">https://data.sdss.org/sas/</a> )	100KB / image	Spectra from the Sloan Digital Sky Server ( <a href="https://www.sdss.org/dr14/spectro/">https://www.sdss.org/dr14/spectro/</a> ). This data set uses a stream data fetcher.
skdaccess. astro. voyager	Dictionary of Data Frames	Space Physics Data Facility ( <a href="https://spdf.gsfc.nasa.gov/pub/data/voyager/">https://spdf.gsfc.nasa.gov/pub/data/voyager/</a> )	≈ 0.1GB	Data from the Voyager mission ( <a href="https://voyager.jpl.nasa.gov/mission/">https://voyager.jpl.nasa.gov/mission/</a> ). This data set uses a cache data fetcher.
skdaccess. engineering. la. traffic_counts	Dictionary of Data Frames	Los Angeles Open Data ( <a href="https://data.lacity.org/">https://data.lacity.org/</a> )	≈ 0.1MB	Traffic count data in Los Angeles ( <a href="https://data.lacity.org/A-Livable-and-Sustainable-City/LADOT-Traffic-Counts-Summary/94wu-3ps3">https://data.lacity.org/A-Livable-and-Sustainable-City/LADOT-Traffic-Counts-Summary/94wu-3ps3</a> ). This data set uses a stream data fetcher.
skdaccess. finance. timeseries	Dictionary of Data Frames	Alpha Vantage ( <a href="https://www.alphavantage.co/">https://www.alphavantage.co/</a> )	Data product dependent	Stock data obtained from Alpha Vantage ( <a href="https://www.alphavantage.co/">https://www.alphavantage.co/</a> ). This data set uses a stream data fetcher.
skdaccess. geo. era_interim	XArray Dataset	The University Corporation for Atmospheric Research ( <a href="https://rda.ucar.edu/datasets/ds627.0/">https://rda.ucar.edu/datasets/ds627.0/</a> )	≈ 0.1GB / day	Atmospheric weather information from the ERA-Interim project at various pressure levels ( <a href="https://www.ecmwf.int/en/forecasts/datasets/archive-datasets/reanalysis-datasets/era-interim">https://www.ecmwf.int/en/forecasts/datasets/archive-datasets/reanalysis-datasets/era-interim</a> ). This data set uses a cache data fetcher.
skdaccess. geo. gldas	Dictionary of Data Frames	NASA Jet Propulsion Laboratory ( <a href="ftp://podaac-ftp.jpl.nasa.gov/allData/tellus/L3/gldas_monthly/netcdf">ftp://podaac-ftp.jpl.nasa.gov/allData/tellus/L3/gldas_monthly/netcdf</a> )	≈ 0.1GB	Land hydrology model produced by NASA. This version of the data is generated to match the GRACE temporal and spatial characteristics and is available as a complementary data product ( <a href="https://grace.jpl.nasa.gov/data/get-data/land-water-content/">https://grace.jpl.nasa.gov/data/get-data/land-water-content/</a> ). This data set uses a download data fetcher.

skdaccess. geo. grace	Dictionary of Data Frames	NASA Jet Propulsion Laboratory ( <a href="ftp://podaac-ftp.jpl.nasa.gov/allData/tellus/L3/land_mass/RL05/netcdf">ftp://podaac-ftp.jpl.nasa.gov/allData/tellus/L3/land_mass/RL05/netcdf</a> )	$\approx 0.1\text{GB}$	GRACE Tellus Monthly Mass Grids. 30-day measurements of changes in Earth's gravity field to quantify equivalent water thickness ( <a href="https://grace.jpl.nasa.gov/data/get-data/monthly-mass-grids-land/">https://grace.jpl.nasa.gov/data/get-data/monthly-mass-grids-land/</a> ). This data set uses a download data fetcher.
skdaccess. geo. grace. mascon	Dictionary of Data Frames	NASA Jet Propulsion Laboratory ( <a href="ftp://podaac.jpl.nasa.gov/allData/tellus/L3/mascon/RL05/JPL/CRI/netcdf">ftp://podaac.jpl.nasa.gov/allData/tellus/L3/mascon/RL05/JPL/CRI/netcdf</a> )	$\approx 1\text{GB}$	GRACE Tellus Monthly Mass Grids - Global Mascons. 30-day measurements of changes in Earth's gravity field to quantify equivalent water thickness ( <a href="https://grace.jpl.nasa.gov/data/get-data/jpl_global_mascons">https://grace.jpl.nasa.gov/data/get-data/jpl_global_mascons</a> ). This data set uses a download data fetcher.
skdaccess. geo. groundwater	Dictionary of Data Frames	USGS National Water Information System ( <a href="https://waterservices.usgs.gov/rest/DV-Service.html">https://waterservices.usgs.gov/rest/DV-Service.html</a> )	$\approx 1\text{GB}$	United States groundwater monitoring wells measuring the depth to water level ( <a href="https://waterservices.usgs.gov/">https://waterservices.usgs.gov/</a> ). This data set uses a download data fetcher.
skdaccess. geo. magnetometer	Dictionary of Data Frames	USGS National Geomagnetism Program ( <a href="https://geomag.usgs.gov/products/downloads.php">https://geomag.usgs.gov/products/downloads.php</a> )	$\approx 1\text{GB}$	Measurement of Earth's magnetic field from the USGS geomagnetism program ( <a href="https://geomag.usgs.gov/">https://geomag.usgs.gov/</a> ). This data set uses a stream data fetcher.
skdaccess. geo. mahali. rinex	List of rinex file paths	MIT-Haystack Observatory ( <a href="http://apollo.haystack.mit.edu/mahali-data/">http://apollo.haystack.mit.edu/mahali-data/</a> )	$\approx 10\text{GB}$	Rinex files from the MIT led NSF project studying the Earth's ionosphere with GPS ( <a href="http://mahali.mit.edu">http://mahali.mit.edu</a> ). This data set uses a cache data fetcher.
skdaccess. geo. mahali. tec	Dictionary of Data Frames	MIT-Haystack Observatory ( <a href="http://apollo.haystack.mit.edu/mahali-data/">http://apollo.haystack.mit.edu/mahali-data/</a> )	$\approx 1\text{GB}$	TEC measurements from the MIT led NSF project studying the Earth's ionosphere with GPS ( <a href="http://mahali.mit.edu">http://mahali.mit.edu</a> ). This data set uses a cache data fetcher.
skdaccess. geo. mahali. temperature	Dictionary of Data Frames	MIT-Haystack Observatory ( <a href="http://apollo.haystack.mit.edu/mahali-data/">http://apollo.haystack.mit.edu/mahali-data/</a> )	$\approx 0.1\text{GB}$	Temperature measurements from the MIT led NSF project studying the Earth's ionosphere with GPS ( <a href="http://mahali.mit.edu">http://mahali.mit.edu</a> ). This data set uses a stream data fetcher.
skdaccess. geo. modis	Dictionary of NumPy arrays	NASA MODIS ( <a href="https://ladsweb.modaps.eosdis.nasa.gov/tools-and-services/">https://ladsweb.modaps.eosdis.nasa.gov/tools-and-services/</a> )	$\approx 100\text{MB}/\text{image}$	Spectroradiometer aboard the NASA Terra and Aqua satellites that generates approximately daily images of the Earth's surface ( <a href="https://modis.gsfc.nasa.gov/">https://modis.gsfc.nasa.gov/</a> ). This data set uses a cache and stream data fetcher.

skdaccess. geo. pbo	Dictionary of Data Frames	UNAVCO Plate Boundary Observatory ( <a href="ftp://data-out.unavco.org/pub/products/position/pbo.nam08.pos.tar.gz">ftp://data-out.unavco.org/pub/products/position/pbo.nam08.pos.tar.gz</a> and <a href="https://www.unavco.org/data/gps-gnss/derived-products/derived-products.html">https://www.unavco.org/data/gps-gnss/derived-products/derived-products.html</a> )	$\approx$ 1GB	Daily GPS displacement time series measurements throughout the United States ( <a href="http://www.unavco.org/projects/major-projects/pbo/pbo.html">http://www.unavco.org/projects/major-projects/pbo/pbo.html</a> ). This data set uses a download data fetcher.
skdaccess. geo. sentinel.1	Dictionary of Numpy arrays	Alaska Satellite Facility ( <a href="https://www.asf.alaska.edu/sentinel/">https://www.asf.alaska.edu/sentinel/</a> )	$\approx$ 1 – 10GB / image	Synthetic Aperture Radar data from the Sentinel 1 satellites operated by the European Space Agency ( <a href="https://www.esa.int/Our_Activities/Observing_the_Earth/Copernicus/Sentinel-1">https://www.esa.int/Our_Activities/Observing_the_Earth/Copernicus/Sentinel-1</a> ). This data set uses a cache data fetcher.
skdaccess. geo. srtm	Dictionary of Numpy arrays	United States Geological Survey ( <a href="https://e4ftl01.cr.usgs.gov/MEASURES/">https://e4ftl01.cr.usgs.gov/MEASURES/</a> )	$\approx$ 100GB	Digital elevation data from the Shuttle Radar Topography Mission ( <a href="https://www2.jpl.nasa.gov/srtm/">https://www2.jpl.nasa.gov/srtm/</a> ). This data set uses a cache data fetcher.
skdaccess. geo. uavasar	Dictionary of Numpy arrays	NASA Jet Propulsion Laboratory ( <a href="https://uavasar.jpl.nasa.gov/cgi-bin/data.pl">https://uavasar.jpl.nasa.gov/cgi-bin/data.pl</a> )	Data product dependent	Synthetic Aperture Radar Single Look Complex data from the Uninhabited Aerial Vehicle Synthetic Aperture Radar ( <a href="https://uavasar.jpl.nasa.gov/">https://uavasar.jpl.nasa.gov/</a> ). This data set uses a cache data fetcher.
skdaccess. geo. wyoming .sounding	Dictionary of Data Frames	University of Wyoming ( <a href="http://weather.uwyo.edu/upperair/sounding.html">http://weather.uwyo.edu/upperair/sounding.html</a> )	$\approx$ 10GB	Sounding data from The University of Wyoming ( <a href="http://weather.uwyo.edu/upperair/sounding.html">http://weather.uwyo.edu/upperair/sounding.html</a> ). This data set has a cache and stream data fetcher.
skdaccess. planetary. .ode	Dictionary of Numpy arrays	Orbital Data Explorer at the University of Washington in St. Louis ( <a href="http://oderest.rsl.wustl.edu">http://oderest.rsl.wustl.edu</a> )	Data product dependent	Planetary data from PDS Geosciences Node's Orbital Data Explorer ( <a href="http://pds-geosciences.wustl.edu/default.htm">http://pds-geosciences.wustl.edu/default.htm</a> ). This data set uses a cache data fetcher
skdaccess. solar. .sdo	Dictionary of Numpy arrays	Solar Dynamics Observatory ( <a href="https://sdo.gsfc.nasa.gov/assets/img/browse/">https://sdo.gsfc.nasa.gov/assets/img/browse/</a> ) and the Joint Science Operations Center ( <a href="http://jsoc2.stanford.edu/data/aia/synoptic/">http://jsoc2.stanford.edu/data/aia/synoptic/</a>	Data product dependent	Images from the Solar Dynamics Observatory ( <a href="https://sdo.gsfc.nasa.gov">https://sdo.gsfc.nasa.gov</a> ). This data set uses a stream data fetcher.

### 3 Installation and Modes of Operation

The package can easily installed by using the standard Python “pip install” command:

```
> pip install scikit-dataaccess
```

After successful installation, a script called “skdaccess” allows users to specify the data sets that should

be downloaded from their original sources to the local machine. The PBO, GRACE and groundwater data sets must be downloaded using this script before they can be used. For example, to download the PBO data use:

```
> skdaccess pbo
```

The script also completes all necessary configurations to make the data access seamlessly available in the Python environment.

### 3.1 Modes of Operation

There are three modes of operation available for accessing the data through the skdaccess package. The two local options are "Download" and "Cache". Using the "Download" option, the dataset is downloaded to local disk before use. The "Cache" option allows for data of interest to be downloaded during use and stored in case of future use. The online option is "Stream", which accesses the data without storing a local copy.

### 3.2 The Skdaccess Script

This script downloads scientific data sets from preconfigured Web sources, makes them available offline on the user's machine, and configures the Python environment for data access.

For the following data sets, the skdaccess script must be used to download and prepare the data.

- GPS data from the Plate Boundary Observatory
- Depth to groundwater for wells in California
- Equivalent water thickness from GRACE Tellus Monthly Land Grids
- Equivalent water thickness from GLDAS

The skdaccess script does not download Kepler data, as the data is downloaded for each star individually the first time the star is accessed by the data fetcher.

To download a dataset, use the command with the dataset name as the argument. For example, to download groundwater data available from California type

```
> skdaccess groundwater
```

The data will be downloaded into the current directory, and the .skdaccess config file located in the user's home directory will be updated. Each data set can be downloaded into different directories depending on the user preferences.

To list all supported data sets, call

```
> skdaccess -l  
This utility can install one of the following data sets:
```

```
PBO - Plate Boundary Observatory GPS Time Series  
GRACE - Monthly Mass Grids  
GLDAS - Monthly estimates from GDLAS model in same resolution as GRACE  
Groundwater - Ground water daily values from across the US
```

Calling the script without any arguments provides a list of available commands as shown below.

```

> skdaccess
usage: skdaccess [-h] [-l] [-i LOCAL_DATA] [-c] [data_set]

The Sci-kit Data Access (skdaccess) package is a tool for integrating various
scientific data sets into the Python environment using a common interface.
This script can download different scientific data sets for offline analysis.

positional arguments:
  data_set           Name of data set

optional arguments:
  -h, --help          show this help message and exit
  -l, --list          List data sets
  -i LOCAL_DATA, --input LOCAL_DATA
                      Use LOCAL_DATA that has already been downloaded
  -c, --check         Print data location for data set

```

## 4 Scientific Data Access in Python

Data is retrieved in a Python program via a DataFetcher object. Each data set has its own data fetcher. There are two ways of handling the data: (1) directly accessing the data structure created by the DataFetcher, or (2) through an iterator interface provided by a data wrapper.

Data Access Example:

```

# First import the data generator for water
# Note: This assumes the groundwater data has been downloaded
from skdaccess.geo.groundwater import DataFetcher as waterDF

# Create a data fetcher and get the data wrapper:
fullDF = waterDF(start_date='2007-01-01', end_date='2011-01-01')
wdata = fullDF.output().get()

```

## 5 Usage Examples

The following examples show how to use the data fetcher for the data sets described earlier and displaying / plotting the data. These notebooks can be accessed at <https://github.com/MITHaystack/scikit-dataaccess/tree/master/skdaccess/examples>.

## 5.1 skdaccess.astro.kepler

MIT Computer-Aided Discovery Demo\_Kepler Last Checkpoint: a minute ago (autosaved) Logout

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In [1]: `#matplotlib notebook  
import matplotlib.pyplot as plt`

In [2]: `# Kepler Exoplanet Light Curves Time Series  
# Source: http://keplerscience.arc.nasa.gov  
# Light curve in relative flux versus phase`

In [3]: `from skdaccess.astro.kepler import DataFetcher as Kepler_DF  
from skdaccess.utilities.kepler_util import normalize  
from skdaccess.framework.param_class import *  
import numpy as np`

In [4]: `kepler_fetcher = Kepler_DF([AutoList(['009941662'])])`

In [5]: `kepler_data = kepler_fetcher.output().get()`  
Downloading data for 1 star(s)

In [6]: `normalize(kepler_data['009941662'])`

In [7]: `kepler_data['009941662'].head()`

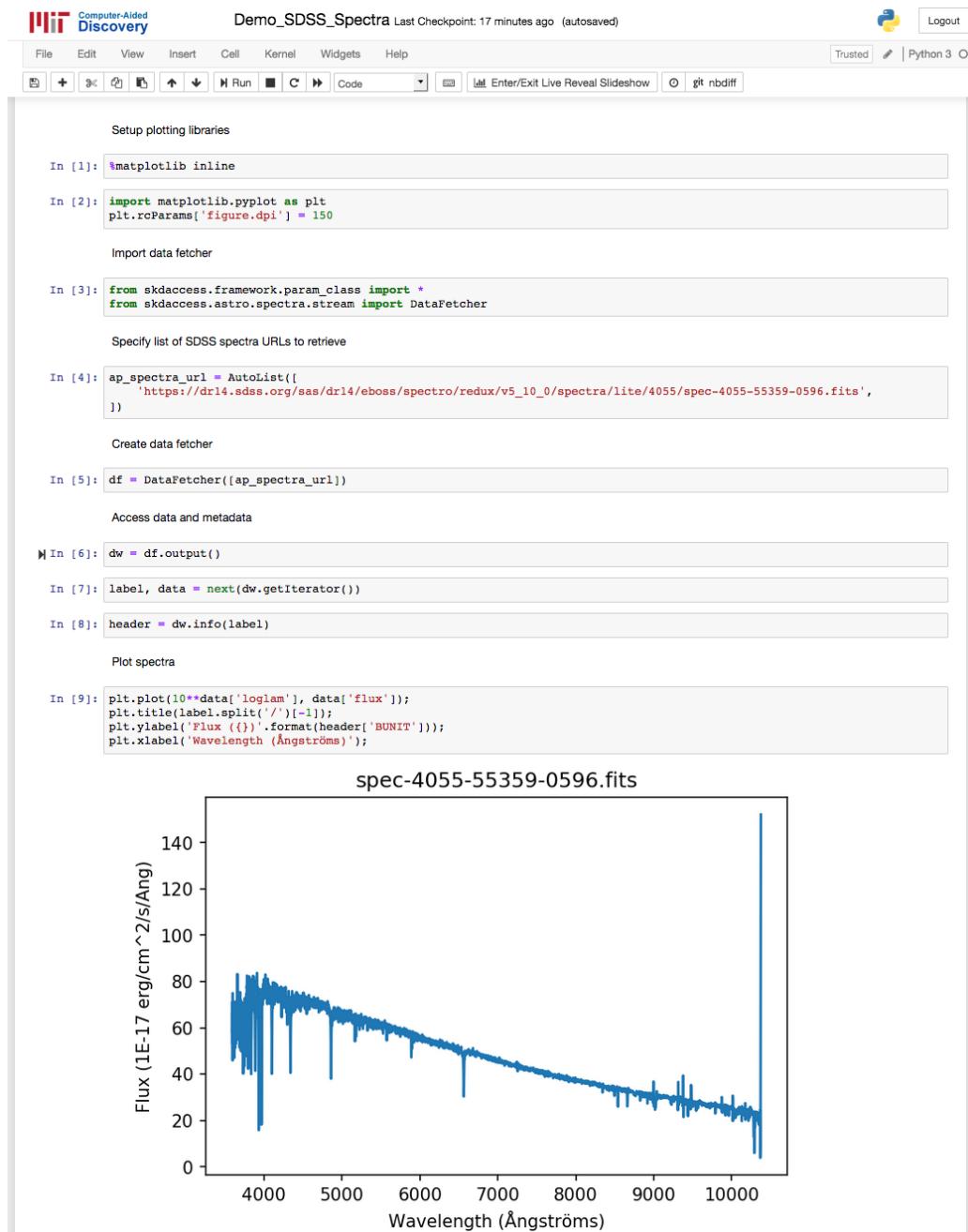
Out [7]:

CADENCENO	TIME	TIMECORR	SAP_FLUX	SAP_FLUX_ERR	SAP_BKG	SAP_BKG_ERR	PDCSAP_FLUX	PDCSAP_FLUX_ERR	SAP_QUALITY	PSF_CEN
568	120.539195	0.001042	1778533.750	33.049557	4841.642090	1.547178	0.995571	32.609024	0	
569	120.559629	0.001043	1778263.875	33.047188	4846.805664	1.546246	0.995457	32.732418	0	
570	120.580063	0.001044	1778347.750	33.048054	4848.539062	1.549641	0.995509	32.837833	0	
571	120.600498	0.001044	1778901.000	33.052914	4847.870117	1.543734	0.995785	32.684124	0	
572	120.620932	0.001045	1781658.250	33.081059	4852.192871	1.546612	0.997348	32.769455	0	

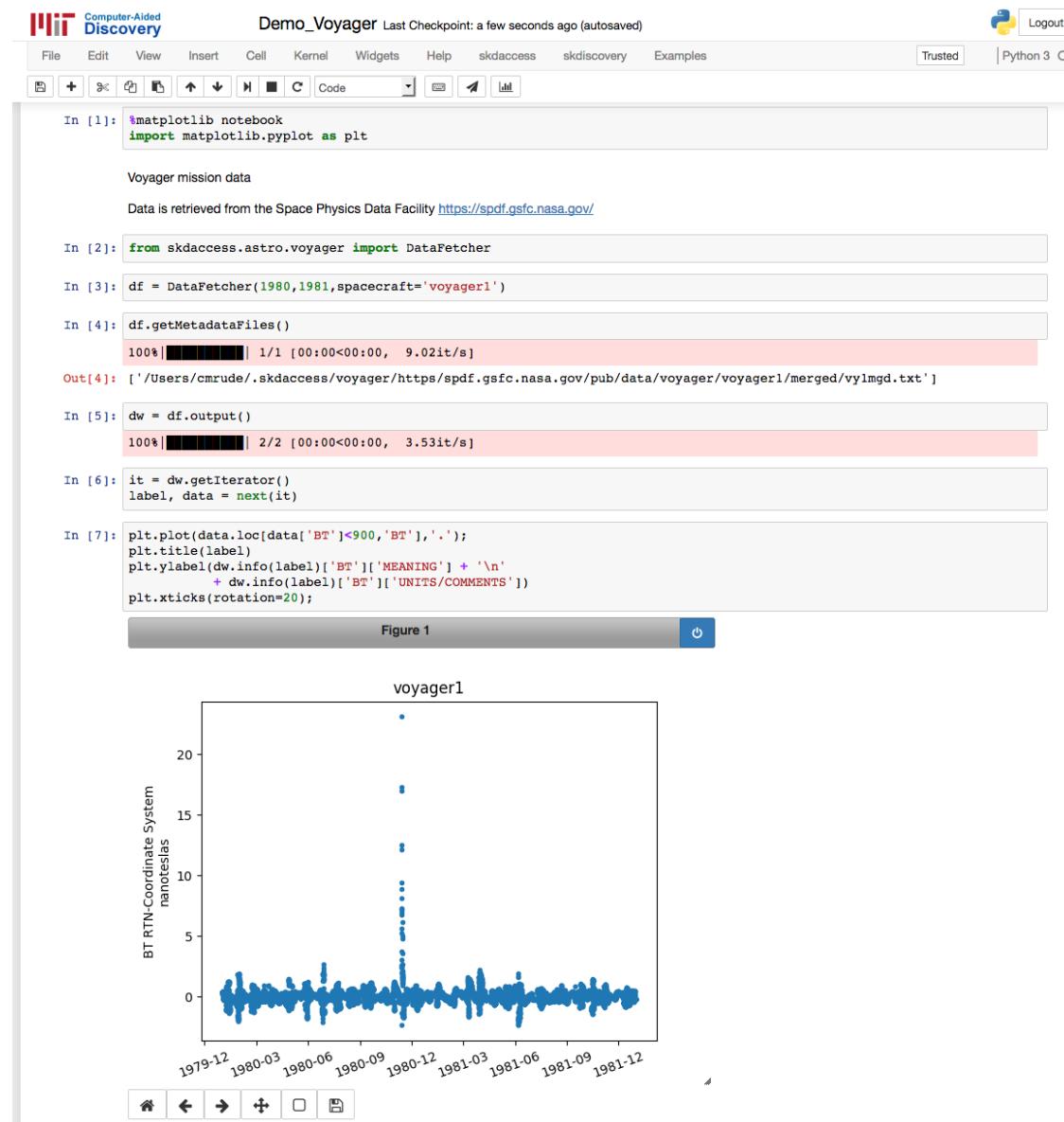
In [8]: `plt.figure(figsize=(8,4));  
data = kepler_data['009941662'].iloc[0:1000]  
plt.plot(np.array(data['TIME']) * 1.7636, data['PDCSAP_FLUX'], '.');  
plt.tight_layout();`

Figure 1

## 5.2 skdaccess.astro.spectra



### 5.3 skdaccess.astro.voyager



## 5.4 skdaccess.engineering.la.traffic\_counts

MIT Computer-Aided Discovery Demo\_Traffic\_Counts (unsaved changes) Logout  
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```
In [1]: %matplotlib inline
In [2]: import matplotlib.pyplot as plt
plt.rcParams['figure.dpi'] = 150
In [3]: import pandas as pd
In [4]: # LA Traffic count data between 2014-01-14 and 2014-01-16
# Source: https://data.lacity.org/A-Livable-and-Sustainable-City/LADOT-Traffic-Counts-Summary/94wu-3ps3
In [5]: from skdaccess.engineering.la.traffic_counts.stream import DataFetcher as TrafficDF
from skdaccess.framework.param_class import *
In [6]: # Create traffic count data fetcher
# Note, use parameter app_token to supply an application token to prevent throttling
# See: https://dev.socrata.com/docs/app-tokens.html
traffic_df = TrafficDF(start_time='2014-01-14', end_time='2014-01-16')
In [7]: # Create a data wapper
traffic_dw = traffic_df.output()
In [8]: # Retrieve results
label, data = next(traffic_dw.getIterator())
In [9]: # Select rows with east bound traffic on 2014-01-14
date = pd.to_datetime('2014-01-14')
cut_data = data[data.count_date == date]
cut_data = cut_data[cut_data.e_b != 0]
In [10]: # Create plot
plt.title('East Bound on {}'.format(date.strftime('%Y-%m-%d')));
plt.ylabel('Count');
tick_labels = cut_data.primary_street + ' and ' + cut_data.cross_street
plt.bar(x=range(1,4), height=cut_data.e_b, tick_label=tick_labels);
plt.xticks(rotation=15);
```

East Bound on 2014-01-14

Intersection	Count
11 th ST and OXFORD AV	1000
26 th ST and PATTON AV	550
27 th ST and PATTON AV	550

## 5.5 skdaccess.finance.timeseries

IIIIT Computer-Aided Discovery Demo\_Finance\_Time\_Series Last Checkpoint: a few seconds ago (unsaved changes) Logout Trusted Python 3 O

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Initial imports

```
In [1]: %matplotlib inline
```

```
In [2]: import matplotlib.pyplot as plt
plt.rcParams['figure.dpi'] = 150
```

```
In [3]: from skdaccess.framework.param_class import *
from skdaccess.finance.timeseries import DataFetcher
```

Select which symbol to retrieve

```
In [4]: stock_ap_list = AutoList(['SPY'])
```

Create a data fetcher

```
In [5]: stockdf = DataFetcher(stock_ap_list, 'daily', '2017-06-01')
```

Access the data

```
In [6]: dw = stockdf.output()
```

```
In [7]: label, data = next(dw.getIterator())
```

List the columns of the data

```
In [8]: data.columns
```

```
Out[8]: Index(['1. open', '2. high', '3. low', '4. close', '5. volume'], dtype='object')
```

Plot the closing price

```
In [9]: ax = data['4. close'].plot(title='SPY daily closing price');
ax.set_ylabel('price')
```

SPY daily closing price

price

date

## 5.6 skdaccess.geo.era\_interim

MIT Computer-Aided Discovery Demo\_ERA\_Interim Last Checkpoint: 02/07/2018 (unsaved changes) Logout  
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Data Citation  
European Centre for Medium-Range Weather Forecasts (2009): ERA-Interim Project. Research Data Archive at the National Center for Atmospheric Research, Computational and Information Systems Laboratory. <https://doi.org/10.5065/D6CR5RD9>.

```
In [1]: %matplotlib inline
import matplotlib.pyplot as plt
plt.rcParams['figure.dpi'] = 150
from getpass import getpass
import pandas as pd

In [2]: from skdaccess.framework.param_class import *
from skdaccess.geo.era_interim.cache import DataFetcher as EDF

Specify list of dates

In [3]: date_list = pd.date_range('2015-06-06 00:00:00', '2015-06-06 06:00:00', freq='6H')

Enter Research Data Archive (NCAR) credentials

In [4]: username='Enter username'
password = getpass()
.......
```

Create data fetcher

```
In [5]: edf = EDF(date_list=date_list, data_names=['Geopotential', 'Temperature'],
username=username, password=password)
```

Access data

```
In [6]: edw = edf.output()

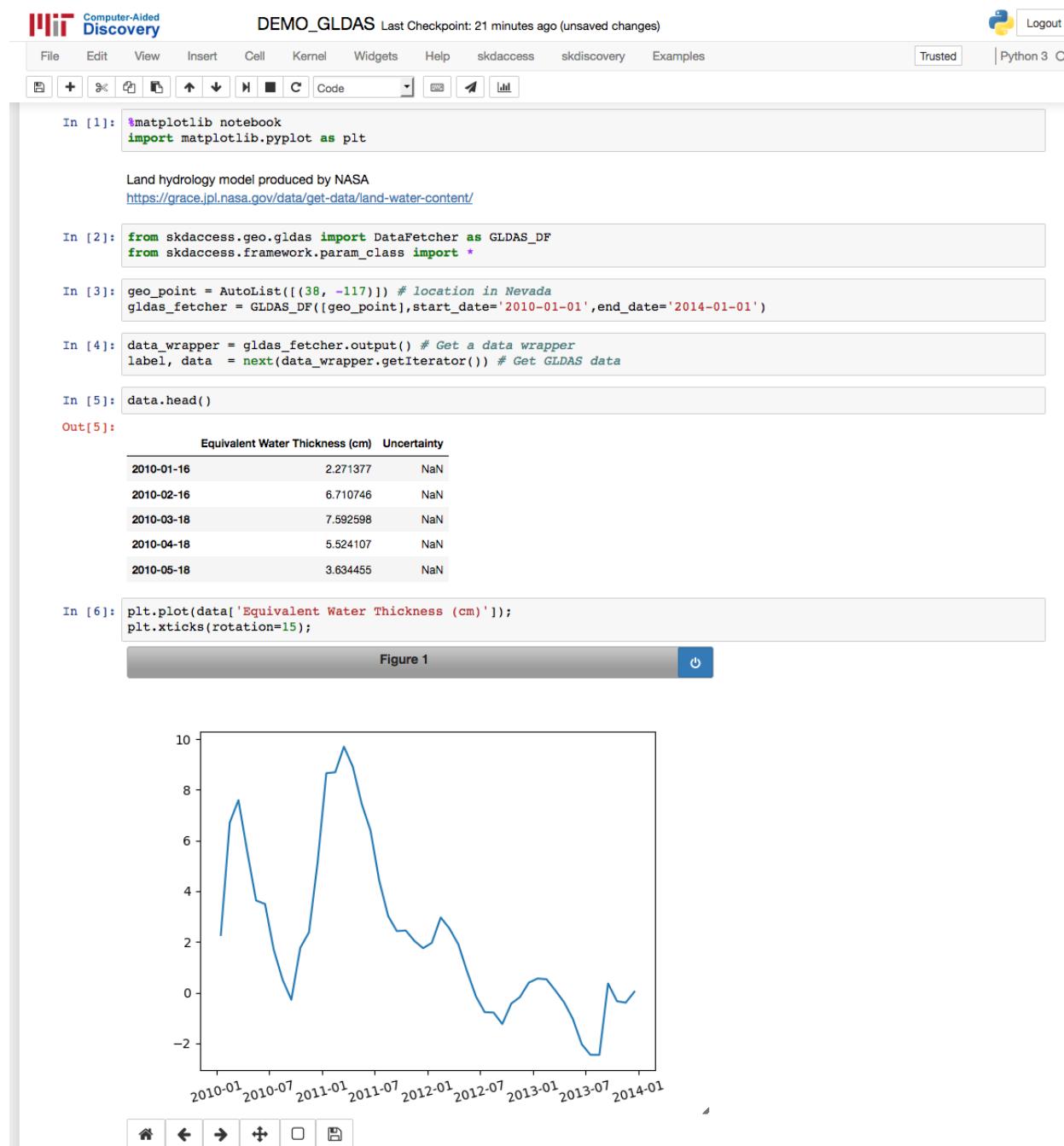
In [7]: iterator = edw.getIterator()
geo_label, geo_data = next(iterator)
temp_label, temp_data = next(iterator)
```

Plot temperature data

```
In [9]: plt.figure(figsize=(5,3.75));
plt.plot(temp_data[0,:75,350], temp_data['pressure']);
plt.gca().invert_yaxis();
plt.xlabel('Pressure');
plt.ylabel('Temperature');
```

Pressure (hPa)	Temperature (K)
1000	~250
800	~240
600	~225
400	~210
200	~190
100	~180

## 5.7 skdaccess.geo.gldas



## 5.8 skdaccess.geo.grace

MIT Computer-Aided Discovery Demo\_GRACE Last Checkpoint: Last Thursday at 1:56 PM (unsaved changes) Logout Trusted Python 3

```
In [1]: %matplotlib notebook
import matplotlib.pyplot as plt

In [2]: # Gravity Recovery and Climate Experiment (GRACE) Data
# Source: http://grace.jpl.nasa.gov/
# Current surface mass change data, measuring equivalent water thickness in cm, versus time
# with a resolution of 1x1 degree

In [3]: from skdaccess.geo.grace import DataFetcher as GR_DF
from skdaccess.utilities.grace.util import computeEWD
from skdaccess.framework.param_class import *

In [4]: geo_point = AutoList([(38, -117)]) # location in Nevada
grace_fetcher = GR_DF(geo_point,start_date='2010-01-01',end_date='2014-01-01')

In [5]: grace_data_wrapper = grace_fetcher.output() # Get a data wrapper
grace_data = grace_data_wrapper.get() # Get GRACE data

In [6]: grace_data['38, -117'].head()

Out[6]:
CSR      GFZ      JPL
Date
2010-01-17 00:00:00 -1.564813 -2.012964 -3.849506
2010-02-15 12:00:00  3.270973  2.980853  2.286743
2010-03-17 00:00:00 5.448462  3.252425  5.566326
2010-04-16 12:00:00 6.534275  4.668304  4.388913
2010-05-17 00:00:00 6.079895  4.922144  2.370604

In [7]: # Calibrate GRACE
scale_factor = grace_data_wrapper.info('38, -117')[['scale_factor']]
combined_ewd = computeEWD(grace_data['38, -117'], scale_factor)

In [8]: # Plot calibrated data
plt.figure();
plt.plot(combined_ewd);
plt.tight_layout()
```

Figure 1

The figure is a line graph titled "Figure 1". The x-axis is labeled with dates from 2010-01 to 2014-01, with major tick marks every six months. The y-axis is labeled with numerical values from -4 to 6, with major tick marks every 2 units. The data is represented by a single blue line. It starts at approximately -1.56 cm in January 2010, rises to a peak of about 3 cm in March 2010, then drops sharply to a minimum of about -3.5 cm in July 2010. It then rises again to a peak of about 6 cm in January 2011, before dropping to another minimum of about -3.5 cm in July 2011. There are several other smaller fluctuations throughout the period, with a notable peak of about 2 cm in January 2012 and another dip to about -4 cm in July 2013.

## 5.9 skdaccess.geo.grace.mascon

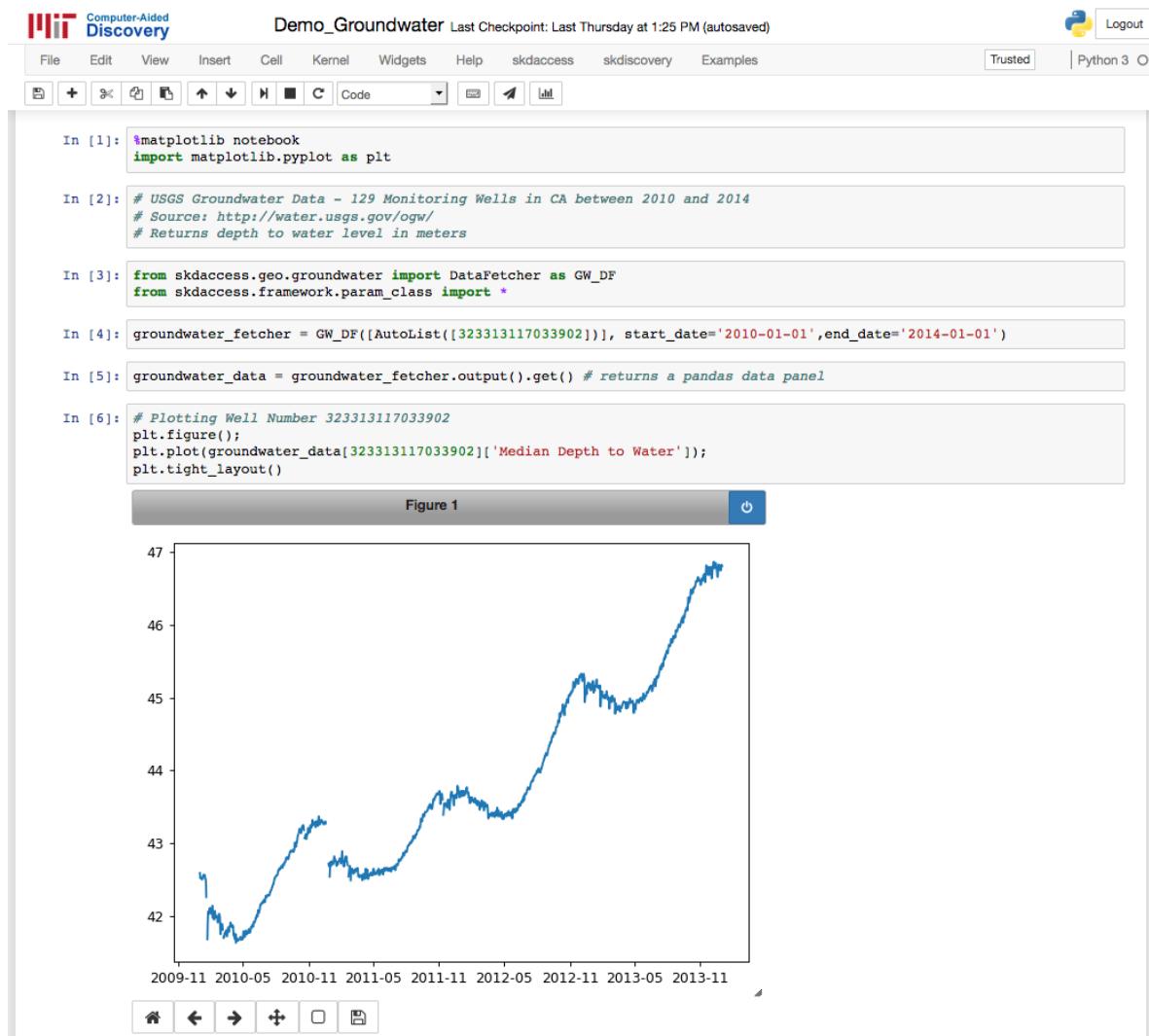
Computer-Aided Discovery Demo\_GRACE\_Mascon Last Checkpoint: 05/07/2018 (unsaved changes) Logout Trusted Python 3

```
In [1]: %matplotlib inline
In [2]: import matplotlib.pyplot as plt
plt.rcParams['figure.dpi']=150
In [3]: # Gravity Recovery and Climate Experiment (GRACE) Data
# Source: http://grace.jpl.nasa.gov/
# Current surface mass change data, measuring equivalent water thickness in cm, versus time
# This data fetcher uses results from the Mascon solutions
In [4]: from skdaccess.geo.grace.mascon.cache import DataFetcher as GR_DF
from skdaccess.framework.param_class import *
In [5]: geo_point = AutoList([(38, -117)]) # location in Nevada
grace_fetcher = GR_DF([geo_point],start_date='2010-01-01',end_date='2014-01-01')
In [6]: grace_data_wrapper = grace_fetcher.output() # Get a data wrapper
grace_label, grace_data = next(grace_data_wrapper.getIterator())# Get GRACE data
In [7]: grace_data.head()
Out[7]:
   EWD  EWD_Error
2010-01-16 12:00:00 -6.871147  1.856682
2010-02-15 00:00:00  2.071728  1.847683
2010-03-16 12:00:00  6.026260  2.089841
2010-04-16 00:00:00  3.524163  1.866020
2010-05-16 12:00:00  1.974874  1.937367

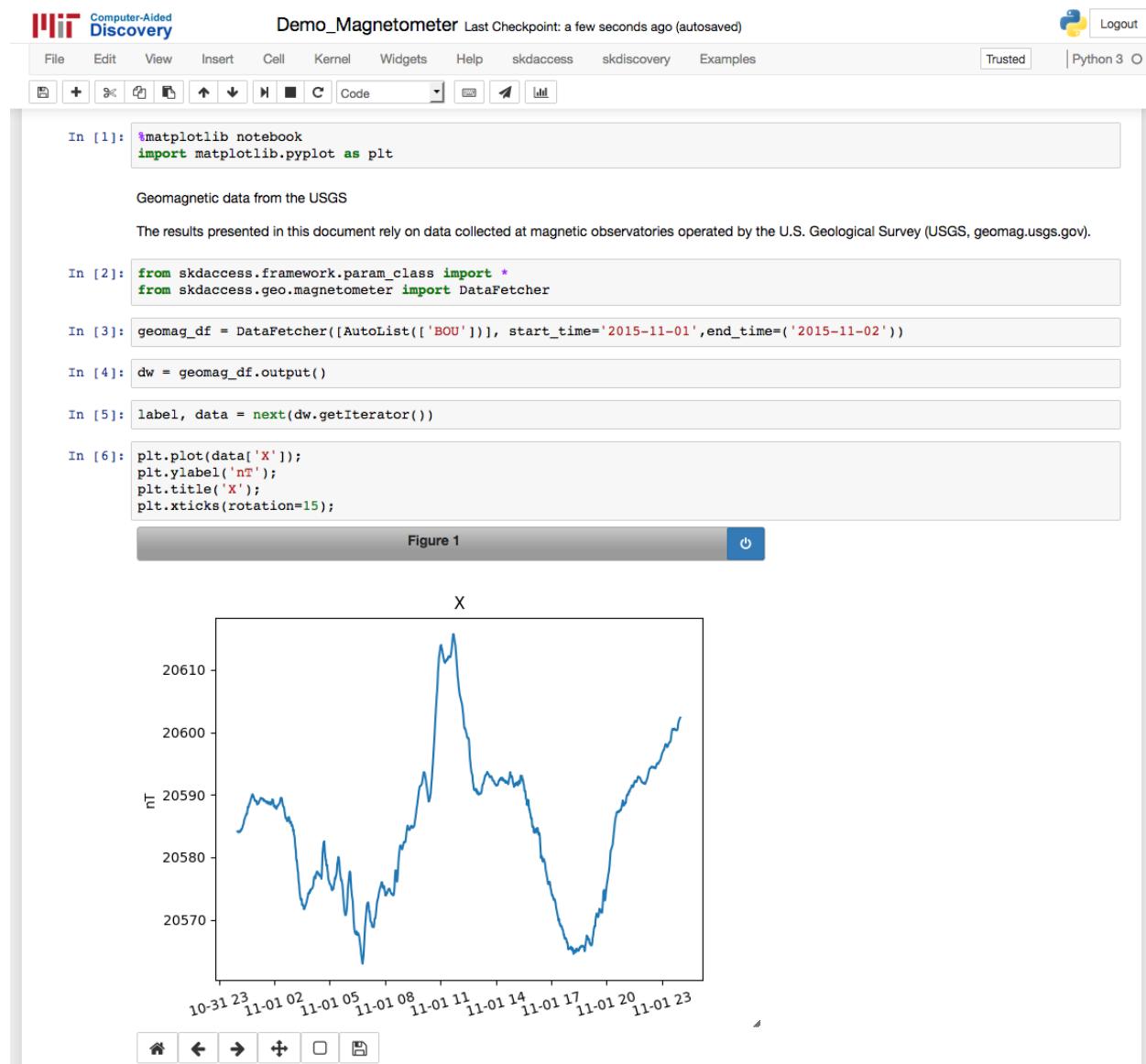
Get scale factor
In [8]: scale_factor = grace_data_wrapper.info(grace_label)['scale_factor']
Plot EWD X scale factor
In [9]: plt.plot(grace_data['EWD']*scale_factor);
plt.xticks(rotation=35);
```

Date	EWD (cm)	EWD_Error
2010-01-16 12:00:00	-6.871147	1.856682
2010-02-15 00:00:00	2.071728	1.847683
2010-03-16 12:00:00	6.026260	2.089841
2010-04-16 00:00:00	3.524163	1.866020
2010-05-16 12:00:00	1.974874	1.937367

## 5.10 skdaccess.geo.groundwater



## 5.11 skdaccess.geo.magnetometer

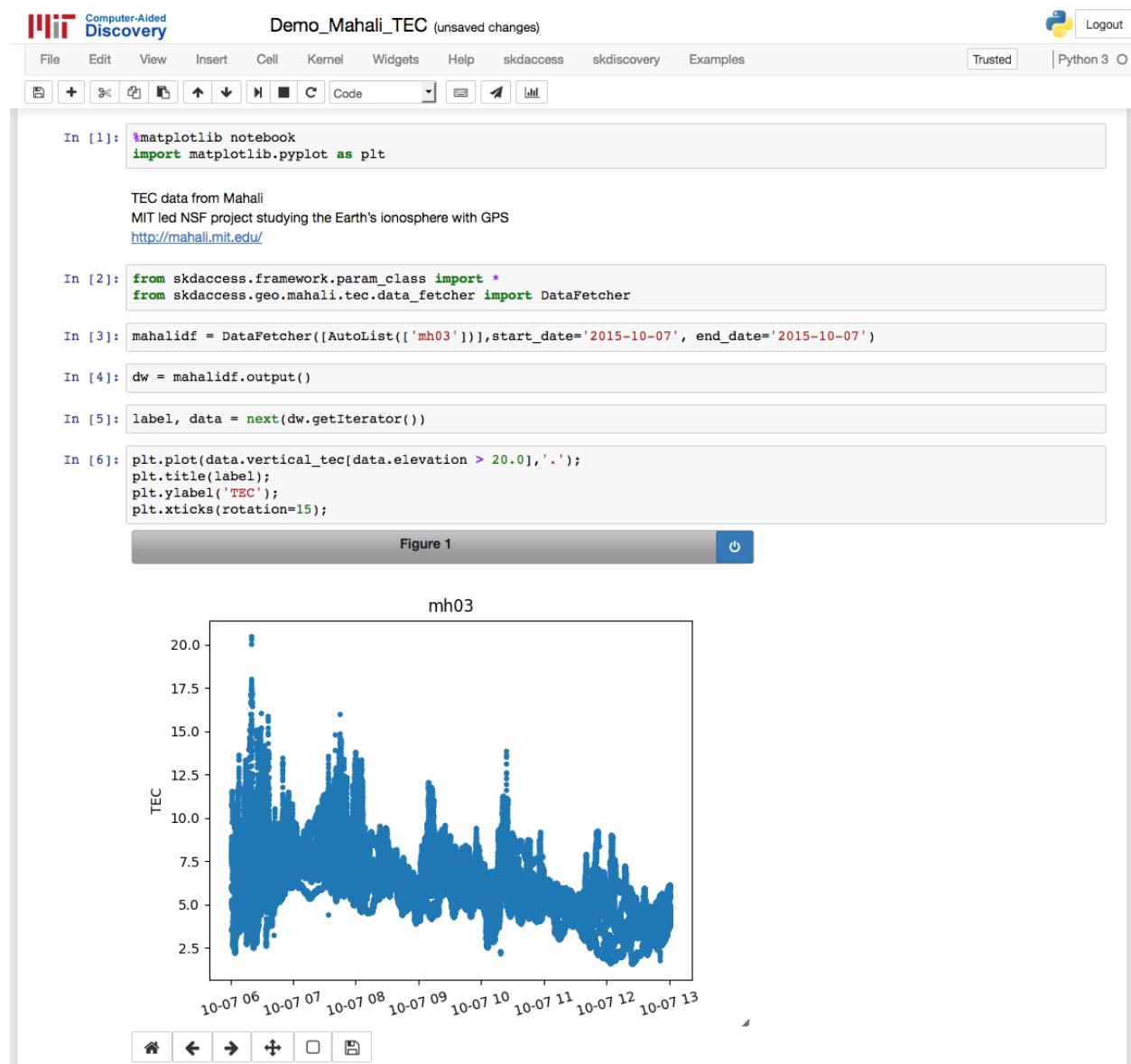


## 5.12 skdaccess.geo.mahali.rinex

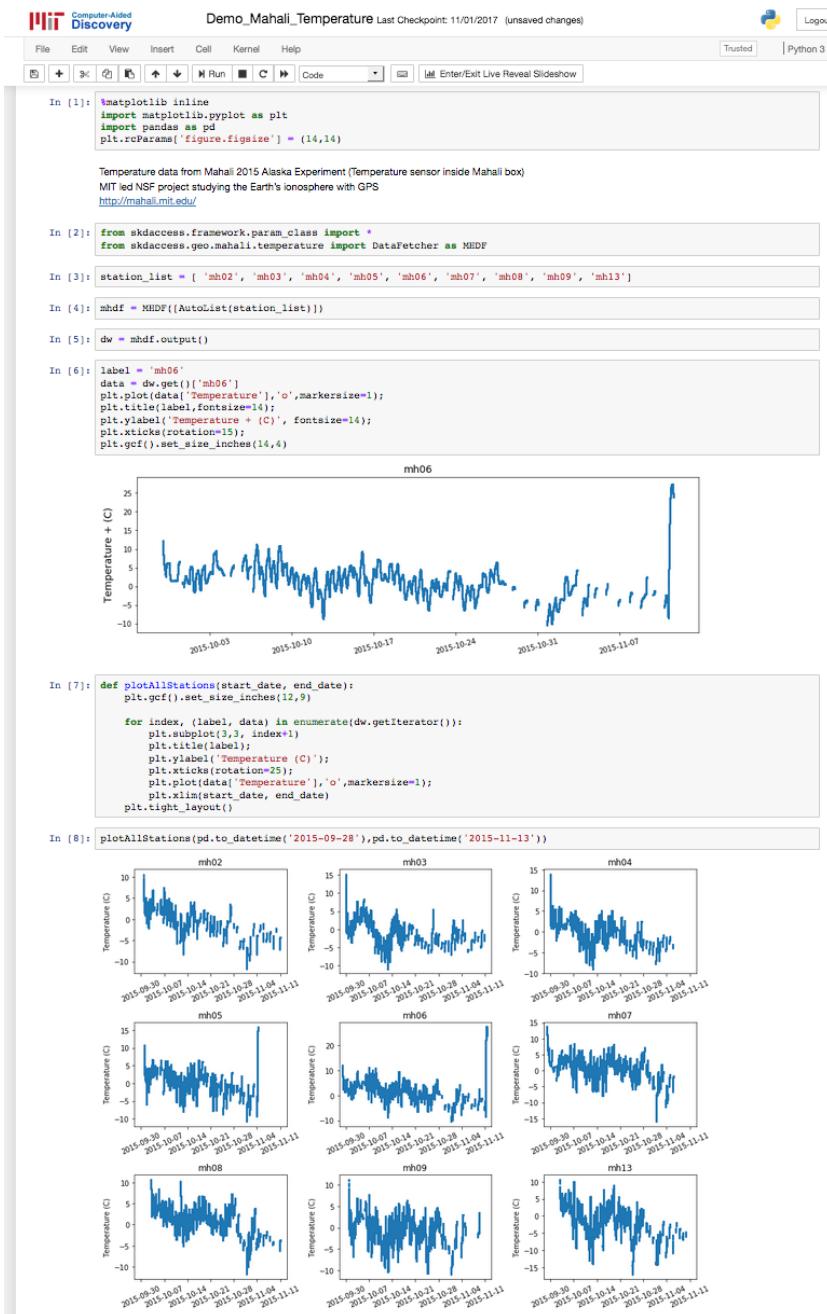
The screenshot shows a Jupyter Notebook interface with the following details:

- Title Bar:** MIT Computer-Aided Discovery Demo\_Mahali Last Checkpoint: a few seconds ago (autosaved)
- Toolbar:** File, Edit, View, Insert, Cell, Kernel, Widgets, Help, skdaccess, skdiscovery, Examples, Logout, Python 3
- Code Cells:**
  - In [1]: `%matplotlib notebook  
import matplotlib.pyplot as plt`
  - In [2]: `from skdaccess.framework.param_class import *  
from skdaccess.geo.mahali import DataFetcher as MAHALI_DF`
  - In [3]: `station_list = AutoList(['mh06']) # List of stations to download  
mahali_df = MAHALI_DF(station_list), start_date='2015-10-10', end_date='2015-10-10')`
  - In [4]: `data_wrapper = mahali_df.output() # Get a data wrapper  
Downloading mahali data  
100% |██████████| 2/2 [00:02<00:00, 1.38s/it]`
  - In [5]: `site, date, nav, obs = next(data_wrapper.getIterator()) # Get data location`
  - In [6]: `site, date, nav, obs # Print information about downloaded data`
- Output Cell (Out[6]):** `('mh06',  
Timestamp('2015-10-10 00:00:00'),  
'/Users/cmruude/.skdaccess/mahali/mh06283.15N',  
'/Users/cmruude/.skdaccess/mahali/mh06283.15O')`

## 5.13 skdaccess.geo.mahali.tec



## 5.14 skdaccess.geo.mahali.temperature



## 5.15 skdaccess.geo.modis.cache.reflectance

MIT Computer-Aided Discovery Demo\_MODIS Last Checkpoint: a few seconds ago (autosaved) Logout Trusted Python 3 O

In [1]: `%matplotlib inline`  
`import matplotlib.pyplot as plt`

MODIS surface reflectance product at 1 km resolution ("MOD09")  
<https://modis.gsfc.nasa.gov/data/>

In [2]: `# Import AutoParams, calibration and rescaling functions, and Stream Data Fetcher`  
`from skdaccess.framework.param_class import *`  
`from skdaccess.utilities.modis_util import calibrateModis, rescale`  
`from skdaccess.geo.modis.cache.reflectance import DataFetcher as MODISDF`

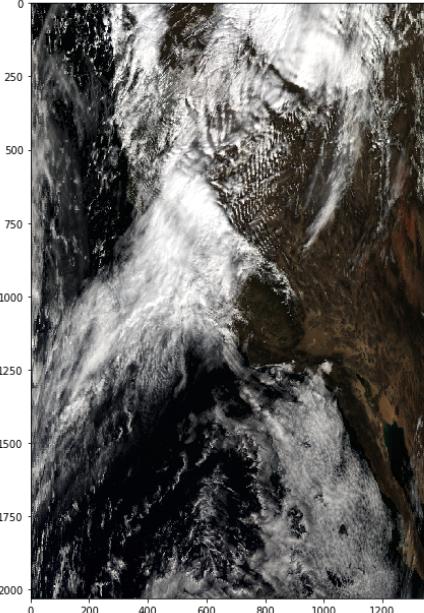
In [3]: `# Create MODIS data fetcher`  
`modis_df = MODISDF([AutoParam(38),AutoParam(119)], '2012-03-13', '2012-03-13')`

In [4]: `# Access data wrapper`  
`modis_dw = modis_df.output()`

In [5]: `# Use iterator to access data`  
`label, data = next(modis_dw.getIterator())`

In [6]: `# Calibrate and scale data`  
`calibrated_data = rescale(calibrateModis(data,modis_dw.info(label)))`

In [7]: `# Plot color image of result`  
`plt.gcf().set_size_inches(7,12);`  
`plt.imshow(calibrated_data);`



## 5.16 skdaccess.geo.pbo

MIT Computer-Aided Discovery Demo\_PBO Last Checkpoint: Last Thursday at 1:25 PM (unsaved changes)

File Edit View Insert Cell Kernel Widgets Help skdaccess skdDiscovery Examples Trusted Python 3 O Logout

```
In [1]: # Plate Boundary Observatory GPS Data
# Source: http://www.unavco.org/instrumentation/networks/status/pbo
# Time series data for GPS sensors (North, East, Up), displacement in meters versus time

In [2]: from skdaccess.geo.pbo import DataFetcher as PBO_DF
from skdaccess.framework.param_class import *

In [3]: %matplotlib notebook
import matplotlib.pyplot as plt

In [4]: # Latitude and Longitude range around Akutan Volcano
lat_range = AutoList((54,54.25))
lon_range = AutoList((-166, -165.6))
start_time = '2006-01-01'
end_time = '2015-06-01'

PBO_data_fetcher = PBO_DF(start_time, end_time, [lat_range, lon_range], mdyratio=.7)

In [5]: PBO_data = PBO_data_fetcher.output().get() # returns an ordered dictionary of data frames
100%|██████████| 6/6 [00:00<00:00, 17.98bit/s]

In [6]: PBO_data['AV06'].head()

Out[6]:
   HHMMSS JJJJJ.JJJJ      X      Y      Z     Sx     Sy     Sz    Rxxy    Rxzx ...    dN     dE     dU     Sn
0  2006-01-01 120000 53736.5 -3.629267e+06 -920656.48751 5.146731e+06 0.00339 0.00167 0.00460 0.508 -0.801 ... 0.00945 0.00935 -0.01095 0.00172
1  2006-01-02 120000 53737.5 -3.629267e+06 -920656.48670 5.146731e+06 0.00323 0.00160 0.00441 0.506 -0.801 ... 0.01064 0.00896 -0.01233 0.00165
2  2006-01-03 120000 53738.5 -3.629267e+06 -920656.48672 5.146731e+06 0.00332 0.00166 0.00450 0.495 -0.806 ... 0.01108 0.00937 -0.01432 0.00166
3  2006-01-04 120000 53739.5 -3.629267e+06 -920656.48650 5.146731e+06 0.00338 0.00169 0.00457 0.492 -0.802 ... 0.00803 0.00947 -0.02076 0.00170
4  2006-01-05 120000 53740.5 -3.629267e+06 -920656.48658 5.146731e+06 0.00331 0.00165 0.00446 0.490 -0.802 ... 0.01132 0.00890 -0.01179 0.00167

5 rows x 24 columns
```

```
In [7]: plt.figure();
plt.plot(PBO_data['AV06'][['dN']]);
plt.tight_layout()
```

Figure 1

## 5.17 skdaccess.geo.sentinel\_1

MIT Computer-Aided Discovery Demo\_Sentinel\_1 Last Checkpoint: 05/04/2018 (unsaved changes) Logout Trusted Python 3

```
In [1]: %matplotlib inline

In [2]: import matplotlib.pyplot as plt
plt.rcParams['figure.dpi'] = 150
import numpy as np
from getpass import getpass

In [3]: from skdaccess.geo.sentinel_1.cache import DataFetcher as S1DF

Supply Earth Data credentials

In [4]: username='Enter username'

In [5]: password = getpass()

Define urls for Sentinel 1 data and precise orbits

In [6]: slc_url_list = ['https://datapool.asf.alaska.edu/SLC/SA/S1A_IW_SLC__1SSV_20141103T195043_20141103T195057_003122_00395A_1'
satellite_url_list = ['https://slqc.asf.alaska.edu/aux_poeorb/S1A_OPER_AUX_POEORB_OPOD_20141124T123237_V20141102T225944']

Create data fetcher

In [7]: sldf = S1DF(slc_url_list, satellite_url_list, username, password, swath=3, polarization='VV')

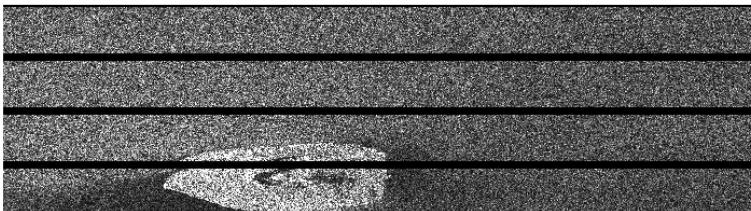
Access data

In [8]: sldw = sldf.output()
Retrieving SLC data
Retrieving orbit files
All files retrieved

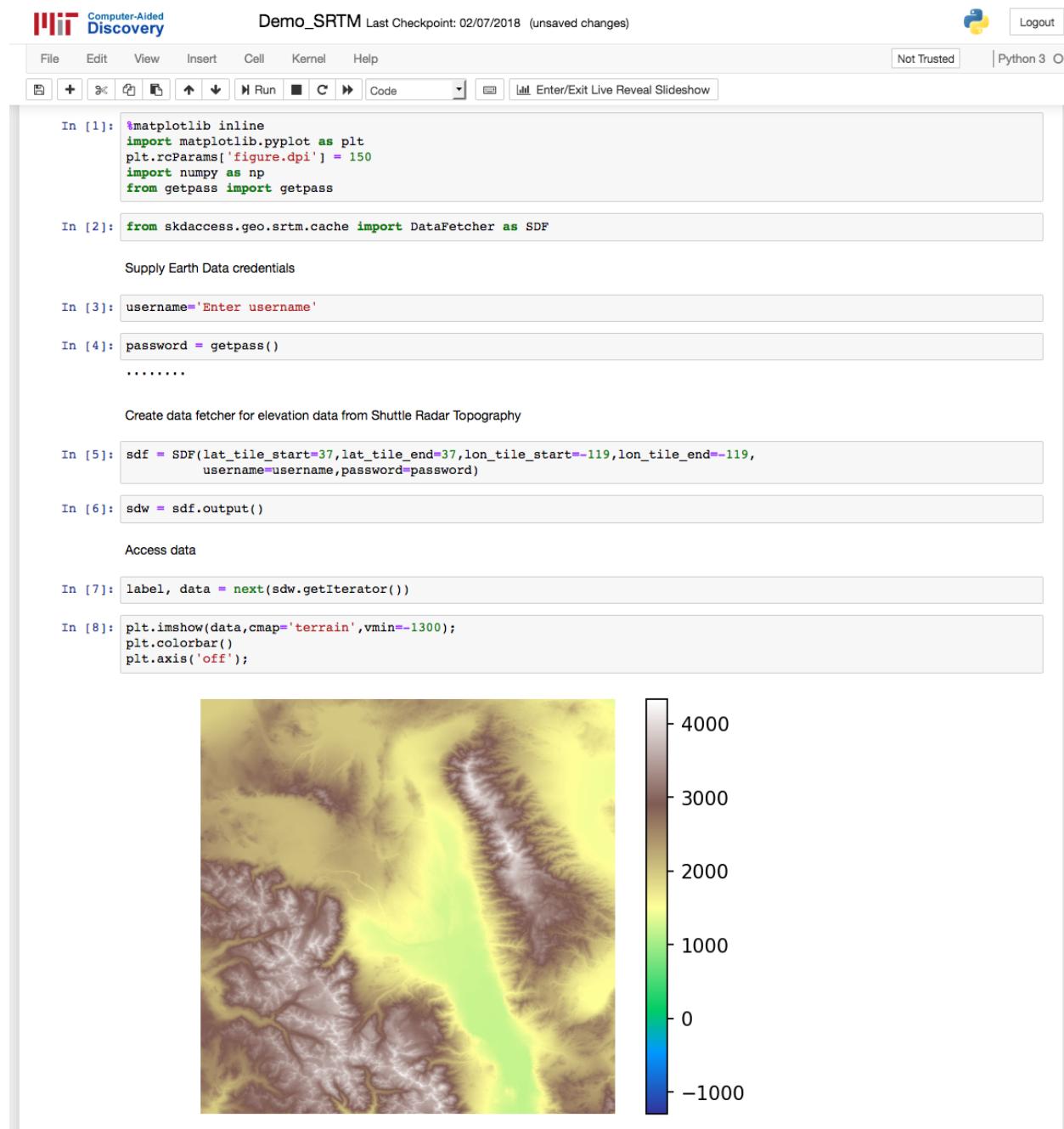
In [9]: label, data = next(sldw.getIterator())

In [10]: plt.title(label, fontsize=8)
plt.imshow(np.abs(data[:,::10]), vmin=0, vmax=100, cmap='gray', origin='lower')
plt.axis('off');

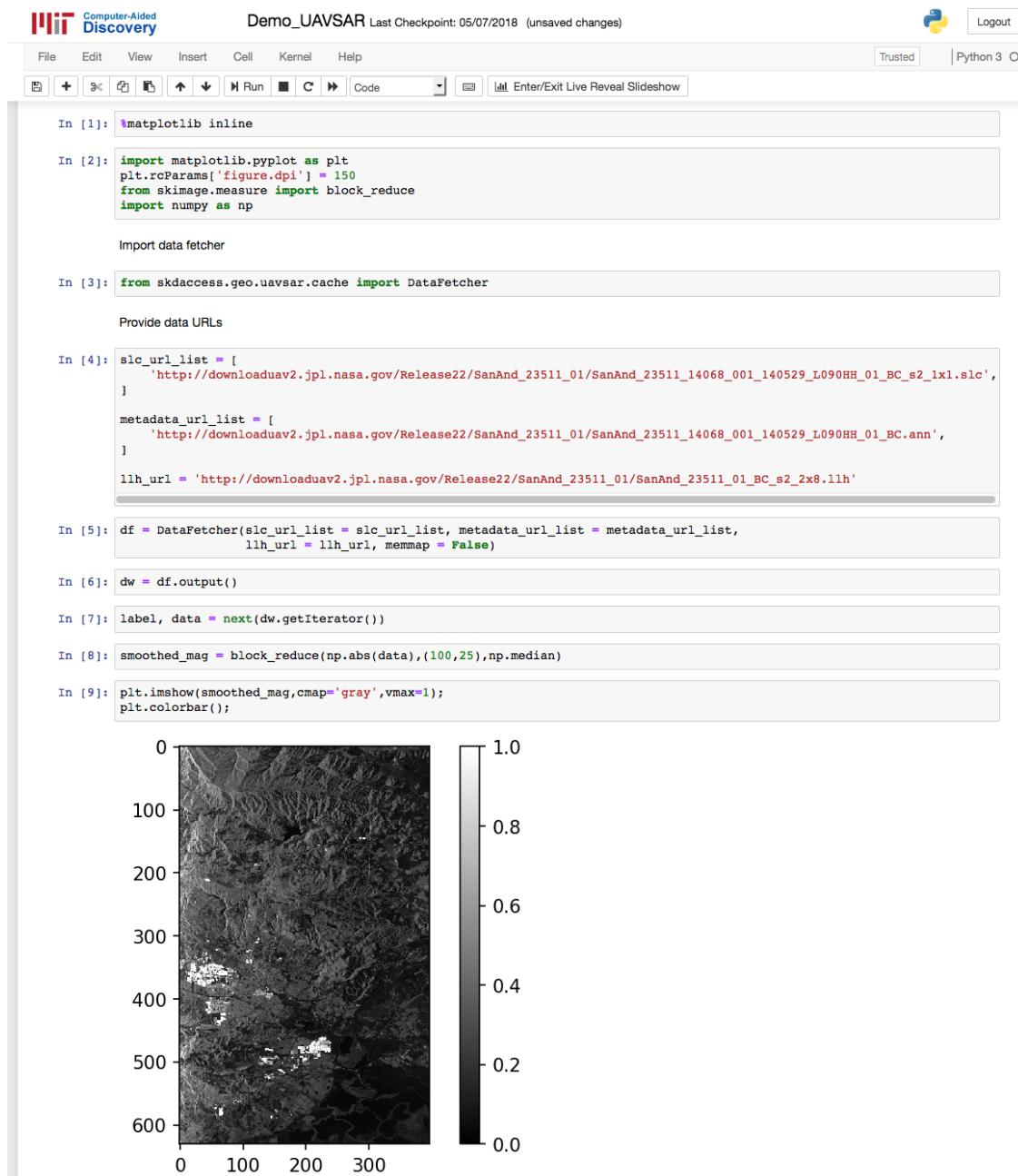
S1A_IW_SLC__1SSV_20141103T195043_20141103T195057_003122_00395A_F396.zip
```



## 5.18 skdaccess.geo.srtm



## 5.19 skdaccess.geo.uavasar



## 5.20 skdaccess.geo.wyoming\_sounding

Demo\_Wyoming\_Sounding Last Checkpoint: 02/07/2018 (unsaved changes) [Logout](#)

File Edit View Insert Cell Kernel Help Trusted Python 3 O

In [1]: `#matplotlib inline  
import matplotlib.pyplot as plt  
plt.rcParams['figure.dpi'] = 150`

In [2]: `from skdaccess.framework.param_class import *  
from skdaccess.geo.wyoming_sounding.cache import DataFetcher`

Create a data fetcher

In [3]: `sdf = DataFetcher(station_number='72493', year=2014, month=5, day_start=30, day_end=30, start_hour=12, end_hour=12)`

Access data

In [4]: `dw = sdf.output()`

In [5]: `label, data = next(dw.getIterator())`

In [6]: `data.head()`

Out[6]:

	PRES	HGHT	TEMP	DWPT	RElh	MIXR	DRCT	SKNT	THTA	THTE	THTV
0	1013.0	3	11.2	9.6	90	7.46	180.0	2.0	283.3	304.0	284.6
1	1012.0	11	11.2	6.2	71	5.91	190.0	2.0	283.4	300.0	284.4
2	1005.0	69	11.0	8.1	82	6.78	257.0	4.0	283.8	302.7	284.9
3	1000.0	111	10.8	8.0	83	6.77	305.0	6.0	283.9	302.9	285.1
4	976.8	305	9.4	7.4	88	6.66	290.0	4.0	284.4	303.1	285.6

In [8]: `plt.figure(figsize=(5,3.75))  
plt.plot(data['TEMP'],data['HGHT']);  
plt.ylabel('Height');  
plt.xlabel('Temperature');`

The figure is a line plot titled 'Height' on the y-axis and 'Temperature' on the x-axis. The y-axis ranges from 0 to 30,000 meters, and the x-axis ranges from -60 to 20 degrees Celsius. The data points show a general downward trend from higher temperatures and pressures at lower heights to lower temperatures and pressures at higher heights, with a notable minimum near the surface.

## 5.21 skdaccess.planetary.ode

The screenshot shows a Jupyter Notebook interface with the title 'Demo\_ODE' (autosaved). The notebook has several cells:

- Package imports** (Cell 11):

```
In [11]: #matplotlib notebook
from skdaccess.planetary.ode.cache import DataFetcher as OODEF
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
```
- Function definitions** (Cell 12):

```
In [12]: def unit_nans(array):
    return array / np.nanmin(array)/(np.isnan(array)) / np.nanmax(array))
```
- Technical foreword** (Cell 13):

SCINet data access relies on ODE's Representational State Transfer (REST) interface to query POS products, and on the [Geospatial Data Abstraction Library \(GDAL\)](#) to open and access the data in `Numpy` arrays. We can then use `matplotlib` to visualize the data.

It is decomposed into two elements: a data fetcher which retrieves the data's URLs and cache the data, and a data wrapper, which provide access to the data through a common interface for the mission, e.g., Mars, Mercury, Moon, Phobos, or Venus.

  - target: Aimed planetary body, (e.g., Mercury, Moon, Phobos, or Venus).
  - mission: Mission name, (e.g., MESSENGER).
  - instrument: Aimed instrument from the mission, e.g., HIRSS or CRISM.
  - product\_type: Type of product to look for, e.g., DTM or ROVIR/1.

You can find more information about the different missions, instruments, and product types supported by ODE on its website and its [user's manual](#). The data fetcher lets you define even other parameters to refine your query:

  - western\_lon: Western longitude to look for the data, from -180 to 180 (default: none).
  - eastern\_lon: Eastern longitude to look for the data, from -180 to 180 (default: none).
  - min\_lat: Minimal latitude to look for the data, from -90 to 90 (default: none).
  - max\_lat: Maximal latitude to look for the data, from -90 to 90 (default: none).
  - max\_obs\_time: Maximal observation time in Julian Julian Day (JULIAN\_DAY) (default: none).
  - max\_dt: Maximal observation time in Julian Julian Day (JULIAN\_DAY) (default: none).
  - product\_id: POS Product ID to look for, with wildcards (\*) allowed (default: none).
  - file\_id: File ID to look for, with wildcards (\*) allowed (default: none).
  - number\_products: Maximal number of products to return (ODE allows 100 at most) (default: 10).
  - result\_offset: Number of offset products, to go beyond the limit of 100 returned products (default: 0).
  - remove\_ndv: Replace the no-data value as mentioned in the data label by np.nan (default: true).
- Examples of data available for Mars** (Cell 14):

Mars data are accessible through the [Mars Orbital Data Explorer](#), whose interface can be used to explore the possible values to use in the data fetcher.
- Mars Global Surveyor** (Cell 15):

**MOLA - Mars Orbital Laser Altimeter**

ODEDF defines the data fetcher, and output() returns the data wrapper to access the data in Python. Here, we use the product ID to avoid the download of all the elevation data from MOLA.

Calling output() actually caches and opens the data; it shows the query URL, which stores the result of the query in XML. You can use it to get more information about the products, but also to assess some errors that would not be shown by the current interface. Then, it shows a list of files to be downloaded, and asks if you want to proceed to the download. If you are not satisfied with the query results, you just have to answer "no", to change the parameters' values of the data fetcher, and to re-run the cell.

```
In [15]: target = 'Mars'
mission = 'MGS'
instrument = 'MOLA'
product_type = 'MOLA'
product_id = '*v1*'

mola_data_fetcher = OODEF(target, mission, instrument, product_type,
product_id=product_id)
mola_data = mola_data_fetcher.output()

Query URL: https://roderest.rsl.vwml.edu/live2/?target=mars&id=MOLA&pt=MOLA&query=productType=&pd=output
=&MOLA[1]&[1]&fast=&product_id=v1#&
```

Files that will be downloaded (if not previously downloaded):

```
Product ID: MOLA29400007B.TMG
File type: TMG
Description: PRODUCT: INV. HEADER FILE
File name: mola29400007b.hdr
Description: PRODUCT: DATA FILE
File name: mola29400007b.tif
Description: PRODUCT: COLOR FILE
File name: mola29400007b.lbl
```

Do you want to proceed? [Y/n] Y

100% ██████████ 3/3 (01.04<00:00, 22.01s/a)

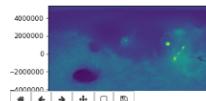
The resulting data wrapper gives you access to the data and some meta-data, for instance the projection or the extent of the raster.

```
In [16]: print(mola_data.keys())
odict_keys(['MOLA29400007B.TMG'])

In [17]: products = list(mola_data.data.keys())
print(products[0])
print(mola_data.data[products[0]].keys())
odict_keys(['mola29400007b.lbl'])

In [18]: file = list(mola_data.data[products[0]].keys())[0]
figure1 = plt.figure()
subfigure1 = figure1.add_subplot(111)
raster_im = plt.imshow(mola_data.data[product][file],
extent = mola_data.data[product][file]['Extent'],
cmap = 'gray',
interpolation = 'None')
raster_im.set_cmap('gray')
raster_im.set_interpolation('None')
raster_im.set_extent(mola_data.data[product][file]['Extent'])
plt.colorbar(raster_im)
plt.show()
```

Figure 1



## 5.22 skdaccess.solar.sdo

The screenshot shows a Jupyter Notebook interface titled "Demo\_SDO Last Checkpoint: a minute ago (unsaved changes)". The notebook has tabs for File, Edit, View, Insert, Cell, Kernel, Widgets, Help, skdaccess, skdiscovery, Examples, Trusted, and Python 3. The code cell In [1] contains:

```
In [1]: import matplotlib.pyplot as plt
import matplotlib.colors as colors
from matplotlib.colors import LogNorm
```

Access data from the Solar Dynamics Observatory  
<https://sdo.gsfc.nasa.gov/>

The code cell In [2] contains:

```
In [2]: from skdaccess.framework.param_class import *
from skdaccess.solar.sdo import DataFetcher as SDODF
```

The code cell In [3] contains:

```
In [3]: al_urls = AutoList(['https://sdo.gsfc.nasa.gov/assets/img/browse/2017/08/28/20170828_000000_256_HMIB.jpg',
                      'http://jsoc2.stanford.edu/data/aia/synoptic/2013/04/04/H0500/AIA20130404_0500_0335.fits'])
```

The code cell In [4] contains:

```
In [4]: df = SDODF(al_urls)
```

The code cell In [5] contains:

```
In [5]: dw = df.output()
dw.download('https://sdo.gsfc.nasa.gov/assets/img/browse/2017/08/28/20170828_000000_256_HMIB.jpg [Done]
Downloaded http://jsoc2.stanford.edu/data/aia/synoptic/2013/04/04/H0500/AIA20130404_0500_0335.fits [Done]
```

The code cell In [6] contains:

```
In [6]: for label, data in dw.getIterator():
    plt.figure()
    plt.axis('off')
    if label[-3:] == 'jpg':
        plt.imshow(data,cmap='gray')
    else:
        plt.imshow(data,cmap='gray',vmin=-5,vmax=20)
    plt.title(label,fontsize=8)
```

Figure 1

Figure 2

## Acknowledgements

Many thanks for support from NASA AIST NNX15AG84G, NSF ACI 1442997, NSF AGS-1343967, and the Amazon Web Services Research grants (PI: V. Pankratius).