測 **產** Measure Space

变間 Apr 28, 2019 · 5 min read

Use Zarr to access cloud storage just like your local file system

1. What is Zarr?

<u>Zarr</u> official website describes <u>Zarr</u> as a Python package providing an implementation of **chunked**, **compressed**, **N-dimensional arrays**. **Chunked** indicates that Zarr can handle very large datasets and fast data access. **Compressed** means that Zarr can save files using reasonable storge size which also means with less cost. **N-dimensional arrarys** reveals that Zarr can handle multi-dimension datasets just like Netcdf (e.g. geoscience datasets with time, x, y, and z four-dimensional datasets).

Some highlights as follow:

- Create N-dimensional arrays with any NumPy dtype.
- Chunk arrays along any dimension.
- Compress and/or filter chunks using any NumCodecs codec.
- Store arrays in memory, on disk, inside a Zip file, on S3, ...
- Read an array concurrently from multiple threads or processes.
- Write to an array concurrently from multiple threads or processes.
- Organize arrays into hierarchies via groups.

The most critical component of Zarr is that it can let you read and write files to cloud storage system (e.g. AWS S3) just like your local file system with the same convience of Netcdf format.

2. Read a netcdf file

Here we will use the surface air temperature data from NCEP Reanalysis Dataset as an example. I first downloaded the 2019 surface air temperature file air.sig995.2019.nc to my laptop. Then I use xarray to read the netcdf file.

```
ds = xr.open dataset('air.sig995.2019.nc')
ds
<xarray.Dataset>
Dimensions:
              (lat: 73, lon: 144, nbnds: 2, time: 116)
Coordinates:
  * lat
               (lat) float32 90.0 87.5 85.0 82.5 ... -82.5 -85.0
-87.5 - 90.0
               (lon) float32 0.0 2.5 5.0 7.5 10.0 ... 350.0 352.5
 * lon
355.0 357.5
  * time
               (time) datetime64[ns] 2019-01-01 2019-01-02 ... 2019-
04 - 26
Dimensions without coordinates: nbnds
Data variables:
    air
               (time, lat, lon) float32 ...
    time bnds (time, nbnds) float64 ...
Attributes:
    Conventions:
                    COARDS
                    mean daily NMC reanalysis (2014)
    title:
    history:
                    created 2017/12 by Hoop (netCDF2.3)
    description: Data is from NMC initialized
reanalysis\n (4x/day). These...
    platform:
                   Model
    References:
http://www.esrl.noaa.gov/psd/data/gridded/data.ncep.reana...
    dataset title: NCEP-NCAR Reanalysis 1
ds.air
<xarray.DataArray 'air' (time: 116, lat: 73, lon: 144)>
[1219392 values with dtype=float32]
Coordinates:
             (lat) float32 90.0 87.5 85.0 82.5 80.0 ... -82.5 -85.0
 * lat
-87.5 - 90.0
  * lon
            (lon) float32 0.0 2.5 5.0 7.5 10.0 ... 350.0 352.5
355.0 357.5
  * time
             (time) datetime64[ns] 2019-01-01 2019-01-02 ... 2019-
04-26
Attributes:
                   mean Daily Air temperature at sigma level 995
    long name:
    units:
                   degK
    precision:
                   2
                   11
    GRIB id:
    GRIB name:
                   TMP
    var desc:
                   Air temperature
    dataset:
                   NCEP Reanalysis Daily Averages
    level desc:
                   Surface
    statistic:
                  Mean
    parent_stat: Individual Obs
valid_range: [185.16 331.16]
    actual range: [198.4 314.]
```

To make Medium work, we log user data. By using Medium, you agree to our $\underline{\text{Privacy Policy}}$, including \times cookie policy.

3. Save it as Zarr format

Now, we are going to save the data as Zarr format. Because I don't have AWS S3 account, so I'm gonna save it to my laptop. Note that you can save it to AWS S3 directly with the help of <u>s3fs</u> package. <u>s3fs</u> is a Pythonic file interface to S3. It builds on top of <u>boto3</u> which is the Amazon Web Services (AWS) SDK for Python..

```
import zarr
import s3fs

# Compare the data if needed
compressor = zarr.Blosc(cname='zstd', clevel=3)
encoding = {vname: {'compressor': compressor} for vname in
ds.data_vars}
# Save to zarr
ds.to_zarr(store='zarr_example', encoding=encoding,
consolidated=True)

<xarray.backends.zarr.ZarrStore at 0x31a4d1ef0>
```

Now, we have saved the data as local zarr file.

The following code can be used to save data to AWS S3 zarr format.

```
import zarr
import s3fs

# AWS S3 path
s3_path = 's3://your_data_path/zarr_example'
# Initilize the S3 file system
s3 = s3fs.S3FileSystem()
store = s3fs.S3Map(root=s3_path, s3=s3, check=False)
# Compare the data if needed
compressor = zarr.Blosc(cname='zstd', clelve=3)
encoding = {vname: {'compressor': compressor} for vname in
ds.data_vars}
# Save to zarr
ds.to_zarr(store=store, encoding=encoding, consolidated=True)
```

4. Access part of the Zarr file

To make Medium work, we log user data. By using Medium, you agree to our $\underline{\text{Privacy Policy}}$, including \otimes \times cookie policy.

can directly access the whole or part of the dataset. Imagine that all the geospatial data (e.g. NCEP Reanalysis) is saved on cloud (it could be thousands of TB or PB), you can use several lines of code to read and download the file directly from AWS S3 without going throught the regular painful data downloading process. How coold is that!?

Again, because I don't have AWS S3 acess, I'm gonna use local zarr file as an example.

```
# Read Zarr file
zarr ds = xr.open zarr(store='zarr example', consolidated=True)
zarr ds
<xarray.Dataset>
Dimensions: (lat: 73, lon: 144, nbnds: 2, time: 116)
Coordinates:
              (lat) float32 90.0 87.5 85.0 82.5 ... -82.5 -85.0
 * lat
-87.5 -90.0
              (lon) float32 0.0 2.5 5.0 7.5 10.0 ... 350.0 352.5
 * lon
355.0 357.5
           (time) datetime64[ns] 2019-01-01 2019-01-02 ... 2019-
 * time
Dimensions without coordinates: nbnds
Data variables:
   air
              (time, lat, lon) float32 dask.array<shape=(116, 73,
144), chunksize=(58, 37, 72)>
   time bnds (time, nbnds) float64 dask.array<shape=(116, 2),
chunksize=(116, 2)>
Attributes:
   Conventions: COARDS
   References:
http://www.esrl.noaa.gov/psd/data/gridded/data.ncep.reana...
   dataset title: NCEP-NCAR Reanalysis 1
   description: Data is from NMC initialized
reanalysis\n (4x/day). These...
   history:
                  created 2017/12 by Hoop (netCDF2.3)
   platform:
                   Model
   title:
                   mean daily NMC reanalysis (2014)
```

See, it's easy. Actually, here zarr only read the metadata of the data file rather than loading all the real data. This is very helpful especially when the data size is large. Now, I'd like to read part of the zarr file. What should I do? For example, we can read the air temperature in January 2019.

```
import pandas as pd
```

```
zarr Jan = zarr ds.sel(time=time period)
zarr Jan
<xarray.Dataset>
Dimensions: (lat: 73, lon: 144, nbnds: 2, time: 31)
Coordinates:
              (lat) float32 90.0 87.5 85.0 82.5 ... -82.5 -85.0
 * lat
-87.5 -90.0
             (lon) float32 0.0 2.5 5.0 7.5 10.0 ... 350.0 352.5
 * lon
355.0 357.5
              (time) datetime64[ns] 2019-01-01 2019-01-02 ... 2019-
 * time
01-31
Dimensions without coordinates: nbnds
Data variables:
              (time, lat, lon) float32 dask.array<shape=(31, 73,
144), chunksize=(31, 37, 72)>
   time bnds (time, nbnds) float64 dask.array<shape=(31, 2),
chunksize=(31, 2)>
Attributes:
   Conventions:
                  COARDS
   References:
http://www.esrl.noaa.gov/psd/data/gridded/data.ncep.reana...
   dataset_title: NCEP-NCAR Reanalysis 1
   description: Data is from NMC initialized
reanalysis\n (4x/day). These...
   history: created 2017/12 by Hoop (netCDF2.3)
   platform:
                  Model
   title:
                   mean daily NMC reanalysis (2014)
```

Now we have the data only in January 2019. Cool!

Note that the following code can be used to access AWS S3 zarr file.

```
# AWS S3 path
s3_path = 's3://your_data_path/zarr_example'
# Initilize the S3 file system
s3 = s3fs.S3FileSystem()
sotre = s3fs.S3Map(root=s3_path, s3=s3, check=False)
# Read Zarr file
ds = xr.open_zarr(store=store, consolidated=True)
```

5. The secret of fast access — `consolidated=True`

Once the data are static and can be regarded as read-only, at least for the metadata/structure of the dataset hierarchy, the many metadata objects can be

6. Summary

Using Zarr, we can easily read and write files to could storage system (e.g. AWS S3). This is extremely helpful for large geospatial data access using cloud storage system. The **compress** and **consolidated** functionality of Zarr can also help use save storge cost and increase data access spead. N-dimensional data reading and writing is a very hot topic right now. Except Zarr, there are some other packages developing now. So far, Zarr is the leading one.

