CNRM_Oahu_Projection

February 23, 2022

1 Climate Change Exploration: Maunalua Bay, Oahu, Hawai'i

1.1 Fall 2021

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1.1.2 2021-12-08

All scripts and data can be accessed from Aloha Aina Repo

Code derived from CMIP6 PanGeo Gallery

Note: This analysis pulls data from CNRM-ESM2-1 only, you will find two other notebook files (BCC_Oahu_Projection.ipynb & IPSL_Oahu_Projection.ipynb) that will pull from other climate model sources (BCC-CSM2-MR & IPSL-CM6A-LR).

```
[2]: ## import libraries:
     from matplotlib import pyplot as plt
     import numpy as np
     import pandas as pd
     import xarray as xr
     import zarr
     import fsspec
     import gcsfs
     import s3fs
     import kedro
     import nc time axis
     import plotly.express as px
     import metpy
     from metpy.units import units
     %matplotlib inline
     %config InlineBackend.figure_format = 'retina'
     plt.rcParams['figure.figsize'] = 12, 6
```

```
[4]:
       activity_id institution_id
                                                        experiment_id member_id \
                                       source_id
     0 HighResMIP
                              CMCC
                                    CMCC-CM2-HR4
                                                  highresSST-present
                                                                       r1i1p1f1
     1 HighResMIP
                              CMCC
                                    CMCC-CM2-HR4
                                                  highresSST-present
                                                                       r1i1p1f1
     2 HighResMIP
                              CMCC
                                    CMCC-CM2-HR4
                                                  highresSST-present
                                                                       r1i1p1f1
     3 HighResMIP
                                                  highresSST-present
                                                                       r1i1p1f1
                              CMCC
                                    CMCC-CM2-HR4
     4 HighResMIP
                              CMCC
                                    CMCC-CM2-HR4
                                                  highresSST-present
                                                                       r1i1p1f1
     5 HighResMIP
                              CMCC
                                    CMCC-CM2-HR4
                                                  highresSST-present
                                                                       r1i1p1f1
     6 HighResMIP
                              CMCC
                                    CMCC-CM2-HR4
                                                  highresSST-present
                                                                       r1i1p1f1
                              CMCC
     7 HighResMIP
                                    CMCC-CM2-HR4
                                                  highresSST-present
                                                                       r1i1p1f1
     8 HighResMIP
                              CMCC
                                    CMCC-CM2-HR4
                                                  highresSST-present
                                                                       r1i1p1f1
                              CMCC
     9 HighResMIP
                                    CMCC-CM2-HR4
                                                  highresSST-present
                                                                       r1i1p1f1
       table_id variable_id grid_label
                         ps
     0
           Amon
                                     gn
     1
           Amon
                       rsds
                                     gn
     2
           Amon
                       rlus
                                     gn
     3
           Amon
                       rlds
                                     gn
     4
           Amon
                        psl
                                     gn
     5
           Amon
                        prw
                                     gn
     6
           Amon
                       hurs
                                     gn
     7
           Amon
                       huss
                                     gn
     8
           Amon
                        hus
                                     gn
     9
           Amon
                       hfss
                                     gn
                                                                               version
                                                     zstore
                                                             dcpp_init_year
        gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      NaN 20170706
       gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      NaN
                                                                           20170706
        gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      {\tt NaN}
                                                                           20170706
        gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      {\tt NaN}
                                                                           20170706
        gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      {\tt NaN}
                                                                           20170706
        gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      NaN
                                                                           20170706
     6 gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      NaN
                                                                           20170706
        gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      NaN
                                                                           20170706
     8 gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      NaN
                                                                           20170706
        gs://cmip6/CMIP6/HighResMIP/CMCC/CMCC-CM2-HR4/...
                                                                      NaN 20170706
[5]: ## Query for projection CMIP6 data
     df_3hr_pr = df[(df.table_id == '3hr') & (df.variable_id == 'pr')]
     len(df_3hr_pr)
     run_counts = df_3hr_pr.groupby(['source_id', 'experiment_id'])['zstore'].count()
     run counts
[5]: source id
                        experiment id
     BCC-CSM2-MR
                       historical
                                               1
                                               1
                       ssp126
                       ssp245
                                               1
```

	ssp370	1
	ssp585	1
CNRM-CM6-1	highresSST-present	1
0111W1 0110 1	historical	3
	ssp126	1
	ssp245	1
	ssp370	1
	ssp585	1
CNRM-CM6-1-HR	highresSST-present	1
CNRM-ESM2-1	historical	1
	ssp126	1
	ssp245	1
	ssp370	1
	ssp585	1
GFDL-CM4	1pctCO2	2
	abrupt-4xCO2	2
	amip	2
	historical	2
	piControl	2
GFDL-CM4C192	highresSST-future	1
	highresSST-present	1
GFDL-ESM4	1pctCO2	1
	abrupt-4xCO2	1
	esm-hist	1
	historical	1
	ssp119	1
	ssp126	1
	ssp370	1
GISS-E2-1-G	historical	2
HadGEM3-GC31-HM	highresSST-present	1
HadGEM3-GC31-LM	highresSST-present	1
HadGEM3-GC31-MM	highresSST-present	1
IPSL-CM6A-ATM-HR	highresSST-present	1
IPSL-CM6A-LR	highresSST-present	1
	historical	15
	piControl	1
	ssp126	3
	ssp245	2
	ssp370	10
	ssp585	1
MRI-ESM2-0	historical	1
Name: zstore, dtype: int64		

 $Fig. 1: Future and historical CO2 emissions scenarios featured in CMIP6 Source: \\ https://www.carbonbrief.org/cmip6-the-next-generation-of-climate-models-explained$

1.1.3 Pulling CNRM-ESM2-1, ssp2-4.5 Projection

```
[6]: ## querty for 3hr, precipitaion for ssp 2-4.5 projection from CNRM-ESM2-1
     df_3hr_ssp245_CNRM_pr = df[(df.table_id == '3hr') & (df.variable_id == 'pr') &__
     → (df.experiment_id== 'ssp245') & (df.source_id== 'CNRM-ESM2-1') ]
     len(df_3hr_ssp245_CNRM_pr)
     df_3hr_ssp245_CNRM_pr
[6]:
                                          source_id experiment_id member_id \
            activity_id institution_id
     68835 ScenarioMIP
                          CNRM-CERFACS CNRM-ESM2-1
                                                           ssp245 r1i1p1f2
           table_id variable_id grid_label \
     68835
                3hr
                             pr
                                        gr
                                                       zstore dcpp init year \
     68835 gs://cmip6/CMIP6/ScenarioMIP/CNRM-CERFACS/CNRM...
                                                                         NaN
             version
     68835 20190328
[7]: ## pull data
     # get the path to a specific zarr store (the first one from the dataframe above)
     zstore = df_3hr_ssp245_CNRM_pr.zstore.values[-1]
     print(zstore)
     # create a mutable-mapping-style interface to the store
     mapper = fsspec.get_mapper(zstore)
     # open it using xarray and zarr
     ds_proj = xr.open_zarr(mapper, consolidated=True)
     ds_proj
    gs://cmip6/CMIP6/ScenarioMIP/CNRM-CERFACS/CNRM-
    ESM2-1/ssp245/r1i1p1f2/3hr/pr/gr/v20190328/
[7]: <xarray.Dataset>
    Dimensions:
                      (lat: 128, lon: 256, time: 251288, axis_nbounds: 2)
     Coordinates:
       * lat
                      (lat) float64 -88.93 -87.54 -86.14 -84.74 ... 86.14 87.54 88.93
                      (lon) float64 0.0 1.406 2.812 4.219 ... 354.4 355.8 357.2 358.6
       * lon
                      (time) datetime64[ns] 2015-01-01T01:30:00 ... 2100-12-31T22:...
       * time
         time_bounds (time, axis_nbounds) datetime64[ns]
     dask.array<chunksize=(62822, 1), meta=np.ndarray>
     Dimensions without coordinates: axis_nbounds
    Data variables:
                      (time, lat, lon) float32 dask.array<chunksize=(600, 128, 256),
         pr
    meta=np.ndarray>
     Attributes: (12/55)
```

CMIP6_CV_version: cv=6.2.3.0-7-g2019642

Conventions: CF-1.7 CMIP-6.2

EXPID: CNRM-ESM2-1_ssp245_r1i1p1f2

activity_id: ScenarioMIP

arpege_minor_version: 6.3.2
branch_method: standard

•••

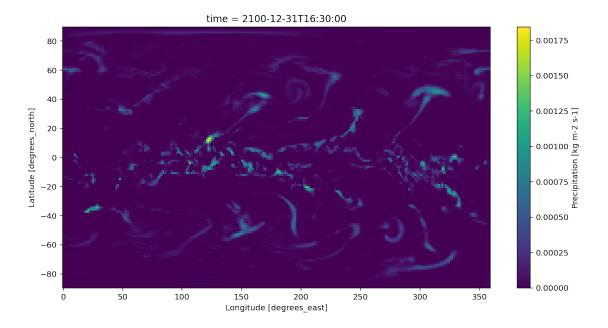
status: 2019-10-25; created; by nhn2@columbia.edu

netcdf_tracking_ids: hdl:21.14100/215d187a-7fa5-41cd-a59b-7fe164306a61...

version_id: v20190328

[8]: ## Plot a map from a specific date: global coverage ds_proj.pr.sel(time='2100-12-31T16:30:00.000000000').squeeze().plot()

[8]: <matplotlib.collections.QuadMesh at 0x7ff244b0e4f0>



- [9]: # # Create logical masks for lat and lon variables for oahu # bouding box: -158.5698,20.9057,-157.406,22.0022 mask_lon = (ds_proj.pr.lon >= 201.43) & (ds_proj.pr.lon <= 202.59) mask_lat = (ds_proj.pr.lat >= 20.91) & (ds_proj.pr.lat <= 22.00)
- [7]: # Apply lat/lon masks to the field, then calculate averages over the lat and one of one dimensions
 oahu_pr_proj=ds_proj.pr.where(mask_lon & mask_lat, drop = True)

```
## remove times associated with leap years (remove feb 29 from all recorded_
     \rightarrow years)
    oahu_pr_proj = oahu_pr_proj.sel(time=~((oahu_pr_proj.time.dt.month == 2) &_
     oahu_pr_proj
     ## group by day of year and avg by day
    oahu_pr_proj['dayofyear'] = xr.DataArray(oahu_pr_proj.indexes['time'].
     ⇒strftime('%Y-%m-%d'), coords=oahu_pr_proj.time.coords)
    oahu_pr_proj_avg = oahu_pr_proj.groupby('dayofyear').mean('time',_
     →keep_attrs=True) #retain attributes for metpy conversion in nxt step
    oahu_pr_proj_avg
[7]: <xarray.DataArray 'pr' (dayofyear: 31390, lat: 1, lon: 1)>
    dask.array<stack, shape=(31390, 1, 1), dtype=float32, chunksize=(1, 1, 1),
    chunktype=numpy.ndarray>
    Coordinates:
      * lat
                   (lat) float64 21.71
                   (lon) float64 202.5
       * lon
      * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
    Attributes:
        cell measures:
                             area: areacella
        cell methods:
                             area: time: mean
                             at surface; includes both liquid and solid phases. ...
        description:
        history:
                             none
        interval_operation: 900 s
        interval_write:
                             3 h
        long name:
                             Precipitation
        online_operation:
                             average
        standard_name:
                             precipitation_flux
        units:
                             kg m-2 s-1
[8]: ## daily sum of precip ssp3-7.0 Projection
    oahu_pr_proj_sum_245 = oahu_pr_proj.groupby('dayofyear').sum('time',_
      →keep attrs=True) #retain attributes for metpy conversion in nxt step
    oahu_pr_proj_sum_245
[8]: <xarray.DataArray 'pr' (dayofyear: 31390, lat: 1, lon: 1)>
    dask.array<stack, shape=(31390, 1, 1), dtype=float32, chunksize=(1, 1, 1),
    chunktype=numpy.ndarray>
    Coordinates:
                   (lat) float64 21.71
      * lat
                   (lon) float64 202.5
       * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
    Attributes:
        cell measures:
                             area: areacella
```

```
description:
                               at surface; includes both liquid and solid phases. ...
          history:
                               none
                               900 s
          interval_operation:
          interval_write:
                               3 h
                               Precipitation
          long_name:
          online_operation:
                               average
          standard_name:
                               precipitation_flux
                               kg m-2 s-1
          units:
 [9]: # Make metpy recognize the units
      oahu_pr_proj_sum_245 = oahu_pr_proj_sum_245.metpy.quantify()
      # convert kg/m2/sec to in/day
      density water = units('kg / m^3') * 1000
      oahu_pr_proj_converted_int_ssp245_sum = (oahu_pr_proj_sum_245 / density_water)
      oahu_pr_proj_converted_int_ssp245_sum = oahu_pr_proj_converted_int_ssp245_sum.
       →metpy.convert_units('inches / day')
      oahu_pr_proj_converted_int_ssp245_sum = oahu_pr_proj_converted_int_ssp245_sum.
       →mean("lon").mean("lat")
      oahu_pr_proj_converted_int_ssp245_sum
 [9]: <xarray.DataArray 'pr' (dayofyear: 31390)>
      <Quantity(dask.array<mean_agg-aggregate, shape=(31390,), dtype=float32,</pre>
      chunksize=(1,), chunktype=numpy.ndarray>, 'inch / day')>
      Coordinates:
        * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
                      Bouding
                               box
                                    coordinates
                                                             projection
                                                                         analysis
                                                                                  Source:
     Fig.
                                                  used for
     https://boundingbox.klokantech.com/
[49]: # Cast our xarray to dataframe -- daily averages
      oahu_pr_proj_df = oahu_pr_proj_converted_int.to_dataframe().reset_index()
      oahu_pr_proj_df.head(20)
[49]:
           dayofyear
                            pr
      0
          2015-01-01 0.000390
      1
          2015-01-02 0.001306
      2
          2015-01-03 0.022132
          2015-01-04 0.196844
      3
      4
          2015-01-05 0.441790
      5
          2015-01-06 0.076026
      6
          2015-01-07 0.059438
      7
          2015-01-08 0.060714
      8
          2015-01-09 0.026974
          2015-01-10 0.005410
      9
      10 2015-01-11 0.001053
```

area: time: mean

cell_methods:

```
11 2015-01-12 0.009215
     12 2015-01-13 0.644469
     13 2015-01-14 0.055369
     14 2015-01-15 0.019334
     15 2015-01-16 0.442971
     16 2015-01-17 0.001024
     17 2015-01-18 0.054217
     18 2015-01-19 0.481935
     19 2015-01-20 0.000166
[10]: # Cast our xarray to dataframe for daily sums
     oahu_pr_proj_converted_int_ssp245_sum_df =_u
      -oahu_pr_proj_converted_int_ssp245_sum.to_dataframe().reset_index()
     oahu_pr_proj_converted_int_ssp245_sum_df.head(20)
[10]:
          dayofyear
                          pr
         2015-01-01 0.003117
     1
         2015-01-02 0.010450
         2015-01-03 0.177055
     2
     3
         2015-01-04 1.574751
     4
         2015-01-05 3.534323
     5
         2015-01-06 0.608212
     6
         2015-01-07 0.475508
     7
         2015-01-08 0.485709
         2015-01-09 0.215790
     8
     9
         2015-01-10 0.043283
     10 2015-01-11 0.008423
     11 2015-01-12 0.073718
     12 2015-01-13 5.155749
     13 2015-01-14 0.442954
     14 2015-01-15 0.154673
     15 2015-01-16 3.543765
     16 2015-01-17 0.008195
     17 2015-01-18 0.433738
     18 2015-01-19 3.855481
     19 2015-01-20 0.001325
[66]: ## to export df, daily avq
     ## oahu pr proj df.to_csv('oahu ssp245_2015_2100_avq.csv', index = False)
[11]: ## to export df, daily sum
     oahu_pr_proj_converted_int_ssp245_sum_df.to_csv('oahu_ssp245_2015_2100_total.
      ⇔csv', index = False)
```

1.1.4 Exploring CNRM-ESM2-1, ssp3-7.0 Projection

```
[19]: | ## querty for 3hr, precipitaion for ssp 3-7.0 projection from CNRM-ESM2-1
      df_3hr_ssp370_CNRM_pr = df[(df.table_id == '3hr') & (df.variable_id == 'pr') &__
      → (df.experiment_id== 'ssp370') & (df.source_id== 'CNRM-ESM2-1') ]
      len(df_3hr_ssp370_CNRM_pr)
      df_3hr_ssp370_CNRM_pr
[19]:
                                           source_id experiment_id member_id \
             activity_id institution_id
      69219 ScenarioMIP
                                                            ssp370 r1i1p1f2
                           CNRM-CERFACS CNRM-ESM2-1
            table_id variable_id grid_label \
      69219
                 3hr
                              pr
                                         gr
                                                        zstore dcpp init year \
      69219 gs://cmip6/CMIP6/ScenarioMIP/CNRM-CERFACS/CNRM...
                                                                          NaN
              version
      69219 20190328
[20]: ## pull data
      # get the path to a specific zarr store (the first one from the dataframe above)
      zstore2 = df_3hr_ssp370_CNRM_pr.zstore.values[-1]
      print(zstore2)
      # create a mutable-mapping-style interface to the store
      mapper2 = fsspec.get_mapper(zstore2)
      # open it using xarray and zarr
      ds_proj_ssp370 = xr.open_zarr(mapper2, consolidated=True)
      ds_proj_ssp370
     gs://cmip6/CMIP6/ScenarioMIP/CNRM-CERFACS/CNRM-
     ESM2-1/ssp370/r1i1p1f2/3hr/pr/gr/v20190328/
[20]: <xarray.Dataset>
     Dimensions:
                       (lat: 128, lon: 256, time: 251288, axis_nbounds: 2)
      Coordinates:
        * lat
                       (lat) float64 -88.93 -87.54 -86.14 -84.74 ... 86.14 87.54 88.93
                       (lon) float64 0.0 1.406 2.812 4.219 ... 354.4 355.8 357.2 358.6
        * lon
        * time
                       (time) datetime64[ns] 2015-01-01T01:30:00 ... 2100-12-31T22:...
          time_bounds (time, axis_nbounds) datetime64[ns]
      dask.array<chunksize=(62822, 1), meta=np.ndarray>
      Dimensions without coordinates: axis_nbounds
      Data variables:
                       (time, lat, lon) float32 dask.array<chunksize=(449, 128, 256),
          pr
     meta=np.ndarray>
      Attributes: (12/55)
```

CF-1.7 CMIP-6.2 Conventions: EXPID: CNRM-ESM2-1_ssp370_r1i1p1f2 ScenarioMIP AerChemMIP activity_id: 6.3.2 arpege_minor_version: branch_method: standard variable_id: pr variant label: r1i1p1f2 xios commit: 1442-shuffle 2019-11-03; created; by nhn2@columbia.edu status: hdl:21.14100/2a291f7e-b9c9-4c68-b33b-cbfc153a587f... netcdf_tracking_ids: version id: v20190328 [21]: # Apply lat/lon masks to the field, then calculate averages over the lat and \rightarrow lon dimensions oahu_pr_proj_ssp370=ds_proj_ssp370.pr.where(mask_lon & mask_lat, drop = True) ## remove times associated with leap years (remove feb 29 from all recorded_ \rightarrow years) oahu_pr_proj_ssp370 = oahu_pr_proj_ssp370.sel(time=~((oahu_pr_proj_ssp370.time. →dt.month == 2) & (oahu_pr_proj_ssp370.time.dt.day == 29))) oahu_pr_proj_ssp370 ## group by day of year and avg by day oahu_pr_proj_ssp370['dayofyear'] = xr.DataArray(oahu_pr_proj_ssp370. →indexes['time'].strftime('%Y-%m-%d'), coords=oahu_pr_proj_ssp370.time.coords) oahu_pr_proj_ssp370_avg = oahu_pr_proj_ssp370.groupby('dayofyear').mean('time',_ →keep attrs=True) #retain attributes for metpy conversion in nxt step oahu_pr_proj_ssp370_avg [21]: <xarray.DataArray 'pr' (dayofyear: 31390, lat: 1, lon: 1)> dask.array<stack, shape=(31390, 1, 1), dtype=float32, chunksize=(1, 1, 1), chunktype=numpy.ndarray> Coordinates: * lat (lat) float64 21.71 (lon) float64 202.5 * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31' Attributes: area: areacella cell measures: cell_methods: area: time: mean description: at surface; includes both liquid and solid phases. ... history: none interval_operation: 900 s interval_write: 3 h long_name: Precipitation online_operation: average

cv=6.2.3.0-7-g2019642

CMIP6_CV_version:

```
units:
                               kg m-2 s-1
[22]: ## daily sum of precip ssp3-7.0 projection
      oahu_pr_proj_sum_370 = oahu_pr_proj_ssp370.groupby('dayofyear').sum('time',_
       →keep_attrs=True) #retain attributes for metpy conversion in nxt step
      oahu pr proj sum 370
[22]: <xarray.DataArray 'pr' (dayofyear: 31390, lat: 1, lon: 1)>
      dask.array<stack, shape=(31390, 1, 1), dtype=float32, chunksize=(1, 1, 1),
      chunktype=numpy.ndarray>
      Coordinates:
        * lat
                     (lat) float64 21.71
        * lon
                     (lon) float64 202.5
        * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
      Attributes:
          cell measures:
                               area: areacella
          cell_methods:
                               area: time: mean
                               at surface; includes both liquid and solid phases. ...
          description:
          history:
                               none
          interval_operation: 900 s
          interval_write:
                               3 h
          long_name:
                               Precipitation
          online_operation:
                               average
          standard_name:
                               precipitation_flux
          units:
                               kg m-2 s-1
[23]: # Make metpy recognize the units
      oahu_pr_proj_ssp370_sum = oahu_pr_proj_sum_370.metpy.quantify()
      # convert kg/m2/sec to in/day
      density water = units('kg / m^3') * 1000
      oahu_pr_proj_ssp370_converted_int_sum = (oahu_pr_proj_ssp370_sum /_
      →density_water)
      oahu_pr_proj_ssp370_converted_int_sum = oahu_pr_proj_ssp370_converted_int_sum.
       →metpy.convert_units('inches / day')
      oahu_pr_proj_ssp370_converted_int_sum = oahu_pr_proj_ssp370_converted_int_sum.
       →mean("lon").mean("lat")
      oahu_pr_proj_ssp370_converted_int_sum
[23]: <xarray.DataArray 'pr' (dayofyear: 31390)>
      <Quantity(dask.array<mean agg-aggregate, shape=(31390,), dtype=float32,</pre>
      chunksize=(1,), chunktype=numpy.ndarray>, 'inch / day')>
      Coordinates:
        * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
```

precipitation_flux

standard_name:

```
[16]: # Cast our xarray to dataframe -- daily avg
     oahu_pr_proj_ssp370_df = oahu_pr_proj_ssp370_converted_int.to_dataframe().
      →reset_index()
     oahu_pr_proj_ssp370_df.head(20)
[16]:
          dayofyear
                               pr
     0
         2015-01-01 3.701659e-04
     1
         2015-01-02 2.282345e-03
     2
         2015-01-03 1.025546e-02
     3
         2015-01-04 3.492512e-01
         2015-01-05 9.594570e-01
     4
         2015-01-06 2.072181e-01
     6
         2015-01-07 1.574929e-01
     7
         2015-01-08 5.034250e-02
         2015-01-09 4.010621e-02
     8
     9
         2015-01-10 2.340790e-02
     10 2015-01-11 3.337233e-03
         2015-01-12 5.087506e-03
     11
     12 2015-01-13 4.717490e-02
     13 2015-01-14 1.209491e+00
     14 2015-01-15 9.867913e-03
     15 2015-01-16 1.273414e-01
     16 2015-01-17 1.849313e+00
     17 2015-01-18 4.772992e-03
     18 2015-01-19 5.384365e-22
     19 2015-01-20 1.341719e-05
[24]: # Cast our xarray to dataframe -- daily sum
     oahu_pr_proj_ssp370_df_sum = oahu_pr_proj_ssp370_converted_int_sum.
      →to_dataframe().reset_index()
     oahu_pr_proj_ssp370_df_sum.head(20)
[24]:
          dayofyear
         2015-01-01 2.961327e-03
         2015-01-02 1.825876e-02
     1
     2
         2015-01-03 8.204365e-02
     3
         2015-01-04 2.794009e+00
     4
         2015-01-05 7.675656e+00
     5
         2015-01-06 1.657745e+00
         2015-01-07 1.259943e+00
     7
         2015-01-08 4.027400e-01
     8
         2015-01-09 3.208497e-01
     9
         2015-01-10 1.872632e-01
     10 2015-01-11 2.669786e-02
     11 2015-01-12 4.070005e-02
     12 2015-01-13 3.773992e-01
     13 2015-01-14 9.675924e+00
```

```
14 2015-01-15 7.894330e-02
      15 2015-01-16 1.018731e+00
      16 2015-01-17 1.479451e+01
      17 2015-01-18 3.818394e-02
      18 2015-01-19 4.307492e-21
      19 2015-01-20 1.073375e-04
[23]: ## to export df
      # oahu pr proj ssp370 df.to csv('oahu ssp370 2015 2100.csv', index = False)
[25]: ## to export df, daily total ssp370
      oahu_pr_proj_ssp370_df_sum.to_csv('oahu_ssp370_2015_2100_total.csv', index = __
       →False)
     1.1.5 Exploring CNRM-ESM2-1, ssp5-8.5 Projection
[26]: ## querty for 3hr, precipitaion for ssp 5-8.5 projection from CNRM-ESM2-1
      df_3hr_ssp585_CNRM_pr = df[(df.table_id == '3hr') & (df.variable_id == 'pr') &_u

→ (df.experiment_id== 'ssp585') & (df.source_id== 'CNRM-ESM2-1') ]
      len(df_3hr_ssp585_CNRM_pr)
      df_3hr_ssp585_CNRM_pr
[26]:
            activity_id institution_id
                                           source_id experiment_id member_id \
      69200 ScenarioMIP
                          CNRM-CERFACS CNRM-ESM2-1
                                                           ssp585 r1i1p1f2
            table_id variable_id grid_label \
      69200
                 3hr
                             pr
                                        gr
                                                       zstore dcpp_init_year \
      69200 gs://cmip6/CMIP6/ScenarioMIP/CNRM-CERFACS/CNRM...
                                                                        NaN
             version
      69200 20190328
[27]: ## pull data
      # get the path to a specific zarr store (the first one from the dataframe above)
      zstore3 = df_3hr_ssp585_CNRM_pr.zstore.values[-1]
      print(zstore3)
      # create a mutable-mapping-style interface to the store
      mapper3 = fsspec.get_mapper(zstore3)
      # open it using xarray and zarr
      ds_proj_ssp585 = xr.open_zarr(mapper3, consolidated=True)
      ds_proj_ssp585
```

gs://cmip6/CMIP6/ScenarioMIP/CNRM-CERFACS/CNRM-

ESM2-1/ssp585/r1i1p1f2/3hr/pr/gr/v20190328/

[27]: <xarray.Dataset>
Dimensions:

```
(lat: 128, lon: 256, time: 251288, axis_nbounds: 2)
      Coordinates:
                       (lat) float64 -88.93 -87.54 -86.14 -84.74 ... 86.14 87.54 88.93
        * lat
                       (lon) float64 0.0 1.406 2.812 4.219 ... 354.4 355.8 357.2 358.6
        * lon
                       (time) datetime64[ns] 2015-01-01T01:30:00 ... 2100-12-31T22:...
        * time
          time bounds (time, axis_nbounds) datetime64[ns]
      dask.array<chunksize=(62822, 1), meta=np.ndarray>
      Dimensions without coordinates: axis_nbounds
      Data variables:
                       (time, lat, lon) float32 dask.array<chunksize=(600, 128, 256),
          pr
     meta=np.ndarray>
      Attributes: (12/55)
          CMIP6_CV_version:
                                  cv=6.2.3.0-7-g2019642
          Conventions:
                                  CF-1.7 CMIP-6.2
          EXPID:
                                  CNRM-ESM2-1_ssp585_r1i1p1f2
                                  ScenarioMIP
          activity_id:
          arpege_minor_version:
                                  6.3.2
          branch_method:
                                  standard
          variable_id:
                                  pr
                                  r1i1p1f2
          variant_label:
          xios_commit:
                                  1442-shuffle
                                  2019-08-26; created; by nhn2@columbia.edu
          status:
          netcdf_tracking_ids:
                                  hdl:21.14100/6fb366f9-6ed1-47fe-918c-08fa5ca8baa3...
          version id:
                                  v20190328
[28]: # Apply lat/lon masks to the field, then calculate averages over the lat and__
      → lon dimensions
      oahu pr proj ssp585-ds proj ssp585.pr.where(mask lon & mask lat, drop = True)
      ## remove times associated with leap years (remove feb 29 from all recordedu
       \rightarrow years )
      oahu_pr_proj_ssp585 = oahu_pr_proj_ssp585.sel(time=~((oahu_pr_proj_ssp585.time.
       →dt.month == 2) & (oahu_pr_proj_ssp585.time.dt.day == 29)))
      oahu_pr_proj_ssp585
      ## group by day of year and avg by day
      oahu_pr_proj_ssp585['dayofyear'] = xr.DataArray(oahu_pr_proj_ssp585.
      →indexes['time'].strftime('%Y-%m-%d'), coords=oahu_pr_proj_ssp585.time.coords)
      oahu_pr_proj_ssp585_avg = oahu_pr_proj_ssp585.groupby('dayofyear').mean('time',_
       →keep attrs=True) #retain attributes for metpy conversion in nxt step
      oahu_pr_proj_ssp585_avg
```

```
[28]: <xarray.DataArray 'pr' (dayofyear: 31390, lat: 1, lon: 1)>
      dask.array<stack, shape=(31390, 1, 1), dtype=float32, chunksize=(1, 1, 1),
      chunktype=numpy.ndarray>
      Coordinates:
                     (lat) float64 21.71
        * lat
        * lon
                     (lon) float64 202.5
        * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
      Attributes:
          cell_measures:
                               area: areacella
          cell_methods:
                               area: time: mean
                               at surface; includes both liquid and solid phases. ...
          description:
          history:
                               none
          interval_operation: 900 s
          interval_write:
                               3 h
          long_name:
                               Precipitation
          online_operation:
                              average
          standard_name:
                               precipitation_flux
          units:
                               kg m-2 s-1
[29]: ## daily sum of precip ssp5-8.5 Projection
      oahu_pr_proj_sum_585 = oahu_pr_proj_ssp585.groupby('dayofyear').sum('time',_
       →keep attrs=True) #retain attributes for metpy conversion in nxt step
      oahu_pr_proj_sum_585
[29]: <xarray.DataArray 'pr' (dayofyear: 31390, lat: 1, lon: 1)>
      dask.array<stack, shape=(31390, 1, 1), dtype=float32, chunksize=(1, 1, 1),
      chunktype=numpy.ndarray>
      Coordinates:
        * lat
                     (lat) float64 21.71
                     (lon) float64 202.5
        * lon
        * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
      Attributes:
          cell_measures:
                               area: areacella
          cell_methods:
                               area: time: mean
          description:
                               at surface; includes both liquid and solid phases. ...
          history:
                               none
          interval_operation: 900 s
          interval_write:
                               3 h
          long name:
                               Precipitation
          online_operation:
                              average
          standard_name:
                               precipitation_flux
          units:
                               kg m-2 s-1
[30]: # Make metpy recognize the units
      oahu_pr_proj_ssp585_sum = oahu_pr_proj_sum_585.metpy.quantify()
      # convert kg/m2/sec to in/day
```

```
density_water = units('kg / m^3') * 1000
     oahu_pr_proj_ssp585_converted_int_sum = (oahu_pr_proj_ssp585_sum /_
      →density_water)
     oahu_pr_proj_ssp585_converted_int_sum = oahu_pr_proj_ssp585_converted_int_sum.
      →metpy.convert_units('inches / day')
     oahu_pr_proj ssp585_converted_int_sum = oahu_pr_proj_ssp585_converted_int_sum.
      →mean("lon").mean("lat")
     oahu_pr_proj_ssp585_converted_int_sum
[30]: <xarray.DataArray 'pr' (dayofyear: 31390)>
     <Quantity(dask.array<mean_agg-aggregate, shape=(31390,), dtype=float32,
     chunksize=(1,), chunktype=numpy.ndarray>, 'inch / day')>
     Coordinates:
       * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
[29]: # Cast our xarray to dataframe
     oahu_pr_proj_ssp585_df = oahu_pr_proj_ssp585_converted_int.to_dataframe().
      →reset_index()
     oahu_pr_proj_ssp585_df.head(20)
[29]:
          dayofyear
         2015-01-01 0.000311
     0
     1
         2015-01-02 0.001061
         2015-01-03 0.019240
     3
         2015-01-04 0.289843
     4
         2015-01-05 0.766865
     5
         2015-01-06 0.173856
     6
         2015-01-07 0.088689
     7
         2015-01-08 0.044970
         2015-01-09 0.023198
         2015-01-10 0.016240
     9
     10 2015-01-11 0.005077
     11 2015-01-12 0.207648
     12 2015-01-13 0.277482
     13 2015-01-14 0.007074
     14 2015-01-15 0.011796
     15 2015-01-16 0.483247
     16 2015-01-17 0.000010
     17 2015-01-18 0.009342
     18 2015-01-19 0.069012
     19 2015-01-20 0.005040
[31]: # Cast our xarray to dataframe
     oahu_pr_proj ssp585_df_sum = oahu_pr_proj ssp585_converted_int_sum.
      →to_dataframe().reset_index()
     oahu_pr_proj_ssp585_df_sum.head(20)
```

```
[31]:
         dayofyear
                           pr
         2015-01-01 0.002490
     0
         2015-01-02 0.008486
     1
     2
         2015-01-03 0.153917
         2015-01-04 2.318743
     3
     4
         2015-01-05 6.134923
     5
         2015-01-06 1.390850
     6
         2015-01-07 0.709511
     7
         2015-01-08 0.359761
     8
         2015-01-09 0.185582
         2015-01-10 0.129921
     9
     10 2015-01-11 0.040615
     11 2015-01-12 1.661184
     12 2015-01-13 2.219857
     13 2015-01-14 0.056589
     14 2015-01-15 0.094369
     15 2015-01-16 3.865978
     16 2015-01-17 0.000081
     17 2015-01-18 0.074732
     18 2015-01-19 0.552095
     19 2015-01-20 0.040323
[30]: | ## to export df
      # oahu_pr_proj_ssp585_df.to_csv('oahu_ssp585_2015_2100.csv', index = False)
[32]: ## to export df, daily total ssp585
     oahu pr proj ssp585 df sum.to csv('oahu ssp585 2015 2100 total.csv', index = 1
      →False)
     1.1.6 Exploring CNRM-ESM2-1, ssp1-2.6 Projection
[33]: | ## querty for 3hr, precipitaion for ssp 1-2.6 from CNRM-ESM2-1
     df 3hr ssp126 CNRM pr = df[(df.table id == '3hr') & (df.variable id == 'pr') & (
      → (df.experiment_id== 'ssp126') & (df.source_id== 'CNRM-ESM2-1') ]
     len(df_3hr_ssp126_CNRM_pr)
     df_3hr_ssp126_CNRM_pr
[33]:
            activity_id institution_id
                                          source_id experiment_id member_id \
     69045 ScenarioMIP
                         CNRM-CERFACS CNRM-ESM2-1
                                                          ssp126 r1i1p1f2
           table id variable id grid label \
     69045
                3hr
                             pr
                                        gr
                                                       zstore dcpp_init_year \
     69045 gs://cmip6/CMIP6/ScenarioMIP/CNRM-CERFACS/CNRM...
                                                                       NaN
             version
```

69045 20190328

```
[34]: ## pull data
      # get the path to a specific zarr store (the first one from the dataframe above)
      zstore4 = df_3hr_ssp126_CNRM_pr.zstore.values[-1]
      print(zstore4)
      # create a mutable-mapping-style interface to the store
      mapper4 = fsspec.get_mapper(zstore4)
      # open it using xarray and zarr
      ds_proj_ssp126 = xr.open_zarr(mapper4, consolidated=True)
      ds_proj_ssp126
     gs://cmip6/CMIP6/ScenarioMIP/CNRM-CERFACS/CNRM-
     ESM2-1/ssp126/r1i1p1f2/3hr/pr/gr/v20190328/
[34]: <xarray.Dataset>
      Dimensions:
                       (lat: 128, lon: 256, time: 251288, axis_nbounds: 2)
      Coordinates:
        * lat
                       (lat) float64 -88.93 -87.54 -86.14 -84.74 ... 86.14 87.54 88.93
                       (lon) float64 0.0 1.406 2.812 4.219 ... 354.4 355.8 357.2 358.6
        * lon
                       (time) datetime64[ns] 2015-01-01T01:30:00 ... 2100-12-31T22:...
        * time
          time_bounds (time, axis_nbounds) datetime64[ns]
      dask.array<chunksize=(62822, 1), meta=np.ndarray>
      Dimensions without coordinates: axis_nbounds
      Data variables:
          pr
                       (time, lat, lon) float32 dask.array<chunksize=(449, 128, 256),
     meta=np.ndarray>
      Attributes: (12/55)
          CMIP6_CV_version:
                                  cv=6.2.3.0-7-g2019642
          Conventions:
                                  CF-1.7 CMIP-6.2
                                  CNRM-ESM2-1_ssp126_r1i1p1f2
          EXPID:
                                  ScenarioMIP
          activity_id:
          arpege_minor_version:
                                  6.3.2
                                  standard
          branch_method:
          variable_id:
                                  pr
          variant_label:
                                  r1i1p1f2
          xios commit:
                                  1442-shuffle
          status:
                                  2019-11-03; created; by nhn2@columbia.edu
          netcdf tracking ids:
                                  hdl:21.14100/6255501d-a196-47b5-be0f-7d61a687e6e1...
          version_id:
                                  v20190328
[35]: # Apply lat/lon masks to the field, then calculate averages over the lat and
       \rightarrow lon dimensions
      oahu_pr_proj_ssp126=ds_proj_ssp126.pr.where(mask_lon & mask_lat, drop = True)
```

```
## remove times associated with leap years (remove feb 29 from records)
      oahu_pr_proj_ssp126 = oahu_pr_proj_ssp126.sel(time=~((oahu_pr_proj_ssp126.time.
       →dt.month == 2) & (oahu_pr_proj_ssp126.time.dt.day == 29)))
      oahu_pr_proj_ssp126
      ## group by day of year and avg by day
      oahu_pr_proj_ssp126['dayofyear'] = xr.DataArray(oahu_pr_proj_ssp126.
       →indexes['time'].strftime('%Y-%m-%d'), coords=oahu_pr_proj_ssp126.time.coords)
      oahu_pr_proj_ssp126_avg = oahu_pr_proj_ssp126.groupby('dayofyear').mean('time',__
       →keep_attrs=True) #retain attributes for metpy conversion in nxt step
      oahu_pr_proj_ssp126_avg
[35]: <xarray.DataArray 'pr' (dayofyear: 31390, lat: 1, lon: 1)>
      dask.array<stack, shape=(31390, 1, 1), dtype=float32, chunksize=(1, 1, 1),
      chunktype=numpy.ndarray>
      Coordinates:
        * lat
                     (lat) float64 21.71
                     (lon) float64 202.5
        * lon
        * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
      Attributes:
                               area: areacella
          cell_measures:
          cell_methods:
                               area: time: mean
                               at surface; includes both liquid and solid phases. \dots
          description:
          history:
          interval_operation: 900 s
          interval write:
                               3 h
          long_name:
                               Precipitation
          online operation:
                              average
          standard name:
                               precipitation_flux
          units:
                               kg m-2 s-1
[36]: ## daily sum of precip ssp126
      oahu_pr_proj_sum_126 = oahu_pr_proj_ssp126.groupby('dayofyear').sum('time',_
       →keep_attrs=True) #retain attributes for metpy conversion in nxt step
      oahu_pr_proj_sum_126
[36]: <xarray.DataArray 'pr' (dayofyear: 31390, lat: 1, lon: 1)>
      dask.array<stack, shape=(31390, 1, 1), dtype=float32, chunksize=(1, 1, 1),
      chunktype=numpy.ndarray>
      Coordinates:
                     (lat) float64 21.71
        * lat
                     (lon) float64 202.5
        * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
      Attributes:
          cell measures:
                               area: areacella
          cell methods:
                               area: time: mean
```

```
history:
                               none
          interval_operation:
                              900 s
          interval_write:
                               3 h
          long_name:
                               Precipitation
          online_operation:
                              average
          standard name:
                               precipitation_flux
          units:
                               kg m-2 s-1
[37]: # Make metpy recognize the units
      oahu pr proj ssp126 sum = oahu pr proj sum 126.metpy.quantify()
      # convert kg/m2/sec to in/day
      density_water = units('kg / m^3') * 1000
      oahu_pr_proj_ssp126_converted_int_sum = (oahu_pr_proj_ssp126_sum /_
      →density_water)
      oahu_pr_proj ssp126_converted_int_sum = oahu_pr_proj ssp126_converted_int_sum.
      →metpy.convert_units('inches / day')
      oahu_pr_proj_ssp126_converted_int_sum = oahu_pr_proj_ssp126_converted_int_sum.
      →mean("lon").mean("lat")
      oahu_pr_proj_ssp126_converted_int_sum
[37]: <xarray.DataArray 'pr' (dayofyear: 31390)>
      <Quantity(dask.array<mean agg-aggregate, shape=(31390,), dtype=float32,</pre>
      chunksize=(1,), chunktype=numpy.ndarray>, 'inch / day')>
      Coordinates:
        * dayofyear (dayofyear) object '2015-01-01' '2015-01-02' ... '2100-12-31'
[14]: # Cast our xarray to dataframe -- daily average
      oahu_pr_proj_ssp126_df = oahu_pr_proj_ssp126_converted_int.to_dataframe().
      →reset_index()
      oahu_pr_proj_ssp126_df.head(20)
[14]:
          dayofyear
                            pr
          2015-01-01 0.001504
      0
      1
         2015-01-02 0.001572
      2
          2015-01-03 0.015226
      3
          2015-01-04 0.296919
      4
          2015-01-05 0.762305
          2015-01-06 0.135178
      5
      6
         2015-01-07 0.150985
      7
          2015-01-08 0.028382
      8
          2015-01-09 0.016181
          2015-01-10 0.011141
      9
      10 2015-01-11 0.001719
      11 2015-01-12 0.156439
      12 2015-01-13 0.533256
```

at surface; includes both liquid and solid phases. ...

description:

```
13 2015-01-14 0.029845
     14 2015-01-15 0.002761
     15 2015-01-16 1.747228
     16 2015-01-17 0.000391
     17 2015-01-18 0.092346
     18 2015-01-19 0.001736
     19 2015-01-20 0.000678
[38]: # Cast our xarray to dataframe -- daily sum
     oahu_pr_proj_ssp126_df_sum = oahu_pr_proj_ssp126_converted_int_sum.
      →to_dataframe().reset_index()
     oahu_pr_proj_ssp126_df_sum.head(20)
[38]:
          dayofyear
                           pr
         2015-01-01
                      0.012032
     0
     1
         2015-01-02 0.012579
     2
         2015-01-03 0.121809
     3
         2015-01-04 2.375351
         2015-01-05 6.098441
     4
     5
         2015-01-06 1.081423
     6
         2015-01-07 1.207878
     7
         2015-01-08
                     0.227057
         2015-01-09 0.129448
     8
     9
         2015-01-10 0.089130
     10 2015-01-11 0.013749
     11 2015-01-12 1.251516
     12 2015-01-13 4.266051
     13 2015-01-14 0.238757
     14 2015-01-15
                    0.022086
     15 2015-01-16 13.977824
     16 2015-01-17 0.003131
     17 2015-01-18
                      0.738765
     18 2015-01-19
                      0.013890
     19 2015-01-20
                      0.005426
[15]: ## to export df
     \#oahu\_pr\_proj\_ssp126\_df.to\_csv('oahu\_ssp126\_2015\_2100.csv', index = False)
[39]: ## to export df, daily total ssp1-2.6 Projection
     oahu_pr_proj_ssp126_df_sum.to_csv('oahu_ssp126_2015_2100_total.csv', index =__
      →False)
```

1.1.7 Exploring CNRM-ESM2-1, historical

```
[42]: | ## querty for 3hr, precipitaion for historical from CNRM-ESM2-1
      df_3hr_historical_CNRM_pr = df[(df.table_id == '3hr') & (df.variable_id ==_
      →'pr') & (df.experiment_id== 'historical') & (df.source_id== 'CNRM-ESM2-1') ]
      len(df_3hr_historical_CNRM_pr)
      df_3hr_historical_CNRM_pr
[42]:
            activity_id institution_id
                                          source_id experiment_id member_id \
      44063
                   CMIP
                          CNRM-CERFACS CNRM-ESM2-1
                                                       historical r1i1p1f2
            table_id variable_id grid_label \
      44063
                 3hr
                              pr
                                         gr
                                                         zstore dcpp init year \
      44063 gs://cmip6/CMIP6/CMIP/CNRM-CERFACS/CNRM-ESM2-1...
                                                                          NaN
              version
      44063 20181206
[43]: ## pull data
      # get the path to a specific zarr store (the first one from the dataframe above)
      zstore5 = df_3hr_historical_CNRM_pr.zstore.values[-1]
      print(zstore5)
      # create a mutable-mapping-style interface to the store
      mapper5 = fsspec.get_mapper(zstore5)
      # open it using xarray and zarr
      ds_proj_historical = xr.open_zarr(mapper5, consolidated=True)
      ds_proj_historical
     gs://cmip6/CMIP6/CMIP/CNRM-CERFACS/CNRM-
     ESM2-1/historical/r1i1p1f2/3hr/pr/gr/v20181206/
[43]: <xarray.Dataset>
     Dimensions:
                       (lat: 128, lon: 256, time: 482120, axis_nbounds: 2)
      Coordinates:
        * lat
                       (lat) float64 -88.93 -87.54 -86.14 -84.74 ... 86.14 87.54 88.93
                       (lon) float64 0.0 1.406 2.812 4.219 ... 354.4 355.8 357.2 358.6
        * lon
        * time
                       (time) datetime64[ns] 1850-01-01T01:30:00 ... 2014-12-31T22:...
          time_bounds (time, axis_nbounds) datetime64[ns]
      dask.array<chunksize=(60265, 1), meta=np.ndarray>
      Dimensions without coordinates: axis_nbounds
      Data variables:
                       (time, lat, lon) float32 dask.array<chunksize=(600, 128, 256),
          pr
      meta=np.ndarray>
      Attributes: (12/55)
```

```
CNRM-ESM2-1_historical_r1i1p1f2_v2
          activity_id:
                                  6.3.2
          arpege_minor_version:
          branch_method:
                                  standard
          variable_id:
          variant label:
                                  r1i1p1f2
          xios commit:
                                  1442-shuffle
                                  2019-10-25; created; by nhn2@columbia.edu
          status:
          netcdf_tracking_ids:
                                  hdl:21.14100/f1e5c10f-c895-46b1-a771-05e33c7947b6...
          version id:
                                  v20181206
[44]: # Apply lat/lon masks to the field, then calculate averages over the lat and
       \rightarrow lon dimensions
      oahu_pr_proj_historical=ds_proj_historical.pr.where(mask_lon & mask_lat, drop = __
       →True)
      ## remove times associated with leap years (remove feb 29 from records)
      oahu pr proj historical = oahu pr proj historical.
       ⇒sel(time=~((oahu_pr_proj_historical.time.dt.month == 2) &
       →(oahu pr proj historical.time.dt.day == 29)))
      oahu_pr_proj_historical
      ## group by day of year and avg by day
      oahu pr proj historical['dayofyear'] = xr.DataArray(oahu pr proj historical.
       →indexes['time'].strftime('%Y-%m-%d'), coords=oahu_pr_proj_historical.time.
       →coords)
      oahu_pr_proj_historical_avg = oahu_pr_proj_historical.groupby('dayofyear').
       →mean('time', keep_attrs=True) #retain attributes for metry conversion in nxt_
       \hookrightarrowstep
      oahu_pr_proj_historical_avg
[44]: <xarray.DataArray 'pr' (dayofyear: 60225, lat: 1, lon: 1)>
      dask.array<stack, shape=(60225, 1, 1), dtype=float32, chunksize=(1, 1, 1),
      chunktype=numpy.ndarray>
      Coordinates:
                     (lat) float64 21.71
        * lat
                     (lon) float64 202.5
        * dayofyear (dayofyear) object '1850-01-01' '1850-01-02' ... '2014-12-31'
      Attributes:
          cell measures:
                               area: areacella
          cell_methods:
                               area: time: mean
          description:
                               at surface; includes both liquid and solid phases. ...
          history:
                               none
          interval_operation: 900 s
```

cv=6.2.3.0-7-g2019642

CF-1.7 CMIP-6.2

CMIP6_CV_version:

Conventions:

EXPID:

```
online_operation:
                               average
          standard_name:
                               precipitation_flux
          units:
                               kg m-2 s-1
[45]: ## daily sum of precip historical
      oahu_pr_proj_sum_historical = oahu_pr_proj_historical.groupby('dayofyear').
       →sum('time', keep_attrs=True) #retain attributes for metry conversion in nxt⊔
      oahu_pr_proj_sum_historical
[45]: <xarray.DataArray 'pr' (dayofyear: 60225, lat: 1, lon: 1)>
      dask.array<stack, shape=(60225, 1, 1), dtype=float32, chunksize=(1, 1, 1),
      chunktype=numpy.ndarray>
      Coordinates:
        * lat
                     (lat) float64 21.71
        * lon
                     (lon) float64 202.5
        * dayofyear (dayofyear) object '1850-01-01' '1850-01-02' ... '2014-12-31'
      Attributes:
          cell_measures:
                               area: areacella
          cell_methods:
                               area: time: mean
          description:
                               at surface; includes both liquid and solid phases. ...
          history:
          interval_operation: 900 s
          interval write:
                               3 h
                               Precipitation
          long_name:
          online_operation:
                              average
          standard_name:
                               precipitation_flux
          units:
                               kg m-2 s-1
[47]: # Make metpy recognize the units
      oahu_pr_proj_historical_sum = oahu_pr_proj_sum_historical.metpy.quantify()
      # convert kg/m2/sec to in/day
      density_water = units('kg / m^3') * 1000
      oahu_pr_proj_historical_converted_int_sum = (oahu_pr_proj_historical_sum /_
       →density_water)
      oahu_pr_proj_historical_converted_int_sum =_
      →oahu_pr_proj_historical_converted_int_sum.metpy.convert_units('inches / day')
      oahu pr proj historical converted int sum = 11
       →oahu_pr_proj_historical_converted_int_sum.mean("lon").mean("lat")
      oahu_pr_proj_historical_converted_int_sum
[47]: <xarray.DataArray 'pr' (dayofyear: 60225)>
      <Quantity(dask.array<mean agg-aggregate, shape=(60225,), dtype=float32,</pre>
      chunksize=(1,), chunktype=numpy.ndarray>, 'inch / day')>
```

interval_write:

long_name:

3 h

Precipitation

```
Coordinates:
```

[22]: # Cast our xarray to dataframe -- daily avq

```
* dayofyear (dayofyear) object '1850-01-01' '1850-01-02' ... '2014-12-31'
```

```
oahu_pr_proj_historical_df = oahu_pr_proj_historical_converted_int.
      →to_dataframe().reset_index()
     oahu_pr_proj_historical_df.head(20)
[22]:
          dayofyear
                               pr
         1850-01-01 7.431174e-04
     1
         1850-01-02 5.950664e-02
         1850-01-03 3.129180e-01
     2
     3
         1850-01-04 4.294988e-04
     4
         1850-01-05 1.384215e-03
     5
         1850-01-06 4.975918e-03
     6
         1850-01-07 9.406500e-05
     7
         1850-01-08 9.380140e-23
     8
         1850-01-09 2.022306e-05
         1850-01-10 2.199838e-22
     9
     10 1850-01-11 5.801964e-04
     11 1850-01-12 1.061206e-01
     12 1850-01-13 4.339770e-02
     13 1850-01-14 2.547624e-02
     14 1850-01-15 2.540510e-01
     15 1850-01-16 1.818833e-05
     16 1850-01-17 2.309185e-03
     17 1850-01-18 1.173254e-01
     18 1850-01-19 1.011662e-01
     19 1850-01-20 7.681608e-05
[48]: # Cast our xarray to dataframe -- daily sum historical
     oahu_pr_proj_historical_df_sum = oahu_pr_proj_historical_converted_int_sum.
      →to_dataframe().reset_index()
     oahu_pr_proj_historical_df_sum.head(20)
```

```
[48]:
          dayofyear
         1850-01-01 5.944939e-03
     1
         1850-01-02 4.760531e-01
     2
         1850-01-03 2.503344e+00
     3
         1850-01-04 3.435990e-03
     4
         1850-01-05 1.107372e-02
     5
         1850-01-06 3.980734e-02
     6
         1850-01-07 7.525200e-04
     7
         1850-01-08 7.504112e-22
     8
         1850-01-09 1.617845e-04
         1850-01-10 1.759870e-21
     10 1850-01-11 4.641572e-03
```

```
11 1850-01-12 8.489646e-01

12 1850-01-13 3.471816e-01

13 1850-01-14 2.038099e-01

14 1850-01-15 2.032408e+00

15 1850-01-16 1.455066e-04

16 1850-01-17 1.847348e-02

17 1850-01-18 9.386032e-01

18 1850-01-19 8.093297e-01

19 1850-01-20 6.145286e-04
```

[23]: ## to export df
#oahu_pr_proj_historical_df.to_csv('oahu_historical_2015_2100.csv', index =□
→False)

[49]: ## to export df, daily total historical oahu_pr_proj_historical_df_sum.to_csv('oahu_historical_1850_2014_total.csv', ⊔ ⇒index = False)