

## 2-1\_compute\_delta\_T\_sensitivity

February 14, 2020

### 1 Plot temperature response over time

This notebook does the same as [2\\_compute\\_delta\\_T.ipynb](#) except that it varies the ECS parameter and outputs a table of changes in temperature with respect to some reference year (defined below).

```
[22]: from ar6_ch6_rcmipfigs.constants import BASE_DIR
      from ar6_ch6_rcmipfigs.constants import OUTPUT_DATA_DIR, INPUT_DATA_DIR,
      ↪RESULTS_DIR

PATH_DATASET = OUTPUT_DATA_DIR + '/forcing_data_rcmip_models.nc'
PATH_DT_OUTPUT = RESULTS_DIR + '/tables/table_sens_dT_cs.csv'
```

Output table found in:

```
[23]: print(PATH_DT_OUTPUT)

/home/sarambl/PHD/IPCC/public/AR6_CH6_RCMIPFIGS/ar6_ch6_rcmipfigs/results/tables
/table_sens_dT_cs.csv
```

#### 1.0.1 General about computing $\Delta T$ :

We compute the change in GSAT temperature ( $\Delta T$ ) from the effective radiative forcing (ERF) estimated from the RCMIP models (Nicholls et al 2020), by integrating with the impulse response function (IRF( $t-t'$ )) (Geoffroy et al 2013). See Nicholls et al (2020) for description of the RCMIP models and output.

For any forcing agent  $x$ , with estimated  $ERF_x$ , the change in temperature  $\Delta T$  is calculated as:

$$\Delta T_x(t) = \int_0^t ERF_x(t') IRF(t-t') dt'$$

**The Impulse response function (IRF):** In these calculations we use the impulse response function (Geoffroy et al 2013):

$$\text{IRF}(t) = 0.885 \cdot \left( \frac{0.587}{4.1} \cdot \exp\left(\frac{-t}{4.1}\right) + \frac{0.413}{249} \cdot \exp\left(\frac{-t}{249}\right) \right)$$
$$\text{IRF}(t) = \frac{1}{\lambda} \sum_{i=1}^2 \frac{a_i}{\tau_i} \cdot \exp\left(\frac{-t}{\tau_i}\right)$$

with  $\frac{1}{\lambda} = 0.885$  (K/Wm<sup>-2</sup>),  $a_1 = 0.587$ ,  $\tau_1 = 4.1$ (yr),  $a_2 = 0.413$  and  $\tau_2 = 249$ (yr) (note that  $i = 1$  is the fast response and  $i = 2$  is the slow response and that  $a_1 + a_2 = 1$ )

## 1.1 Input data:

See [README.md](#)

## 2 Code + figures

```
[24]: from ar6_ch6_rcmipfigs.constants import BASE_DIR
      from ar6_ch6_rcmipfigs.constants import OUTPUT_DATA_DIR, INPUT_DATA_DIR, \
      ↪RESULTS_DIR

      PATH_DATASET = OUTPUT_DATA_DIR + '/forcing_data_rcmip_models.nc'
      PATH_DT_OUTPUT = RESULTS_DIR + '/tables/table_sens_dT_cs.csv'
```

Output table found in:

```
[25]: print(PATH_DT_OUTPUT)

/home/sarambl/PHD/IPCC/public/AR6_CH6_RCMIPFIGS/ar6_ch6_rcmipfigs/results/tables
/table_sens_dT_cs.csv
```

## 2.1 Imports:

```
[26]: import xarray as xr
      from IPython.display import clear_output
      import numpy as np
      import os
      import re
      from pathlib import Path
      import pandas as pd
      import tqdm
      from scmdata import df_append, ScmDataFrame
      import matplotlib.pyplot as plt

      %load_ext autoreload
```

```
%autoreload 2
```

The autoreload extension is already loaded. To reload it, use:

```
%reload_ext autoreload
```

```
[27]: climatemodel = 'climatemodel'  
      scenario = 'scenario'  
      variable = 'variable'  
      time = 'time'
```

## 2.2 Set values:

ECS parameters:

```
[28]: ECS2ecsf = {'ECS = 2K':0.526, 'ECS = 3.4K':0.884, 'ECS = 5K': 1.136 }
```

Year to integrate from and to:

```
[29]: first_y = '1850'  
      last_y = '2100'
```

Set reference year for temperature change:

```
[30]: ref_year = '2021'
```

Years to output change in

```
[31]: years= ['2040', '2100']
```

## 2.3 IRF:

```
[32]: def IRF(t, l=0.885, alpha1=0.587 / 4.1, alpha2=0.413 / 249, tau1=4.1, tau2=249):  
      """  
      Returns the IRF function for:  
      :param t: Time in years  
      :param l: climate sensitivity factor  
      :param alpha1:  
      :param alpha2:  
      :param tau1:  
      :param tau2:  
      :return:  
      IRF  
      """  
      return l * (alpha1 * np.exp(-t / tau1) + alpha2 * np.exp(-t / tau2))
```

## 2.4 ERF:

Read ERF from file

### 2.4.1 Define variables to look at:

```
[33]: # variables to plot:
variables_erf_comp = [
    'Effective Radiative Forcing|Anthropogenic|CH4',
    'Effective Radiative Forcing|Anthropogenic|Aerosols',
    'Effective Radiative Forcing|Anthropogenic|Tropospheric Ozone',
    'Effective Radiative Forcing|Anthropogenic|F-Gases|HFC',
    'Effective Radiative Forcing|Anthropogenic|Other|BC on Snow']
# total ERFs for anthropogenic and total:
variables_erf_tot = ['Effective Radiative Forcing|Anthropogenic',
                    'Effective Radiative Forcing']
# Scenarios to plot:
scenarios_fl = ['ssp119', 'ssp126', 'ssp245', 'ssp370',
    ↪ 'ssp370-lowNTCF-aerchemmip', # 'ssp370-lowNTCF', Due to mistake here
    'ssp585', 'historical']
```

### 2.4.2 Open dataset:

```
[34]: ds = xr.open_dataset(PATH_DATASET)
```

## 2.5 Integrate:

The code below integrates the read in ERFs with the pre defined impulse response function (IRF).

$$\Delta T(t) = \int_0^t ERF(t')IRF(t-t')dt'$$

```
[35]: name_deltaT = 'Delta T'

def new_varname(var, nname):
    """
    var:str
        Old variable of format varname/bla/bla
    nname:str
        name for the resulting variable, based on var
    Returns
```

```

-----
new variable name with nname/bla/bla
"""
return nname + '|' + '|'.join(var.split('|')[1:])

def integrate_(i, var, nvar, ds, ds_DT, csfac=0.885):
    """
    Parameters
    -----
    i:int
        the index for the integral
    var:str
        the name of the EFR variables to integrate
    nvar:str
        the name of output integrated value

    ds:xr.Dataset
        the ds with the input data
    ds_DT: xr.Dataset
        the ouptut ds with the integrated results
    csfac: climate sensitivity factor (for IRF)
    Returns
    -----
    None

    """
    # lets create a ds that goes from 0 to i inclusive
    ds_short = ds[{'time': slice(0, i + 1)}].copy()
    # lets get the current year
    current_year = ds_short['time'][{'time': i}].dt.year
    # lets get a list of years
    years = ds_short['time'].dt.year
    # lets get the year delta until current year(i)
    ds_short['end_year_delta'] = current_year - years

    # lets get the irf values from 0 until i
    ds_short['irf'] = IRF(
        ds_short['end_year_delta'] * ds_short['delta_t'], l=csfac
    )

    # lets do the famous integral
    ds_short['to_integrate'] = \
        ds_short[var] * \
        ds_short['irf'] * \
        ds_short['delta_t']

```

```

# lets sum all the values up until i and set
# this value at ds_DT
# If whole array is null, set value to nan
if np.all(ds_short['to_integrate'].isnull()): # or last_null:
    _val = np.nan
else:
    #

    _ds_int = ds_short['to_integrate'].sum(['time'])
    # mask where last value is null (in order to not get integral
    _ds_m1 = ds_short['to_integrate'].isel(time=-1)
    # where no forcing data)
    _val = _ds_int.where(_ds_m1.notnull())
# set value in dataframe:
ds_DT[nvar][{'time': i}] = _val

def integrate_to_dT(ds, from_t, to_t, variables, csfac=0.885):
    """
    Integrate forcing to temperature change.

    :param ds: dataset containing the forcings
    :param from_t: start time
    :param to_t: end time
    :param variables: variables to integrate
    :param csfac: climate sensitivity factor
    :return:
    """
    # slice dataset
    ds_sl = ds.sel(time=slice(from_t, to_t))
    len_time = len(ds_sl['time'])
    # lets create a result DS
    ds_DT = ds_sl.copy()

    # lets define the vars of the ds
    vars = variables # variables_erf_comp+ variables_erf_tot #['EFR']
    for var in variables:
        namevar = new_varname(var, name_deltaT)
        # set all values to zero for results dataarray:
        ds_DT[namevar] = ds_DT[var] * 0
        # Units Kelvin:
        ds_DT[namevar].attrs['unit'] = 'K'
        if 'unit' in ds_DT[namevar].coords:
            ds_DT[namevar].coords['unit'] = 'K'

    for i in range(len_time):

```

```

        # da = ds[var]
        if (i % 20) == 0:
            print('%s of %s done' % (i, len_time))
        for var in variables:
            namevar = new_varname(var, name_deltaT) # 'Delta T/' + '/'.
            join(var.split('/')[1:])

            # print(var)
            integrate_(i, var, namevar, ds_sl, ds_DT, csfac=csfac)
        clear_output()

    fname = 'DT_%s-%s.nc' % (from_t, to_t)
    # save dataset.
    ds_DT.to_netcdf(fname)
    return ds_DT

```

## 2.6 Compute $\Delta T$ with 3 different climate sensitivities

```

[36]: dic_ds = {}
      for key in ECS2ecsf:
          dic_ds[key] = integrate_to_dT(ds, first_y,
            join(last_y, (variables_era_comp+variables_era_tot), csfac=ECS2ecsf[key])

```

## 2.7 Table

### 2.7.1 Setup table:

```

[37]: iterables = [list(ECS2ecsf.keys()), years]

def setup_table(scenario_n=''):
    _i = pd.MultiIndex.from_product(iterables, names=['', ''])
    table = pd.DataFrame(columns=[var.split('/')[-1] for var in
            join(variables_era_comp, index = _i).transpose()
            table.index.name=scenario_n
    return table

```

```

[51]: # Dictionary of tables with different ESC:
      scntab_dic = {}
      for scn in scenarios_fl:
          # Loop over scenarios
          tab = setup_table(scenario_n=scn) # make table
          for var in variables_era_comp:
              # Loop over variables
              tabvar = var.split('/')[-1]

```

```

dtvar = new_varname(var, name_deltaT)
for key in ECS2ecsf:
    # Loop over ESC parameters
    for year in years:
        _tab_da = dic_ds[key][dtvar].sel(scenario=scn,
        ↪time=slice(year,year))- dic_ds[key][dtvar].sel(scenario=scn,
        ↪time=slice(ref_year,ref_year)).squeeze()
        tab.loc[tabvar,key][year]=_tab_da.squeeze().
        ↪mean('climatemodel').values
    scntab_dic[scn]=tab.copy()

```

```

/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
    return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
    return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
    return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
    return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
    return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
    return np.nanmean(a, axis=axis, dtype=dtype)

```

```

[52]: from IPython.display import display

for key in scntab_dic:
    display(scntab_dic[key])

```

	ECS = 2K	\
	2040	2100
ssp119		
CH4	-0.02659828770048701	-0.1296912754447551
Aerosols	0.15646387941855683	0.2249195956185493
Tropospheric Ozone	-0.04992508208229368	-0.10048092297791049
HFC	0.003468253587250494	-0.0029761684916363817
BC on Snow	-0.011329101981495735	-0.01697901215240332

  

	ECS = 3.4K	\
	2040	2100
ssp119		



CH4	-0.04470130480462069	-0.2179602423824401
Aerosols	0.26295450457415265	0.37800175385322743
Tropospheric Ozone	-0.08390451057176351	-0.16886907968150738
HFC	0.005828775990740371	-0.005001773662750117
BC on Snow	-0.019039783558255186	-0.02853506985308847

ECS = 5K

2040

2100

ssp119

CH4	-0.05744421069915064	-0.2800937051430452
Aerosols	0.3379143859685943	0.4857579099290342
Tropospheric Ozone	-0.10782299096099926	-0.2170082290929777
HFC	0.0074903727663812964	-0.006427618643534087
BC on Snow	-0.024467414165359597	-0.03666950153066572

ECS = 2K

2040

2100

ssp126

CH4	-0.017367724889572122	-0.1231558283831311
Aerosols	0.12215597107108145	0.22774103852286798
Tropospheric Ozone	-0.033763352537327304	-0.09404727045262026
HFC	0.011032221399586888	-0.00132423401564886
BC on Snow	-0.009436533713336538	-0.01532641975381759

ECS = 3.4K

2040

2100

ssp126

CH4	-0.02918834373076379	-0.20697671538153598
Aerosols	0.2052963468190799	0.3827434943996489
Tropospheric Ozone	-0.0567429727053181	-0.15805662942987894
HFC	0.018540843568887466	-0.002225518763942189
BC on Snow	-0.0158591174954173	-0.025757709244058458

ECS = 5K

2040

2100

ssp126

CH4	-0.03750900280333449	-0.2659791274586253
Aerosols	0.2638197398037043	0.4918513683687795
Tropospheric Ozone	-0.07291857125932286	-0.20311349664292128
HFC	0.023826242414316925	-0.0028599426649754825
BC on Snow	-0.020380042392300957	-0.033100404639423535

ECS = 2K

2040

2100

ssp245

CH4	0.028937393882524302	-0.004182704112309194
Aerosols	0.03637900641138267	0.1367591718072312

Tropospheric Ozone	0.0009328311281484147	-0.039034660403500676
HFC	0.018234694544699594	0.038451376778719744
BC on Snow	-0.003402782920874044	-0.012488796472324282

ECS = 3.4K \

2040 2100

ssp245		
CH4	0.04863242622082048	-0.007029487519546129
Aerosols	0.06113886248604996	0.22983860813230494
Tropospheric Ozone	0.0015677237971163881	-0.06560197679979962
HFC	0.030645380185388666	0.06462170546081415
BC on Snow	-0.005718745441164742	-0.020988775820408107

ECS = 5K

2040 2100

ssp245		
CH4	0.06249596853716284	-0.009033368577154454
Aerosols	0.07856758799112298	0.2953582113555412
Tropospheric Ozone	0.002014631485887075	-0.08430299281060225
HFC	0.039381393541404436	0.08304327760575213
BC on Snow	-0.007348976042039757	-0.026972001506768793

ECS = 2K \

2040 2100

ssp370		
CH4	0.05457162376258964	0.21565799113002856
Aerosols	-0.011068885443324872	-0.007421477076560101
Tropospheric Ozone	0.024602751780866523	0.06349719049159727
HFC	0.01704878795819055	0.06493327819133399
BC on Snow	0.0036842324966297967	0.006754058664053657

ECS = 3.4K \

2040 2100

ssp370		
CH4	0.0917135273880785	0.36243662387632175
Aerosols	-0.018602461467488916	-0.012472596455663742
Tropospheric Ozone	0.041347590445410684	0.106713909495384
HFC	0.028652335655970435	0.10912741049646246
BC on Snow	0.006191751952510926	0.01135092748863771

ECS = 5K

2040 2100

ssp370		
CH4	0.11785810759372976	0.46575566145192465
Aerosols	-0.02390542559623008	-0.01602813300184842
Tropospheric Ozone	0.05313446011989424	0.13713461672709976
HFC	0.03682019604658643	0.14023612932577073

BC on Snow 0.007956821513633937 0.014586712247842122

	ECS = 2K	
	2040	2100
ssp370-lowNTCF-aerchemmip		
CH4	0.05457162366753363	0.21565799026157
Aerosols	0.06273291382163072	0.14244876622095556
Tropospheric Ozone	-0.0032062710739075963	0.012046389288997983
HFC	0.01704878793732071	0.0649332767375719
BC on Snow	-0.0037433274391409396	-0.00704176174526431

	ECS = 3.4K	
	2040	2100
ssp370-lowNTCF-aerchemmip		
CH4	0.09171352722832657	0.3624366224167831
Aerosols	0.10542945973064936	0.23940058809757553
Tropospheric Ozone	-0.005388485987327574	0.02024526260736541
HFC	0.0286523356208964	0.10912740805325775
BC on Snow	-0.006291067407225456	-0.011834443693562069

	ECS = 5K	
	2040	2100
ssp370-lowNTCF-aerchemmip		
CH4	0.11785810738843767	0.4657556595763185
Aerosols	0.13548401159956747	0.30764600461407887
Tropospheric Ozone	-0.006924570228058991	0.026016536563311144
HFC	0.03682019600151392	0.14023612618608688
BC on Snow	-0.008084448613810082	-0.015208063389011895

	ECS = 2K	
	2040	2100
ssp585		
CH4	0.044123353916243016	0.11489216246985463
Aerosols	0.047101148762349374	0.0979238452792276
Tropospheric Ozone	0.01531662537912159	0.011390724974023542
HFC	0.02710784418483706	0.14984300008060603
BC on Snow	-0.0018159626749880552	-0.0035326595202287737

	ECS = 3.4K	
	2040	2100
ssp585		
CH4	0.07415407768433238	0.19308872932196103
Aerosols	0.07915858461200927	0.16457163351109735
Tropospheric Ozone	0.025741248736014292	0.01914334767497496
HFC	0.04555766969466912	0.2518273993750109
BC on Snow	-0.0030519220621472187	-0.005937017140460517

	ECS = 5K	
	2040	2100
ssp585		
CH4	0.09529302290656273	0.24813212274858323
Aerosols	0.10172415398104348	0.21148571908213415
Tropospheric Ozone	0.03307925176935772	0.024600501084583216
HFC	0.0585446977071766	0.3236153005543126
BC on Snow	-0.003921926993890548	-0.0076294699904560365

	ECS = 2K	
	2040	2100
historical		
CH4	0.0077132387149825565	0.02703927122654534
Aerosols	-0.009908627912094015	-0.04251408931871051
Tropospheric Ozone	0.005241636725860588	0.01849257773823193
HFC	0.0007444353141619498	0.0017742750627926067
BC on Snow	nan	nan

	ECS = 3.4K	
	2040	2100
historical		
CH4	0.012962933505788199	0.045442425407349996
Aerosols	-0.016652522954926197	-0.0714495341401904
Tropospheric Ozone	0.008809138527872185	0.031078780837636932
HFC	0.0012511042161961299	0.0029818615123738887
BC on Snow	nan	nan

	ECS = 5K	
	2040	2100
historical		
CH4	0.016658249392053515	0.058396600975961
Aerosols	-0.02139962225881925	-0.09181750088603652
Tropospheric Ozone	0.011320340913645699	0.03993834279587727
HFC	0.0016077538343877845	0.0038318944321908757
BC on Snow	nan	nan

## 2.7.2 Make table with all scenarios:

```
[40]: iterables = [list(ECS2ecsf.keys()), years]
iterables2 = [scenarios_fl, [var.split('|')[-1] for var in variables_erf_comp]]

def setup_table2():#scenario_n='':
    _i = pd.MultiIndex.from_product(iterables, names=['', ''])
    _r = pd.MultiIndex.from_product(iterables2, names=['', ''])
```

```

table = pd.DataFrame(columns=_r, index = _i).transpose()
return table

```

```

[55]: tab = setup_table2()#scenario_n=scn

for scn in scenarios_fl:
    for var in variables_erf_comp:
        tabvar = var.split('|')[-1]
        dtvar = new_varname(var,name_deltaT)
        print(dtvar)
        for key in ECS2ecsf:
            for year in years:
                # compute difference between year and ref year
                _da_y = dic_ds[key][dtvar].sel(scenario=scn,␣
↪time=slice(year,year))#.squeeze()
                _da_refy = dic_ds[key][dtvar].sel(scenario=scn,␣
↪time=slice(ref_year,ref_year)).squeeze()
                #_tab_da = dic_ds[key][dtvar].sel(scenario=scn,␣
↪time=slice(year,year))- dic_ds[key][dtvar].sel(scenario=scn,␣
↪time=slice(ref_year,ref_year)).squeeze()
                _tab_da = _da_y - _da_refy

                tab.loc[(scn, tabvar), (key,year)] =_tab_da.squeeze().
↪mean('climatemodel').values#[0]

```

```

Delta T|Anthropogenic|CH4
Delta T|Anthropogenic|Aerosols
Delta T|Anthropogenic|Tropospheric Ozone
Delta T|Anthropogenic|F-Gases|HFC
Delta T|Anthropogenic|Other|BC on Snow
Delta T|Anthropogenic|CH4
Delta T|Anthropogenic|Aerosols
Delta T|Anthropogenic|Tropospheric Ozone
Delta T|Anthropogenic|F-Gases|HFC
Delta T|Anthropogenic|Other|BC on Snow
Delta T|Anthropogenic|CH4
Delta T|Anthropogenic|Aerosols
Delta T|Anthropogenic|Tropospheric Ozone
Delta T|Anthropogenic|F-Gases|HFC
Delta T|Anthropogenic|Other|BC on Snow
Delta T|Anthropogenic|CH4
Delta T|Anthropogenic|Aerosols
Delta T|Anthropogenic|Tropospheric Ozone
Delta T|Anthropogenic|F-Gases|HFC
Delta T|Anthropogenic|Other|BC on Snow
Delta T|Anthropogenic|CH4
Delta T|Anthropogenic|Aerosols

```

Delta T|Anthropogenic|Tropospheric Ozone  
Delta T|Anthropogenic|F-Gases|HFC  
Delta T|Anthropogenic|Other|BC on Snow  
Delta T|Anthropogenic|CH4  
Delta T|Anthropogenic|Aerosols  
Delta T|Anthropogenic|Tropospheric Ozone  
Delta T|Anthropogenic|F-Gases|HFC  
Delta T|Anthropogenic|Other|BC on Snow  
Delta T|Anthropogenic|CH4  
Delta T|Anthropogenic|Aerosols  
Delta T|Anthropogenic|Tropospheric Ozone  
Delta T|Anthropogenic|F-Gases|HFC  
Delta T|Anthropogenic|Other|BC on Snow

```
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
  return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
  return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
  return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
  return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
  return np.nanmean(a, axis=axis, dtype=dtype)
/home/sarambl/miniconda3/envs/rcmip_ipcc/lib/python3.7/site-
packages/xarray/core/nanops.py:142: RuntimeWarning: Mean of empty slice
  return np.nanmean(a, axis=axis, dtype=dtype)
```

[56]: tab

[56]: ECS = 2K \ 2040

ssp119	CH4	-0.02659828770048701
	Aerosols	0.15646387941855683
	Tropospheric Ozone	-0.04992508208229368
	HFC	0.003468253587250494
	BC on Snow	-0.011329101981495735
ssp126	CH4	-0.017367724889572122
	Aerosols	0.12215597107108145
	Tropospheric Ozone	-0.033763352537327304
	HFC	0.011032221399586888
	BC on Snow	-0.009436533713336538

ssp245	CH4	0.028937393882524302
	Aerosols	0.03637900641138267
	Tropospheric Ozone	0.0009328311281484147
	HFC	0.018234694544699594
	BC on Snow	-0.003402782920874044
ssp370	CH4	0.05457162376258964
	Aerosols	-0.011068885443324872
	Tropospheric Ozone	0.024602751780866523
	HFC	0.01704878795819055
	BC on Snow	0.0036842324966297967
ssp370-lowNTCF-aerchemmip	CH4	0.05457162366753363
	Aerosols	0.06273291382163072
	Tropospheric Ozone	-0.0032062710739075963
	HFC	0.01704878793732071
	BC on Snow	-0.0037433274391409396
ssp585	CH4	0.044123353916243016
	Aerosols	0.047101148762349374
	Tropospheric Ozone	0.01531662537912159
	HFC	0.02710784418483706
	BC on Snow	-0.0018159626749880552
historical	CH4	0.0077132387149825565
	Aerosols	-0.009908627912094015
	Tropospheric Ozone	0.005241636725860588
	HFC	0.0007444353141619498
	BC on Snow	nan

2100

ssp119	CH4	-0.1296912754447551
	Aerosols	0.2249195956185493
	Tropospheric Ozone	-0.10048092297791049
	HFC	-0.0029761684916363817
	BC on Snow	-0.01697901215240332
ssp126	CH4	-0.1231558283831311
	Aerosols	0.22774103852286798
	Tropospheric Ozone	-0.09404727045262026
	HFC	-0.00132423401564886
	BC on Snow	-0.01532641975381759
ssp245	CH4	-0.004182704112309194
	Aerosols	0.1367591718072312
	Tropospheric Ozone	-0.039034660403500676
	HFC	0.038451376778719744
	BC on Snow	-0.012488796472324282
ssp370	CH4	0.21565799113002856
	Aerosols	-0.007421477076560101
	Tropospheric Ozone	0.06349719049159727

	HFC	0.06493327819133399
	BC on Snow	0.006754058664053657
ssp370-lowNTCF-aerchemmip	CH4	0.21565799026157
	Aerosols	0.14244876622095556
	Tropospheric Ozone	0.012046389288997983
	HFC	0.0649332767375719
	BC on Snow	-0.00704176174526431
ssp585	CH4	0.11489216246985463
	Aerosols	0.0979238452792276
	Tropospheric Ozone	0.011390724974023542
	HFC	0.14984300008060603
	BC on Snow	-0.0035326595202287737
historical	CH4	0.02703927122654534
	Aerosols	-0.04251408931871051
	Tropospheric Ozone	0.01849257773823193
	HFC	0.0017742750627926067
	BC on Snow	nan

ECS = 3.4K \

2040

ssp119	CH4	-0.04470130480462069
	Aerosols	0.26295450457415265
	Tropospheric Ozone	-0.08390451057176351
	HFC	0.005828775990740371
	BC on Snow	-0.019039783558255186
ssp126	CH4	-0.02918834373076379
	Aerosols	0.2052963468190799
	Tropospheric Ozone	-0.0567429727053181
	HFC	0.018540843568887466
	BC on Snow	-0.0158591174954173
ssp245	CH4	0.04863242622082048
	Aerosols	0.06113886248604996
	Tropospheric Ozone	0.0015677237971163881
	HFC	0.030645380185388666
	BC on Snow	-0.005718745441164742
ssp370	CH4	0.0917135273880785
	Aerosols	-0.018602461467488916
	Tropospheric Ozone	0.041347590445410684
	HFC	0.028652335655970435
	BC on Snow	0.006191751952510926
ssp370-lowNTCF-aerchemmip	CH4	0.09171352722832657
	Aerosols	0.10542945973064936
	Tropospheric Ozone	-0.005388485987327574
	HFC	0.0286523356208964
	BC on Snow	-0.006291067407225456
ssp585	CH4	0.07415407768433238



	Aerosols	0.07915858461200927
	Tropospheric Ozone	0.025741248736014292
	HFC	0.04555766969466912
historical	BC on Snow	-0.0030519220621472187
	CH4	0.012962933505788199
	Aerosols	-0.016652522954926197
	Tropospheric Ozone	0.008809138527872185
	HFC	0.0012511042161961299
	BC on Snow	nan

2100

ssp119	CH4	-0.2179602423824401
	Aerosols	0.37800175385322743
	Tropospheric Ozone	-0.16886907968150738
	HFC	-0.005001773662750117
	BC on Snow	-0.02853506985308847
ssp126	CH4	-0.20697671538153598
	Aerosols	0.3827434943996489
	Tropospheric Ozone	-0.15805662942987894
	HFC	-0.002225518763942189
	BC on Snow	-0.025757709244058458
ssp245	CH4	-0.007029487519546129
	Aerosols	0.22983860813230494
	Tropospheric Ozone	-0.06560197679979962
	HFC	0.06462170546081415
	BC on Snow	-0.020988775820408107
ssp370	CH4	0.36243662387632175
	Aerosols	-0.012472596455663742
	Tropospheric Ozone	0.106713909495384
	HFC	0.10912741049646246
	BC on Snow	0.01135092748863771
ssp370-lowNTCF-aerchemmip	CH4	0.3624366224167831
	Aerosols	0.23940058809757553
	Tropospheric Ozone	0.02024526260736541
	HFC	0.10912740805325775
	BC on Snow	-0.011834443693562069
ssp585	CH4	0.19308872932196103
	Aerosols	0.16457163351109735
	Tropospheric Ozone	0.01914334767497496
	HFC	0.2518273993750109
	BC on Snow	-0.005937017140460517
historical	CH4	0.045442425407349996
	Aerosols	-0.0714495341401904
	Tropospheric Ozone	0.031078780837636932
	HFC	0.0029818615123738887

	BC on Snow	nan
		ECS = 5K \
		2040
ssp119	CH4	-0.05744421069915064
	Aerosols	0.3379143859685943
	Tropospheric Ozone	-0.10782299096099926
	HFC	0.0074903727663812964
	BC on Snow	-0.024467414165359597
ssp126	CH4	-0.03750900280333449
	Aerosols	0.2638197398037043
	Tropospheric Ozone	-0.07291857125932286
	HFC	0.023826242414316925
	BC on Snow	-0.020380042392300957
ssp245	CH4	0.06249596853716284
	Aerosols	0.07856758799112298
	Tropospheric Ozone	0.002014631485887075
	HFC	0.039381393541404436
	BC on Snow	-0.007348976042039757
ssp370	CH4	0.11785810759372976
	Aerosols	-0.02390542559623008
	Tropospheric Ozone	0.05313446011989424
	HFC	0.03682019604658643
	BC on Snow	0.007956821513633937
ssp370-lowNTCF-aerchemmip	CH4	0.11785810738843767
	Aerosols	0.13548401159956747
	Tropospheric Ozone	-0.006924570228058991
	HFC	0.03682019600151392
	BC on Snow	-0.008084448613810082
ssp585	CH4	0.09529302290656273
	Aerosols	0.10172415398104348
	Tropospheric Ozone	0.03307925176935772
	HFC	0.0585446977071766
	BC on Snow	-0.003921926993890548
historical	CH4	0.016658249392053515
	Aerosols	-0.02139962225881925
	Tropospheric Ozone	0.011320340913645699
	HFC	0.0016077538343877845
	BC on Snow	nan
		2100
ssp119	CH4	-0.2800937051430452
	Aerosols	0.4857579099290342
	Tropospheric Ozone	-0.2170082290929777

ssp126	HFC	-0.006427618643534087
	BC on Snow	-0.03666950153066572
	CH4	-0.2659791274586253
	Aerosols	0.4918513683687795
	Tropospheric Ozone	-0.20311349664292128
ssp245	HFC	-0.0028599426649754825
	BC on Snow	-0.033100404639423535
	CH4	-0.009033368577154454
	Aerosols	0.2953582113555412
	Tropospheric Ozone	-0.08430299281060225
ssp370	HFC	0.08304327760575213
	BC on Snow	-0.026972001506768793
	CH4	0.46575566145192465
	Aerosols	-0.01602813300184842
	Tropospheric Ozone	0.13713461672709976
ssp370-lowNTCF-aerchemmip	HFC	0.14023612932577073
	BC on Snow	0.014586712247842122
	CH4	0.4657556595763185
	Aerosols	0.30764600461407887
	Tropospheric Ozone	0.026016536563311144
ssp585	HFC	0.14023612618608688
	BC on Snow	-0.015208063389011895
	CH4	0.24813212274858323
	Aerosols	0.21148571908213415
	Tropospheric Ozone	0.024600501084583216
historical	HFC	0.3236153005543126
	BC on Snow	-0.0076294699904560365
	CH4	0.058396600975961
	Aerosols	-0.09181750088603652
	Tropospheric Ozone	0.03993834279587727
	HFC	0.0038318944321908757
	BC on Snow	nan

## 2.8 Save output

```
[57]: tab.to_csv(PATH_DT_OUTPUT)
```