

# Documentation for forcing.py

## 1 Introduction

**Radiative Forcing** is the difference between sunlight absorbed by the Earth and energy radiated back to space. Positive radiative forcing means Earth receives more incoming energy from sunlight than it radiates to space. This net gain of energy will cause warming. Conversely, negative radiative forcing means that Earth loses more energy to space than it receives from the sun, which produces cooling. This is to say, positive forcing warms the system, while negative forcing cools it. Causes of radiative forcing include changes in insolation and the concentrations of radiatively active gases, commonly known as greenhouse gases and aerosols.

According to Wikipedia, radiative forcing can be used to estimate a subsequent change in equilibrium surface temperature  $\Delta T_t$  arising from that forcing via the equation

$$\Delta T_t = \lambda \Delta F \quad (1)$$

where  $\lambda$  is the climate sensitivity and  $\Delta F$  is the radiative forcing.

In our case, we take radiative forcing into consideration when we calculate damage for both the baseline scenarios as well as the simulated paths. To serve this end, forcing.py is essentially a class that computes the radiative forcing for the EZ-Climate model. It helps determine the excess energy created by GHGs in the atmosphere. Based on the business-as-usual scenario and the given mitigation level, forcing.py calculates the realized GHG levels under mitigation and the forcing component. To distinguish the GHG level under mitigation from its business-as-usual benchmark, we denote the benchmark as  $\bar{G}$  and the realized GHG level  $G$ .

## 2 Model of Excess Energy from GHGs

The climate damage function  $D_t(CRF_t, \theta_t)$  depends on  $CRF_t$ , the cumulative solar radiative forcing up to time  $t$  that determines global average temperature.  $CRF_t$  depends on the mitigation path up to point  $t$  in time. The cumulative mitigation is given by:

$$X_t = \frac{\sum_{s=0}^t g_s \cdot x_s}{\sum_{s=0}^t g_s} \quad (2)$$

where  $g_s$  is the flow of GHG emissions into the atmosphere in period  $s$ , for each period up to  $t$ , absent any mitigation. The cumulative GHG emissions that must be absorbed into the atmosphere or oceans is  $G_t(1 - X_t)$ , where  $G_t = \sum_{s=0}^t g_s$  denotes the cumulative emissions under the BAU scenario.

According to the paper, radiative forcing in a ten-year interval is given by

$$5.351 \cdot [\log(\text{GHG}) - \log(278.063)] \quad (3)$$

where GHG is the average level of atmospheric  $CO_2$ . The carbon absorption in a ten-year interval is given by

$$0.94835 \cdot \left| \text{GHG} - (285.6268 + 0.88414 \cdot \sum \text{absorption}) \right|^{0.741547} \quad (4)$$

where the sum is over absorption in previous periods. Here is the list of the parameters defined in the code, those in red are still unknown:

- `sink_start` = 35.596
- `forcing_start` = 4.926
- `forcing_p1` = 0.13173
- `forcing_p2` = 0.607773
- `forcing_p3` = 315.3785
- `absorption_p1` = 0.94835
- `absorption_p2` = 0.741547
- `lsc_p1` = 285.6268
- `lsc_p2` = 0.88414

### 3 Inputs

The methods embedded in the Forcing class require similar inputs that are explained below.

- **m**: an array of fractional mitigation levels that are denoted  $x_t$  in the paper.
- **node**: an integer that represents the node in the **TreeModel** for which forcing is to be calculated.
- **tree**: the **TreeModel** object whose tree structure is to be used.
- **bau**: the **BusinessAsUsual** object that gives the emission levels under the business-as-usual scenario.
- **subinterval\_len**: a float that represents the length of a subinterval. It is the size of intervals we model the probability of hitting a Tipping Point in the damage function. Within one period, the subintervals of this length are call **increments**. In our example, this value is 5.
- **returning**: an optional string that selects the output. Valid returns include "forcing", "ghg", and "both".

- "forcing": returns the forcing only;
- "ghg": returns the GHGs level only;
- "both": returns both the GHGs level and forcing.

## 4 Python: Forcing

```
from __future__ import division
import numpy as np
```

### 4.1 Attributes

The attributes of the Forcing class are basically parameters of the theoretical model, so they are all of the float type. See Equation 4 for reference.

- **sink\_start**: Its default value is 35.596.
- **forcing\_start**: Its default value is 4.926.
- **forcing\_p1**: Its default value is 0.13173.
- **forcing\_p2**: Its default value is 0.607773.
- **forcing\_p3**: Its default value is 315.3785.
- **absorption\_p1**: Its default value is 0.94835.
- **absorption\_p2**: Its default value is 0.741547.
- **lsc\_p1**: Its default value is 285.6268.
- **lsc\_p2**: Its default value is 0.88414.

```
class Forcing(object):
    """Radiative forcing for the EZ-Climate model. Determines the excess energy cr
    by GHGs in the atmosphere.

    Attributes
    -----
    sink_start : float
        sinking constant
    forcing_start : float
        forcing start constant
    forcing_p1 : float
        forcing constant
    forcing_p2 : float
        forcing constant
```

```

forcing_p3 : float
    forcing constant
absorption_p1 : float
    absorption constant
absorption_p2 : float
    absorption constant
lsc_p1 : float
    class constant
lsc_p2 : float
    class constant

"""

# parameters that I have no idea about
sink_start = 35.596
forcing_start = 4.926
forcing_p1 = 0.13173
forcing_p2 = 0.607773
forcing_p3 = 315.3785
absorption_p1 = 0.94835
absorption_p2 = 0.741547
lsc_p1 = 285.6268
lsc_p2 = 0.88414

```

## 4.2 Methods

There are 3 methods in this file, while **forcing\_at\_node** and **ghg\_level\_at\_node** simply executes **forcing\_and\_ghg\_at\_node** upon different requests. **forcing\_and\_ghg\_at\_node** implements the computation along a deterministic path and gives outputs specified by **forcing\_at\_node** and **ghg\_level\_at\_node**.

**forcing\_and\_ghg\_at\_node**: calculates the radiative forcing based on GHGs evolution that leads up to damage calculation. The steps are given below.

```

@classmethod
def forcing_and_ghg_at_node(cls, m, node, tree, bau, subinterval_len, returning=
    """Calculates the radiative forcing based on GHG evolution leading up to
    damage calculation in `node`.

    Parameters
    -----
    m : ndarray
        array of mitigations
    node : int
        node for which forcing is to be calculated

```

```

tree : `TreeModel` object
      tree structure used
bau : `BusinessAsUsual` object
      business-as-usual scenario of emissions
subinterval_len : float
      subinterval length
returning : string, optional
          * "forcing": implies only the forcing is returned
          * "ghg": implies only the GHG level is returned
          * "both": implies both the forcing and GHG level is returned

Returns
-----
tuple or float
    if `returning` is
        * "forcing": only the forcing is returned
        * "ghg": only the GHG level is returned
        * "both": both the forcing and GHG level is returned

"""

```

- Specify the case when the node is 0.

```

#for the start state, return 0 for forcing and ghg_start for the g
#call bau to get the ghg level
if node == 0:
    if returning == "forcing":
        return 0.0
    elif returning=="ghg":
        return bau.ghg_start
    else:
        return 0.0, bau.ghg_start

```

- Based on the node given, find its period, path, and decision times through **TreeModel**.
- Determine the number of increments within each period, of the length specified by **subinterval\_len**. (This number is converted into integer later, for convenience of computation.)

```

# get the period and the path that the target node are in
period = tree.get_period(node)
path = tree.get_path(node, period)
# the decision time is the time when we make a mitigation, i.e. an
period_lengths = tree.decision_times[1:period+1] - tree.decision_ti
#increments are the number counts of subintervals within a period
increments = period_lengths/subinterval_len

```

- Assign the starting values of forcing and GHGs level.

```

#assign beginning values
cum_sink = cls.sink_start
cum_forcing = cls.forcing_start
ghg_level = bau.ghg_start

```

- For each period that a node has undergone, determine its beginning and ending GHG levels under mitigation with the base case outputs from **bau.emission\_by\_decisions**. We assume that the emission level remains constant since the second to last period.

```

for p in range(0, period):
    #for each period, we calculate the start_emission and end_
    #! problem: when will the act takes in to effect? either u
    #emission_by_decision: the emission level at a decision po
    start_emission = (1.0 - m[path[p]]) * bau.emission_by_decis
    if p < tree.num_periods-1: #if not too late to implement m
        end_emission = (1.0 - m[path[p]]) * bau.emission_by
    else:
        end_emission = start_emission #emission level reman
    increment = int(increments[p])

```

- Within each period that a node has undergone, interpolate linearly the emission levels at the beginnings of all the increments in concern. Based on the emission levels, calculate the amounts of consequent ppm, absorption, and forcing.
- Calculate the GHG level at the beginning of each increment by updating values for forcing and absorption based on the GHG emissions.

– **p\_co2** precise meanings unknown

– **p\_c**

– **add\_p\_ppm**

– **absorption**:  $0.5 \times 0.94835 |GHG - (285.6268 + 0.88414 \cdot \sum \text{absorption})|^{0.741547}$

This is the carbon absorption in a 5-year interval. Since our subinterval is of length 5, the absorption should be half of that in Equation 4.

– **cum\_forcing**: the cumulative forcing is given by  $0.13173 \times |GHG - 315.3785|^{0.607773}$

```

# for each increment in a period, the forcing is affecting
for i in range(0, increment):
    #allocate the emission level change across time
    p_co2_emission = start_emission + i * (end_emission - start_emission)
    p_co2 = 0.71 * p_co2_emission
    p_c = p_co2 / 3.67
    add_p_ppm = subinterval_len * p_c / 2.13
    lsc = cls.lsc_p1 + cls.lsc_p2 * cum_sink
    absorption = 0.5 * cls.absorption_p1 * np.sign(ghg_level - cum_sink)
    cum_sink += absorption

```

```

        cum_forcing += cls.forcing_p1*np.sign(ghg_level-cls
        ghg_level += add_p_ppm - absorption

```

- Return the outcome according to the selection.

```

        if returning == "forcing":
            return cum_forcing
        elif returning == "ghg":
            return ghg_level
        else:
            return cum_forcing, ghg_level

```

**forcing\_at\_node:** returns the forcing based mitigation at a given node that leads up to the damage calculation.

```

@classmethod
def forcing_at_node(cls, m, node, tree, bau, subinterval_len):
    """Calculates the forcing based mitigation leading up to the
    damage calculation in `node`.

    Parameters
    -----
    m : ndarray
        array of mitigations in each node.
    node : int
        the node for which the forcing is being calculated.

    Returns
    -----
    float
        forcing

    """

    return cls.forcing_and_ghg_at_node(m, node, tree, bau, subinterval_len,

```

**ghg\_level\_at\_node:** returns the GHGs level at a given node that leads up to the damage calculation.

```

@classmethod
def ghg_level_at_node(cls, m, node, tree, bau, subinterval_len):
    """Calculates the GHG level leading up to the damage calculation in `n

    Parameters
    -----
    m : ndarray
        array of mitigations in each node.
    node : int

```

```

        the node for which the GHG level is being calculated.

Returns
-----
float
    GHG level at node

"""
return cls.forcing_and_ghg_at_node(m, node, tree, bau, subinterval_len, r

```