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DIGITAL ORGANISM SIMULATION ENVIRONMENT (DOSE) VERSION 1.0.4

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ABSTRACT

Evolution is a fundamental aspect of biology but examining evolution is difficult, time-consuming and costly. At the same time, molecular analysis of biological organisms is generally destructive, which presents a conundrum between observing possible evolutionary outcomes and in depth molecular analysis to decipher the corresponding evolutionary outcomes. Artificial life simulations via the use of digital organisms (DO) had been proposed as a feasible means of examining evolution *in silico* and had yield biologically relevant findings. Being digital, identical replicates can be made for analysis; thereby, resolving the conundrum. Recently, original implementation of DOSE (Ling, 2012a) had been improved (Castillo and Ling, 2014a) for use as a Python library for simplified construction of simulation, enabling database logging and revival of simulations. This manuscript documents the implementation and improvement of DOSE, which is released as DOSE version 1.0.4 (<https://github.com/mauriceling/dose/releases/tag/v1.0.4>) and licensed under GNU General Public License version 3. DOSE codebase is hosted and available for forking at <https://github.com/mauriceling/dose>.

Keywords: Simulation Environment, Digital Organism, Artificial Life

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INTRODUCTION

Evolution is a fundamental aspect of biology where organism adapt to the environment, resulting in emergent responses. One of the earlier iconic examples of evolution (Clarke et al., 1985) is the phenotype colour change of peppered moth from light coloured to dark coloured, as observed by higher occurrence of dark coloured peppered moth, during increased air pollution in the second half of 1800s. This is followed by reversal from darker coloured to lighter coloured, where occurrence of lighter coloured pepper moth increased from 6% to 30%, during reduced air pollution from 1959 to 1984 (Clarke et al., 1985). More recent examples include *Anolis carolinensis*, a lizard, evolving larger toepads to inhibit higher perches after invasion by another lizard, *Anolis sagrei* (Stuart et al., 2014). House mouse (*Mus musculus domesticus*) developing resistance to warfarin (Song et al., 2011), an anti-coagulant used as rodenticide.

However, testing evolutionary hypotheses is a challenge (Batut et al., 2013) as it is highly time consuming and expensive, if not impossible. Generally, many generations are required for adaptation; hence, the primary hurdle of evolutionary experiment in biology is the generation time. Thus, most experimental evolution can only be carried out using organisms with short generation time (Buckling et al., 2009); such as bacteria, yeast, or viruses. In addition, most genetic analysis to find out the basis for such adaptations requires the extraction of DNA from the cells; hence, it is destructive testing. At the same time, it is generally prohibitively expensive to analyse the genome of each cell. As a result, most genetic analyses are performed on groups of cells rather than individual cell.

Christopher Langton had conceptualized that by casting chemical reactions, reactants, and products into computable operations, operands, and outputs respectively, it may be possible to simulate artificial life (organisms in the digital world, or digital organisms) as cellular automata “living” on artificial chemistries (Langton, 1986). A study on the definition of life has also concluded that artificial life can be considered “alive” (Koh and Ling, 2013). Being digital, an organism or a cell can be replicated, by making another computational instance of the cell or by “reading” the cell from hard disk without the need to destroy the original copy, for analysis. Thereby, removing the dilemma related to destructive testing – the very sample of cells that is being genetically analysed are the cells containing emergent features. Moreover, it is also possible to analyse each genome as computational analysis of a genome generally requires less cost compared to biological analysis. As a result, a large volume of studies had used digital organisms to study biological evolution (Ling, 2014); including the effects of genome size (Gupta et al., 2016), effects of population (LaBar and Adami, 2016), speciation (Anderson and Harmon, 2014), and resistance to quorum quenching (Beckmann et al., 2012). Batut et al. (2013) argue that artificial life or digital organisms (DO) is a valuable tool to enable experimental evolution despite its drawbacks as repeated simulations can be carried

out with recording of all events. Bersini argued that artificial life and theoretical biology shared many common grounds and presented genetic algorithm (GA) as an important model to bridge the two fields (Bersini, 2009).

A Python implementation of GA, grounded on the biological hierarchy from genome to organisms to population (Lim et al., 2010), had been used as a basis to develop into a simulation framework for digital organisms, known as Digital Organisms Simulation Environment (DOSE) (Ling, 2012a). The genetic material of DOSE organisms is an executable DNA based on a parameterless 3-character instruction code, known as Ragaraja (Ling, 2012b), which Ling (2012a) argued that such an instruction set mimics the naturally occurring DNA.

Recently, the original implementation of DOSE (Ling, 2012a) had been improved (Castillo and Ling, 2014a) for use as a Python library for simplified construction of simulation, enabling database logging and revival of simulations. A utility case study of DOSE demonstrated that adjacent migration, such as foraging or nomadic behavior, increases heterozygosity (local genetic distance) while long distance migration, such as flight covering the entire ecosystem, does not increase heterozygosity. These results are consistent with previous studies (Raymond et al., 2013; Relethford, 1988). In addition, DOSE had been used to simulate antibiotics resistance (Castillo and Ling, 2014b; Castillo et al., 2015). This manuscript documents the implementation of DOSE version 1.0.4 (<https://github.com/mauriceling/dose/releases/tag/v1.0.4>), which is licensed under GNU General Public License version 3, except for COPADS (Collection of Python Algorithms and Data Structures, which is released under Python Software Foundation License version 2). DOSE codebase is hosted and available for forking at <https://github.com/mauriceling/dose>.

The rest of the manuscript is organized as follows. Section 2 provides a brief code description of DOSE version 1.0.4 (for detailed description, please see (Castillo and Ling, 2014a; Lim et al., 2010; Ling, 2012b)). Section 3 presents all 5 new and improved codes of DOSE; namely, *dose.py*, *simulation_calls.py*, *database_calls.py*, *dose_world.py* (bug fixed; original version in (Ling, 2012a)) and *genetic.py* (original version in (Lim et al., 2010)). Two other code files (original version in (Ling, 2012b)); namely, *register_machine.py* and *ragaraja.py*; were unchanged from the original version and will not be presented. Section 4 presents the simulation implementations for the first case study described (Castillo and Ling, 2014a), which examined the effects of different migration schemes on genetic heterozygosity (local genetic distance). Section 5 presents an example of reviving a simulation from Section 4 from the database file and continuing the simulation for another 200 generations. A brief guide on how to write a simulation is provided to end this article.

CODE DESCRIPTION

DOSE version 1.0.4 builds on previous versions of DOSE (Castillo and Ling, 2014a; Ling, 2012a; Ling, 2012b), which used a previous implementation of GA (*genetic.py*; (Lim et al., 2010)). The architecture of DOSE version 1.0.4 is given in Figure 1.

The most important enhancement to the original implementation of DOSE (Ling, 2012a) is the ability for the simulations to log all events into a SQLite database and subsequently, the ability to revive a simulation from the logged records and continuing the simulation. This is achieved in *database_calls.py* by logging all data attributes of each organism and the entire ecological system into the SQLite database, as well as re-creating all simulation objects (organisms and ecosystem) at a specific generation from the database. Other enhancements (in GA library; *genetic.py*; original version in Lim et al., 2010) are as follows: firstly, each organism is given a 32-character randomly generated name at initialization to aid in ancestry tracking. Secondly, chromosomes will inherit parental background mutation after crossover. Lastly, the statuses of each organism are increased from 6 (Lim et al., 2010) to 12. New statuses are

- parents (identity of parents),
- gender (gender of organism),
- blood (result of genomic interpretation or expression by Rajaraga interpreter),
- identity (32-character randomly generated name),
- deme (defined as a sub-population or local population), and
- location (location of the organism within the ecosystem).

The main application programming interface (API) is provided by *dose.py*, which contains the main functions needed to write a simulation. Once all needed simulation parameters and functions are defined (for details, please see Castillo and Ling, 2014a), *dose.simulate* function is called (see examples in Section 4). This will in turn call functions in *simulation_calls.py* prepare the simulation (Figure 2) by constructing, initializing the DOs, and mapping each DO onto the world. Simulation preparation is dependent on whether the simulation is a new simulation (regular simulation) or revival and continuation of an existing simulation (simulation revival) from database or files. For regular simulation, if the initial chromosome/genome is given, it will be used for organism construction; otherwise, a chromosome of single base will be used. The constructed organisms (population(s)) will be deployed accordingly before running the simulation (simulation core in Figure 2). On the other hand, a simulation can only be revived if either the simulation log files or the simulation results database with all required parameters and attributes logged are present. These will be used to reconstruct the

state of each organism, the world, and mapping each organism to the ecosystem before running the simulation.

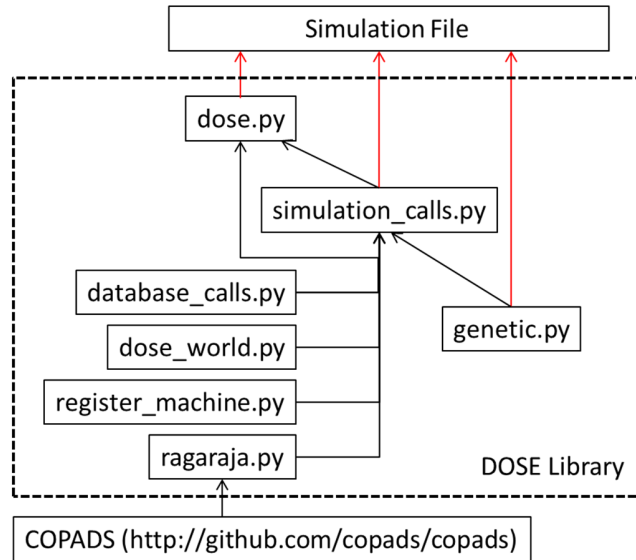


Figure 1. Architecture of DOSE Library. DOSE consists of 7 main files. Four of the files; `database_calls.py`, `dose_world.py`, `register_machine.py` and `ragaraja.py`; were used by `dose.py` and `simulation_calls.py`. `genetic.py` provides the GA for simulation. COPADS (released under Python Software Foundation License version 2) is required by Rajaraja interpreter (`ragaraja.py`; Ling, 2012b). Simulation File represents the simulation implementation, such as those in Sections 4 and 5 and requires `dose.py` and `simulation_calls.py`. `genetic.py`.

After simulation preparation is completed, a unique folder for storing text-based simulation results (results folder) is created using the datetime stamp of the simulation execution. Logging database will be created if database logging is requested and the database file is not present. The simulation is then executed from first generation to the maximum generation as defined in the parameter, and the events are reported into a text file or database as required (Figure 3; Castillo and Ling, 2014a). After DO initialization, the current simulation driver simulates each organism and ecological cell sequentially (Ling, 2012a). Hence, `simulation_calls.py` performs the task of the original command line simulation runner, `run_dose.py` (Ling, 2012a).

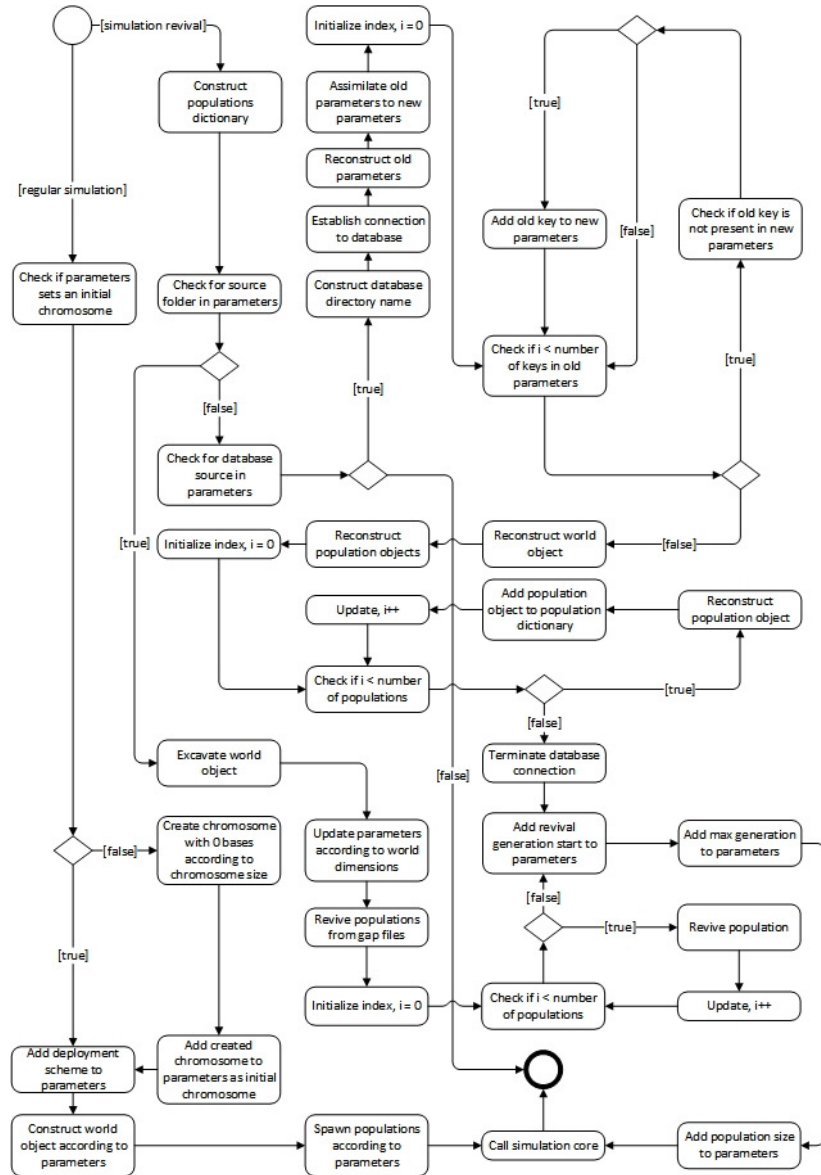


Figure 2. Activity Diagram for Simulation Preparation. Simulation preparation procedure is dependent on whether the simulation is a new simulation (regular simulation) or revival and continuation of an existing simulation (simulation revival) from database or files.

FORMATTED DOSE CODE FILES¹

FILE NAME: METADATA.PY

```
'''Digital Organism Simulation Environment (DOSE) Metadata

Date created: 14th February 2014'''

pypi_name = 'digital organism simulation environment'
description = '''Digital Organism Simulation \
Environment (DOSE)'''
version = '1.0.4'
maintainer = 'Maurice HT Ling'
email = 'mauriceling@acm.org'
download_url = \
'https://github.com/mauriceling/dose/releases'
project_start_year = '2010'
project_website = 'https://github.com/mauriceling/dose'
documentation_website = \
'http://maurice.vodien.com/project-dose'
short_license = 'GNU General Public License version 3'
long_license = '''Unless otherwise specified, dose.copads
package will be licensed under Python Software Foundation
License version 2; all other files will be licensed GNU
General Public License version 3.'''
platforms = ['Operating System :: OS Independent']
trove_classifiers = \
    ['Development Status :: 3 - Alpha',
     'Intended Audience :: Developers',
     'Intended Audience :: Education',
     'Intended Audience :: Science/Research',
     'License :: OSI Approved :: GNU General Public \
License v3 (GPLv3)',
     'Natural Language :: English',
     'Operating System :: OS Independent',
     'Programming Language :: Python :: 2 :: Only',
     'Topic :: Education',
     'Topic :: Scientific/Engineering :: Artificial Life',
     'Topic :: Scientific/Engineering :: Bio-Informatics',
     'Topic :: Software Development :: Libraries :: \
Python Modules']
```

¹ These code files had been formatted for display and printing purposes; hence, different from the original codes in <http://github.com/mauriceling/dose> on a line-by-line basis. However, they are logically, syntactically, and semantically identical to the original codes.

```
from datetime import datetime
copyright_year = '-'.join([project_start_year,
                           str(datetime.now().year)])
```

```
long_description = '''%s
```

```
Life is fascinating and deeply intriguing. Despite so, life
forms on Earth or carbon-based life forms as a group is just
one form, one possible sample of possibly a whole magnitude
of life. Even then, there are many aspects of
life that cannot be deciphered even by examining current
life forms; for example, how did chemical reactions organize
themselves into biochemical pathways? How did life start?
How is intelligence formed?
```

```
To answer such questions, we will have to restart our
evolutionary time to the very beginning - clearly an
impossibly gargantuan task. At the same time, studying
biological/carbon-based life forms is expensive, time
consuming and destructive. As a molecular biologist, there
is no way I can examine the entire genome of even a bacteria
in an inanimate state, then somehow allow it to continue
living as if time had just stopped while I am examining it.
```

```
However, if I can simulate a bacteria or any life form in a
computer, then I can make a digital copy of the bacterium,
pull it apart to study it while the original bacterium
continues "living" in my virtual world without even knowing
that it had been duplicated. Many biologists thought of
virtual life forms as a new way to learn about life itself.
Studying of virtual life forms is known as Artificial Life
and I term "virtual life forms" as "digital organisms".
There are several advantages in experimenting using digital
organisms. Firstly, generation time can be much faster
compared to most biological life. Secondly, it is usually
cheaper to examine computer simulations than working on
actual biological life. Perhaps the most important advantage
of looking at life from this perspective is that by
recreating life in a different medium, we are not limited to
our own system of carbon-based life; hence, studying life as
what-it-could-be.
```

```
Digital Organisms Simulation Environment (DOSE) is
essentially a virtual world simulator for studying digital
organisms. I will argue that digital organisms are
considered living organisms (Koh and Ling, 2013). Despite
so, being a molecular biologist by training, I have a hard
```

time mapping components of digital organisms into biological life whenever such components are too abstract.

Hence, I decided to design an artificial life / digital organism simulator that bears resemblance to biological life and ecology. These are the foundation papers:

1. Lim, JZR, Aw, ZQ, Goh, DJW, How, JA, Low, SXZ, Loo, BZL, Ling, MHT. 2010. A genetic algorithm framework grounded in biology. The Python Papers Source Codes 2: 6.

This manuscript describes the implementation of a GA framework that uses biological hierarchy – from chromosomes to organisms to population.

2. Ling, MHT. 2012. An Artificial Life Simulation Library Based on Genetic Algorithm, 3-Character Genetic Code and Biological Hierarchy. The Python Papers 7: 5.

Genetic algorithm (GA) is inspired by biological evolution of genetic organisms by optimizing the genotypic combinations encoded within each individual with the help of evolutionary operators, suggesting that GA may be a suitable model for studying real-life evolutionary processes. This paper describes the design of a Python library for artificial life simulation, Digital Organism Simulation Environment (DOSE), based on GA and biological hierarchy starting from genetic sequence to population. A 3-character instruction set that does not take any operand is introduced as genetic code for digital organism. This mimics the 3-nucleotide codon structure in naturally occurring DNA. In addition, the context of a 3-dimensional world composing of ecological cells is introduced to simulate a physical ecosystem.

3. Ling, MHT. 2012. Ragaraja 1.0: The Genome Interpreter of Digital Organism Simulation Environment (DOSE). The Python Papers Source Codes 4: 2.

This manuscript describes the implementation and test of Ragaraja instruction set version 1.0, which is the core genomic interpreter of DOSE.

From this foundation, the complete suite of Digital Organisms Simulation Environment (DOSE) can be build.

Project website: U{%s}

Documentation can be found at %s

License: %s

Copyright %s, Maurice HT Ling (on behalf of all authors).

```
''' % (description,
      project_website,
      documentation_website,
      long_license,
      copyright_year)
```

FILE NAME: __INIT__.PY

```
'''Digital Organism Simulation Environment (DOSE)
```

```
Date created: 27th September 2013'''
```

```
import metadata
```

```
__version__ = metadata.version
__maintainer__ = metadata.maintainer
__email__ = metadata.email
__description__ = metadata.long_description
```

```
# Package imports (in ascending order of package names)
```

```
import copads
```

```
# Module imports (in ascending order of module names)
```

```
import database_calls
```

```
import dose
```

```
import genetic
```

```
import register_machine
```

```
import ragaraja
```

```
# COPADS Class imports (in ascending order of module
# names, then class names)
```

```
from copads.lindenmayer import lindenmayer
```

```
# DOSE Class imports (in ascending order of module names,
# then class names)
```

```
from dose import dose_functions
```

```
from dose_world import World
```

```
from genetic import Chromosome
```

```
from genetic import Organism
```

```
from genetic import Population
```

```
# Function imports (in ascending order of module names,
# then function names)
```

```
from database_calls import connect_database
```

```

from database_calls import db_list_datafields
from database_calls import db_list_generations
from database_calls import db_list_simulations
from database_calls import db_list_population_name
from database_calls import db_get_ecosystem
from database_calls import \
db_get_organisms_chromosome_sequences
from database_calls import db_get_organisms_genome
from database_calls import db_get_organisms_status
from dose import database_report_populations
from dose import database_report_world
from dose import filter_age
from dose import filter_deme
from dose import filter_gender
from dose import filter_location
from dose import filter_status
from dose import filter_vitality
from dose import revive_simulation
from dose import simulate
from genetic import crossover
from genetic import population_constructor
from genetic import population_simulate

```

FILE NAME: DOSE.PY

```

'''Application Programming Interface (API) for DOSE (digital
organism simulation environment). This contains the main
functions and operations needed to write a DOSE simulation.
This file will be imported as top level (from dose import *)
when DOSE is imported; hence, all functions in this file can
be assessed at top level.

```

```

Date created: 27th September 2013'''
import sys, os, random, inspect

```

```

import ragaraja, register_machine
import dose_world
import genetic, simulation_calls

```

```

from simulation_calls import spawn_populations
from simulation_calls import simulation_core
from simulation_calls import excavate_world
from simulation_calls import revive_population

```

```

from database_calls import connect_database
from database_calls import \
db_reconstruct_simulation_parameters
from database_calls import db_reconstruct_population

```

```
from database_calls import db_reconstruct_world
from database_calls import db_list_simulations
```

```
class dose_functions():
    '''Abstract class to contain all of the simulation specific
    functions (functions that vary with each simulation) that
    are to be defined / implemented by the user to be used in a
    simulation. This class should be inherited by every
    simulation to over-ride each function / method.
```

```

    This set of functions / methods is consolidated functions /
    methods to be over-ridden from genetic.Organism,
    genetic.Population, and dose_world.World classes. As a
    result, the functions / methods can be working at different
    levels - at the level of individual organisms, at the level
    of entire population(s), or at the level of the world.
```

```

    Please see the examples in examples directory on its use.
    '''
```

```

    def mutation_scheme(self, organism):
        '''Method / function to trigger mutational events in
        each chromosome of the genome within an organism.
        This function works at the level of individual
        organisms.
```

```

        @param organism: genetic.Organism object
        @return: None'''
```

```
        raise NotImplementedError
```

```

    def prepopulation_control(self, Populations,
                              pop_name):
```

```

        '''Method / function to trigger population control
        events before mating event in each generation. For
        example, it can be used to simulate pre-puberty
        (childhood) death. This function works at the level
        of entire population(s).
```

```

        @param Populations: A dictionary containing one or
        more populations where the value is a
        genetic.Population object.
```

```

        @param pop_name: Name of the population which is used
        as key in the the dictionary (Populations parameter).
        @return: None'''
```

```
        raise NotImplementedError
```

```

    def fitness(self, Populations, pop_name):
```

```

        '''Method / function to calculate the fitness score
        of each organism within the population(s). This
        function works at the level of entire population(s)
        even though fitness calculation occurs at the
```

organism level. The fitness of each organism may be stored in `Organism.status['fitness']` and may be used by mating scheme.

```
@param Populations: A dictionary containing one or
more populations where the value is a
genetic.Population object.
@param pop_name: Name of the population which is used
as key in the the dictionary (Populations parameter).
@return: None'''
    raise NotImplementedError
def mating(self, Populations, pop_name):
    '''Method / function to trigger mating events in each
generation. For example, it can be used to simulates
mate choices and progeny size. A support function
provided is genetic.crossover() function which
generates one random crossover operation between 2
chromosomes, to simulate meiosis crossover. This
function may also use one or more of the
dose.filter_XXX() functions to select or choose
suitable mates. This function works at the level of
entire population(s), which means that this function
will have
    - to manage mating scheme and progeny (offspring)
generation for the entire population
    - add or replace offsprings into the respective
population(s)
    - (optional) store identity of parent(s) as list;
for example, [parentA identity, parentB
identity], in offspring's status['parents'] for
ancestral tracing.
```

```
@param Populations: A dictionary containing one or
more populations where the value is a
genetic.Population object.
@param pop_name: Name of the population which is used
as key in the the dictionary (Populations parameter).
@return: None'''
    raise NotImplementedError
def postpopulation_control(self, Populations,
                           pop_name):
    '''Method / function to trigger population control
events after mating event in each generation. For
example, it can be used to simulate old-age death.
This function works at the level of entire
population(s).
```

```
@param Populations: A dictionary containing one or
```

```

more populations where the value is a
genetic.Population object.
@param pop_name: Name of the population which is used
as key in the the dictionary (Populations parameter).
@return: None'''
    raise NotImplementedError
def generation_events(self, Populations, pop_name):
    '''Method / function to trigger other defined events
    in each generation. For example, it can be used to
    simulate catastrophe or epidemic that does not occur
    regularly, or simulates unusual occurrences of
    multiple mutation events. This function works at the
    level of entire population(s).

    @param Populations: A dictionary containing one or
    more populations where the value is a
    genetic.Population object.
    @param pop_name: Name of the population which is used
    as key in the the dictionary (Populations parameter).
    @return: None'''
    raise NotImplementedError
def population_report(self, Populations, pop_name):
    '''Method / function to generate a text report of the
    population(s) and/or each organisms within the
    population at regular intervals, within the
    simulation, as determined by "print_frequency" in the
    simulation parameters. This function works at the
    level of entire population(s).

    @param Populations: A dictionary containing one or
    more populations where the value is a
    genetic.Population object.
    @param pop_name: Name of the population which is used
    as key in the the dictionary (Populations parameter).
    @return: Entire report of a population at a
    generation count in a string. To make it human-
    readable, usually the report string is both tab-
    delimited (for one organism) and newline-delimited
    (for entire population).'''
    raise NotImplementedError
def organism_movement(self, Populations, pop_name,
                      World):
    '''organism_movement and organism_location are both
    methods / functions to execute movement of organisms
    within the world. The semantic difference between
    organism_movement and organism_location is that
    organism_movement is generally used for short travels
    while organism_location is used for long travel. For

```

example, `organism_movement` can be used to simulate foraging or nomadic behaviour. This function works at both the level of entire population(s) and world.

For each organism to move, this function will have to

- update the number of organisms in each `World.ecosystem[x-axis][y-axis][z-axis]['organisms']`
- update the respective Organism's location in the status dictionary (`Population[pop_name].agents[<index>].status['location']`)

@param Populations: A dictionary containing one or more populations where the value is a `genetic.Population` object.
 @param pop_name: Name of the population which is used as key in the the dictionary (Populations parameter).
 @param World: `dose_world.World` object.
 @return: None'''

```

    raise NotImplementedError
def organism_location(self, Populations,
                      pop_name, World):
    '''organism_movement and organism_location are both
    methods / functions to execute movement of organisms
    within the world. The semantic difference between
    organism_movement and organism_location is that
    organism_movement is generally used for short travels
    while organism_location is used for long travel. For
    example, organism_movement can be used to simulate
    foraging or nomadic behaviour. This function works at
    both the level of entire population(s) and world.
  
```

For each organism to move, this function will have to

- update the number of organisms in each `World.ecosystem[x-axis][y-axis][z-axis]['organisms']`
- update the respective Organism's location in the status dictionary (`Population[pop_name].agents[<index>].status['location']`)

@param Populations: A dictionary containing one or more populations where the value is a `genetic.Population` object.
 @param pop_name: Name of the population which is used as key in the the dictionary (Populations parameter).
 @param World: `dose_world.World` object.
 @return: None'''

```

    raise NotImplementedError
  
```

```

def ecoregulate(self, World):
    '''Method / function for broad spectrum management of
    the entire ecosystem defined as World.ecosystem[x-
    axis][y-axis][z-axis]['local_input'] and
    World.ecosystem[x-axis][y-axis][z-axis]
    ['local_output']). For example, it can be used to
    simulate temperature, solar radiation, or resource
    gradients. This function works at the level of the
    world.

    @param World: dose_world.World object.
    @return: None'''
    raise NotImplementedError

def update_ecology(self, World, x, y, z):
    '''Method / function to process the input and output
    from the activities of the organisms in the current
    ecological cell (defined as World.ecosystem[x-
    axis][y-axis][z-axis]['temporary_input'] and
    World.ecosystem[x-axis][y-axis][z-
    axis]['temporary_output']) into a local ecological
    cell condition (defined as World.ecosystem
    [x-axis][y-axis][z-axis]['local_input'] and
    World.ecosystem[x-axis][y-axis][z-
    axis]['local_output']), and update the ecosystem
    (which is essentially the ecological cell adjacent to
    World.ecosystem[x-axis][y-axis][z-axis]). For
    example, it can be used to simulate secretion of
    chemicals or use of resources (such as food) by
    organisms, and diffusion of secretions to the
    neighbouring ecological cells.

    Essentially, this function simulates the "diffusion"
    of local situation outwards. This function works at
    the level of the world.

    @param World: dose_world.World object.
    @param x: x-axis of the World.ecosystem to identify
    the cell.
    @param y: y-axis of the World.ecosystem to identify
    the cell.
    @param z: z-axis of the World.ecosystem to identify
    the cell.
    @return: None'''
    raise NotImplementedError

def update_local(self, World, x, y, z):
    '''Method / function to update local ecological cell
    condition (defined as World.ecosystem[x-axis][y-
    axis][z-axis]['local_input'] and World.ecosystem[x-

```

```
axis][y-axis][z-axis]['local_output']) from the
ecosystem (World.ecosystem [x-axis][y-axis][z-
axis]['local_input'] and World.ecosystem[x-axis][y-
axis][z-axis]['local_output'] of adjacent ecological
cells). For example, it can be used to simulates
movement or diffusion of resources from the ecosystem
to local.
```

Essentially, this function is the reverse of `update_ecology()` function. In this case, the local ecological cell is affected by adjacent conditions. This function works at the level of the world.

```
@param World: dose_world.World object.
@param x: x-axis of the World.ecosystem to identify
the cell.
@param y: y-axis of the World.ecosystem to identify
the cell.
@param z: z-axis of the World.ecosystem to identify
the cell.
@return: None'''
    raise NotImplementedError
def report(self, World):
    '''Method / function to generate a text report of the
ecosystem status (World.ecosystem) at regular
intervals, within the simulation, as determined by
"print_frequency" in the simulation parameters. This
function works at the level of world.

@param World: dose_world.World object.
@return: Entire report of a population at a
generation count in a string. To make it human-
readable, usually the report string is both tab-
delimited (for one organism) and newline-delimited
(for entire population).'''
    raise NotImplementedError
def deployment_scheme(self, Populations, pop_name,
                      World):
    '''Method / function to implement a user-specific /
simulation-specific deployment scheme used to deploy
organisms into the World. This function will only be
used when "deployment_code" in simulation parameters
dictionary equals to 0. This function works at all
three levels - organism(s), population(s), and world.

@param Populations: A dictionary containing one or
more populations where the value is a
genetic.Population object.
```

```

@param pop_name: Name of the population which is used
as key in the the dictionary (Populations parameter).
@param World: dose_world.World object.
@return: None'''
    raise NotImplementedError
def database_report(self, con, cur, start_time,
                    Population, World,
                    generation_count):
    '''Method / function to implement database logging of
    each organism in each population, and the ecosystem
    status. The frequency of logging is determined by
    "database_logging_frequency" in the
    simulation_parameters.

```

Three database tables had been defined for use in this function:

- organisms where the structure is (start_time text, pop_name text, org_name text, generation text, key text, value text)
- world where the structure is (start_time text, x text, y text, z text, generation text, key text, value text)
- miscellaneous where the structure is (start_time text, generation text, key text, value text)

The logical purpose of these tables are:

- organisms, to log status of each organism. Starting time of current simulation (start_time), population name (pop_name), organism name (org_name), and generation count (generation) are used as complex primary key to identify the organism within a specific simulation at a specific generation. Key and value pair makes up the actual data to be logged where key is the field and value is the datum or attribute.
- world, to log the status of the ecosystem. Starting time of current simulation (start_time), location of ecological cell (x, y, z as coordinates), and generation count (generation) are used as complex primary key to identify the ecological cell within a specific simulation at a specific generation. Key and value pair makes up the actual data to be logged where key is the field and value is the datum or attribute.
- miscellaneous, is used to log other undefined data. Starting time of current simulation (start_time), and generation count (generation)

are used as complex primary key to identify a specific generation within a specific simulation. Key and value pair makes up the actual data to be logged where key is the field and value is the datum or attribute.

```
@param con: Database connector. See Python DB-API for
details.
@param cur: Database cursor. See Python DB-API for
details.
@param start_time: Starting time of current
simulation in the format of <date>-<seconds since
epoch>; for example, 2013-10-11-1381480985.77.
@param Population: A dictionary containing one or
more populations where the value is a
genetic.Population object.
@param World: dose_world.World object.
@param generation_count: Current number of
generations simulated.
@return: None'''
    raise NotImplementedError

def database_report_populations(con, cur, start_time,
                               Populations,
                               generation_count):
    '''Function to log organisms' status and genome into
    database. Organisms' status is implemented as a dictionary
    and each key-value pair in the status is logged as a
    separate record. Similarly, each chromosome is logged as a
    separately record. A combination of starting time of the
    simulation, population name, organism's name, and the
    generation can identify all the data specific to an organism
    within a population, at a specific generation within a
    simulation. This function is a complete logger - it logs
    everything there is about an organism. There is nothing to
    log for populations.

    The following transformations of data are made:
    - Organism.status['blood'] is a list of numbers
    resulting from interpreting the genome by Ragaraja
    interpreter. The numbers are concatenated and
    delimited by '|'. For example, [1, 2, 3] ==> 1|2|3
    - Organism.status['location'] is a tuple of 3
    integers (x, y, z) for location of ecological cell.
    The numbers are concatenated and delimited by '|'.
    For example, (2, 3, 4) ==> 2|3|4
    - Each chromosome in Organism.genome is a list of
    bases. These bases are concatenated with no
```

```

sequence))

con.commit()

def database_report_world(con, cur, start_time, World,
                          generation_count):
    '''Function to log World.ecosystem into database. The
    ecosystem is made up of a collection of ecological cells,
    identified by the (x, y, z) coordinates within the
    ecosystem. Each ecological cell is implemented as a
    dictionary of ecological status. This function logs the
    entire set of ecological cells; thus, a complete logger.
    Each ecosystem within a specific simulation, at a specific
    time/generation, can be identified by a combination of
    starting time of simulation and generation. Each ecological
    cell within the ecosystem can be further identified by the
    (x, y, z) coordinates.

    @param con: Database connector. See Python DB-API for
    details.
    @param cur: Database cursor. See Python DB-API for details.
    @param start_time: Starting time of current simulation in
    the format of <date>-<seconds since epoch>; for example,
    2013-10-11-1381480985.77.
    @param World: dose_world.World object.
    @param generation_count: Current number of generations
    simulated.
    @return: None'''
    generation = str(generation_count)
    ecosystem = World.ecosystem
    location = [(x, y, z)
                 for x in range(len(ecosystem))
                 for y in range(len(ecosystem[x]))
                 for z in
                     range(len(ecosystem[x][y]))]
    for cell in location:
        eco_cell = ecosystem[cell[0]][cell[1]][cell[2]]
        for key in list(eco_cell.keys()):
            value = str(eco_cell[key])
            cur.execute('insert into world values
                        (?, ?, ?, ?, ?, ?, ?)',
                        (str(start_time), str(cell[0]),
                         str(cell[1]), str(cell[2]),
                         generation, key, value))

    con.commit()

def filter_deme(deme_name, agents):
    '''Function to identify organisms (agents) with a specific
    sub-population name (also known as deme) within a

```

population. Demes can be considered as strains in bacteria, breed or sub-species in animals, races in humans, and cultivars in plants. This function is can be used to support the identification of suitable mates for mating schemes.

```
@param deme_name: Name of deme (sub-population name)
@type deme_name: string
@param agents: A list of organisms, such as
Population.agents.
@return: List of Organism objects'''
    extract = [individual for individual in agents
                if individual.status['deme'].upper() == \
                deme_name.upper()]
    return extract

def filter_gender(gender, agents):
    '''Function to identify organisms (agents) with a specific
    gender within a population. This function is can be used to
    support the identification of suitable mates for mating
    schemes.

    @param gender: Gender
    @type gender: string
    @param agents: A list of organisms, such as
    Population.agents.
    @return: List of Organism objects'''
    extract = [individual for individual in agents
                if individual.status['gender'].upper() == \
                gender.upper()]
    return extract

def filter_age(minimum, maximum, agents):
    '''Function to identify organisms (agents) within a certain
    age range in a population. This function is can be used to
    support the identification of suitable mates for mating
    schemes.

    @param minimum: Minimum age
    @type minimum: float
    @param maximum: Maximum age
    @type maximum: float
    @param agents: A list of organisms, such as
    Population.agents.
    @return: List of Organism objects'''
    extract = [individual for individual in agents
                if float(individual.status['age']) > \
                (float(minimum) - 0.01) \
                and float(individual.status['age']) < \
```

```

        float(maximum) + 0.01]
    return extract

def filter_location(location, agents):
    '''Function to identify organisms (agents) of a population
    within a specific ecological cell. This function is can be
    used to support the identification of suitable mates for
    mating schemes.

    @param location: (x, y, z) coordinates within the
    World.ecosystem
    @param location: tuple
    @param agents: A list of organisms, such as
    Population.agents.
    @return: List of Organism objects'''
    extract = [individual for individual in agents
                if individual.status['location'] == \
                location]
    return extract

def filter_vitality(minimum, maximum, agents):
    '''Function to identify organisms (agents) within a certain
    vitality score in a population. This function is can be used
    to support the identification of suitable mates for mating
    schemes.

    @param minimum: Minimum vitality score
    @type minimum: float
    @param maximum: Maximum vitality score
    @type maximum: float
    @param agents: A list of organisms, such as
    Population.agents.
    @return: List of Organism objects'''
    extract = [individual for individual in agents
                if float(individual.status['vitality']) > \
                (float(minimum) - 0.01) \
                and float(individual.status['vitality']) < \
                float(maximum) + 0.01]
    return extract

def filter_status(status_key, condition, agents):
    '''Generic function to identity organisms (agents) within a
    population via the status of organisms (Organism.status
    dictionary). This function is can be used to support the
    identification of suitable mates for mating schemes.

    @param status_key: Status (key in Organism.status
    dictionary) to filter

```

```

@type status_key: string
@param condition: Condition of the status (value in
Organism.status dictionary) to filter. This can be a unique
condition, such as "True", or a range, such as (minimum,
maximum) to define the minimum and maximum value of the
condition.
@param agents: A list of organisms, such as
Population.agents.
@return: List of Organism objects'''
    if type(condition) in (str, int, float, bool):
        extract = [individual for individual in agents
                    if individual.status[status_key] == \
                    condition]
    else:
        extract = [individual for individual in agents
                    if float(individual.status[status_key]) > \
                    float(condition[0]) - 0.01 \
                    and float(individual.status[status_key]) < \
                    float(condition[1]) + 0.01]
    return extract

def revive_simulation(rev_parameters, sim_functions):
    print('\n[' + \
    rev_parameters["simulation_name"].upper() + \
    ' REVIVAL SIMULATION]')
    Populations = {}
    if "sim_folder" in rev_parameters:
        print('Accessing simulation files directory...')
        print('Excavating World entity: ' + \
              rev_parameters['eco_file'] + '...')
        World = \
        excavate_world(rev_parameters['sim_folder'] + \

        rev_parameters['eco_file'])
        print('Updating parameters with World \
dimensions...')
        rev_parameters["world_z"] = \
            len(World.ecosystem[0][0][0])
        rev_parameters["world_y"] = \
            len(World.ecosystem[0][0])
        rev_parameters["world_x"] = \
            len(World.ecosystem[0])
        for i in range(len(rev_parameters["pop_files"])):
            print('\nReviving population file: ' + \
                  rev_parameters["pop_files"][i] + '...')
            pop_file = rev_parameters["sim_folder"] + \
            rev_parameters["pop_files"][i]
            Populations[rev_parameters["population_names"][i]]

```

```

    = revive_population(pop_file)
    print('\nUpdating revival generation start in \
simulation parameters...')
    rev_parameters["rev_start"] = \
        [Populations[pop_name].generation
         for pop_name in Populations]
elif "database_source" in rev_parameters:
    print('Constructing database directory...')
    dbpath = os.sep.join([os.getcwd(),
                          'Simulations',
                          rev_parameters["database_source"]])
    print('Connecting to database file: ' + \
          rev_parameters["database_source"] + '...')
    (con, cur) = connect_database(dbpath, None)
    if rev_parameters["simulation_time"] == \
        'default':
        print('Acquiring simulation starting \
time...')
        rev_parameters["simulation_time"] = \
            db_list_simulations(cur)[0][0]
    print('Reconstructing old simulation \
parameters...')
    temp_parameters = \
        db_reconstruct_simulation_parameters(cur,
        rev_parameters["simulation_time"])
    print('Assimilating old simulation parameters \
with new simulation parameters...')
    for key in temp_parameters:
        if key not in rev_parameters:
            rev_parameters[key] = \
                temp_parameters[key]
    print('Reconstructing World entity...')
    World = db_reconstruct_world(cur,
        rev_parameters["simulation_time"],
        rev_parameters["rev_start"][0])
    print('\nUpdating population names parameter...')
    for pop_name in \
        rev_parameters["population_names"]:
        print('Reconstructing population: ' + \
              pop_name + '...')
        Populations[pop_name] = \
            db_reconstruct_population(cur,
            rev_parameters["simulation_time"], pop_name,
            rev_parameters["rev_start"][rev_parameters
            ["population_names"].index(pop_name)])
    print('Terminating database connection...')
    con.close()
    print('Updating last generation revival and \

```

```

population size simulation parameters...')
rev_parameters["rev_finish"] = \
[(Populations[pop_name].generation + \
rev_parameters["extend_gen"])
for pop_name in Populations]
rev_parameters["rev_pop_size"] = \
[len(Populations[pop_name].agents)
for pop_name in Populations]
print('\nStarting simulation core...')
simulation_core(sim_functions, rev_parameters,
                Populations, World)

```

```

def simulate(sim_parameters, sim_functions):
'''Function called by simulation to run the actual
simulation based on a set of parameters and functions.

```

Simulation parameters dictionary contains the following keys:

- simulation_name: Short name of the simulation.
- population_names: Population name(s) in a list.
- population_locations: A list of lists containing (x, y, z) coordinates in the world to deploy the populations. There must be equal numbers of items in this list as population_names. However, one population can be deployed into one or more ecological cell(s).
- deployment_code: Defines the type of deployment. Allowable values are 0 (custom deployment scheme which users can implement by over-riding dose.dose_functions.deployment_scheme() function), 1 (all organisms are in one location), 2 (organisms are randomly deployed across a list of population locations given for the population), 3 (organisms are randomly deployed across a list of population locations given for the population), 4 (organisms are dispersed from a specific location where the defined location will have the maximum allocation before dispersal happens). (see simulation_calls.deployment for more details)
- chromosome_bases: List containing allowable bases in the chromosome.
- background_mutation: Defines background mutation rate where 0.01 represents 1% mutation and 0.5 represents 50% mutation.
- additional_mutation: Defines mutation rate on top of background mutation rate where 0.01 represents 1% mutation and 0.5 represents 50% mutation.
- mutation_type: Type of mutation for background

mutation rate. Allowable values are 'point' (point mutation), 'insert' (insert a base), 'delete' (delete a base), 'invert' (invert a stretch of the chromosome), 'duplicate' (duplicate a stretch of the chromosome), 'translocate' (translocate a stretch of chromosome to another random position).

- chromosome_size: Number of bases for a chromosome.
- genome_size: Number of chromosome(s) in a genome.
- max_tape_length: Maximum number of cells for array used in Ragaraja interpreter, which will be stored as Organism.status['blood'].
- clean_cell: Toggle to define if a new tape will be used for Ragaraja interpreter. If True, at every Ragaraja interpretation of the source (genome), a new tape ([0] * max_tape_length) will be provided. If False, the tape stored as Organism.status['blood'] (results from previous interpretation of genome) will be used.
- interpret_chromosome: Toggle to define whether genomes will be interpreted by Ragaraja. If True, chromosomes of each organism will be interpreted by Ragaraja.
- max_codon: Maximum number of Ragaraja instructions to execute per organism, which can be used as force-break from endless loop.
- population_size: Number of organisms per population for initial deployment.
- eco_cell_capacity: Maximum number of organisms per ecological cell at the time of deployment prior to the start of simulation.
- world_x: Number of ecological cells in the x-axis of the World.ecosystem.
- world_y: Number of ecological cells in the y-axis of the World.ecosystem.
- world_z: Number of ecological cells in the z-axis of the World.ecosystem.
- goal: Goal for population to reach. This provides a goal for use in fitness functions.
- maximum_generations: Number of generations to simulate.
- fossilized_ratio: Proportion (less than 1) of the population to fossilize or freeze.
- fossilized_frequency: Number of generations (intervals) for each population fossilization or freezing event.
- print_frequency: Number of generations (intervals) for each reporting event into files.
- ragaraja_version: Ragaraja instruction version to

```

activate. (see ragaraja.activate_version() for more
details)
- ragaraja_instructions: A list defining a set of
Ragaraga instructions to activate. This is only
useful when ragaraja_version is 0.
- eco_buried_frequency: Number of generations
(intervals) for each ecological freezing or burial
event.
- database_file: Logging database file name. This
file will be located in <current working
directory>/Simulations folder
- database_logging_frequency: Number of generations
(intervals) for each database logging event.

```

Methods / Functions from dose.dose_functions class to be over-ridden as simulation_functions (for more details, please look at dose.dose_functions class):

```

- organism_movement: organism_movement and
organism_location are both methods / functions to
execute movement of organisms within the world. The
semantic difference between organism_movement and
organism_location is that organism_movement is
generally used for short travels while
organism_location is used for long travel.
- organism_location: organism_movement and
organism_location are both methods / functions to
execute movement of organisms within the world. The
semantic difference between organism_movement and
organism_location is that organism_movement is
generally used for short travels while
organism_location is used for long travel.
- ecoregulate: Broad spectrum management of the
entire ecosystem.
- update_ecology: Process the input and output from
the activities of the organisms in the current
ecological cell into a local ecological cell
condition, and update the ecosystem.
- update_local: Update local ecological cell
condition from the ecosystem.
- report: Generate a text report of ecosystem
(World.ecosystem) at regular intervals, within the
simulation, as determined by "print_frequency" in the
simulation parameters.
- fitness: Calculate the fitness score of each
organism within the population(s).
- mutation_scheme: Trigger mutational events in each
chromosome of the genome within an organism.
- prepopulation_control: Trigger population control

```

```

events before mating event in each generation.
- mating: Trigger mating events in each generation.
- postpopulation_control: Trigger population control
events after mating event in each generation.
- generation_events: Trigger other defined events in
each generation.
- population_report: Generate a text report of the
population(s) and/or each organisms within the
population at regular intervals, within the
simulation, as determined by "print_frequency" in the
simulation parameters
- database_report: Implement database logging of each
organism in each population, and the ecosystem
status. The frequency of logging is determined by
"database_logging_frequency" in the
simulation parameters.
- deployment_scheme: Implement a user-specific /
simulation-specific deployment scheme used to deploy
organisms into the World. This function will only be
used when "deployment_code" in simulation parameters
dictionary equals to 0.

@param sim_parameters: Dictionary of simulation parameters
@param sim_functions: A class inherited from
dose.dose_functions class to implement all the needed
simulation functions.'''
print('\n[ ' + \
sim_parameters["simulation_name"].upper() + \
' SIMULATION]')
if "initial_chromosome" not in sim_parameters:
    print('Adding initial chromosome to simulation \
parameters...')
    sim_parameters["initial_chromosome"] = ['0'] * \
sim_parameters["chromosome_size"]
print('Adding deployment scheme to simulation \
parameters...')
sim_parameters["deployment_scheme"] = \
sim_functions.deployment_scheme
print('Constructing World entity...')
World = dose_world.World(sim_parameters["world_x"],
                        sim_parameters["world_y"],
                        sim_parameters["world_z"])
print('Spawning populations...')
Populations = spawn_populations(sim_parameters)
print('\nStarting simulation core...')
simulation_core(sim_functions, sim_parameters,
                Populations, World)

```

FILE NAME: DOSE_WORLD.PY

```
'''World structure for DOSE (digital organism simulation
environment)
```

```
Date created: 13th September 2012
```

```
Reference: Ling, MHT. 2012. An Artificial Life Simulation
Library Based on Genetic Algorithm, 3-Character Genetic Code
and Biological Hierarchy. The Python Papers 7: 5.'''
```

```
import copy
```

```
class World(object):
```

```
    '''Representation of a 3-dimensional ecological world.
```

```
    The ecosystem is made up of ecological cells. Each
    ecological cell is modelled as a dictionary of
```

- local_input: A list containing processed input,
representing the partial local ecological condition,
to be used as input to the organisms in the current
ecological cell. This is updated by
World.update_local function.
- local_output: A list containing processed output,
representing the partial local ecological condition.
This is updated by World.update_local function.
- temporary_input: A list acting as temporary holding
for input after being fed to the organisms in the
current ecological cell, which is to be used to
update local_input and local_output lists by
World.update_local and World.update_ecology
functions.
- temporary_output: A list acting as temporary
holding for output from the organisms in the current
ecological cell, which is to be used to update
local_input and local_output lists by
World.update_local and World.update_ecology
functions.
- organisms: The number of organisms in the current
ecological cell which is updated by
World.organism_movement and World.organism_location
functions.

```
@see: Ling, MHT. 2012. An Artificial Life Simulation Library
Based on Genetic Algorithm, 3-Character Genetic Code and
Biological Hierarchy. The Python Papers 7: 5.'''
```

```
def __init__(self, world_x, world_y, world_z):
    '''Setting up the world and ecosystem
```

```

    @param world_x: number of ecological cells on the x-
    axis
    @type world_x: integer
    @param world_y: number of ecological cells on the y-
    axis
    @type world_y: integer
    @param world_z: number of ecological cells on the z-
    axis
    @type world_z: integer'''
    self.ecosystem = {}

    eco_cell = {'local_input': [],
                'local_output': [],
                'temporary_input': [],
                'temporary_output': [],
                'organisms': 0}
    self.world_x = int(world_x)
    self.world_y = int(world_y)
    self.world_z = int(world_z)
    for x in range(self.world_x):
        eco_x = {}
        for y in range(self.world_y):
            eco_y = {}
            for z in range(self.world_z):
                eco_y[z] = copy.deepcopy(eco_cell)
            eco_x[y] = copy.deepcopy(eco_y)
        self.ecosystem[x] = copy.deepcopy(eco_x)

def eco_burial(self, filename):
    '''Function to preserve the entire ecosystem.

    @param filename: file name of preserved ecosystem.'''
    try:
        import cPickle as pickle
    except ImportError:
        import pickle

    f = open(filename, 'wb')
    pickle.dump(self.ecosystem, f)
    f.close()

def eco_excavate(self, filename):
    '''Function to excavate entire ecosystem.

    @param filename: file name of preserved ecosystem.'''
    try:
        import cPickle as pickle
    except ImportError:

```

```
import pickle

self.ecosystem = \
    pickle.load(open(filename, 'rb'))

def ecoregulate(self):
    '''Function to simulate events to the entire
    ecosystem. B{This function may be over-ridden by the
    inherited class or substituted to cater for
    ecological schemes but not an absolute requirement
    to do so.}'''
    pass

def organism_movement(self, x, y, z):
    '''Function to trigger organism movement from current
    ecological cell to an adjacent ecological cell.
    B{This function may be over-ridden by the inherited
    class or substituted to cater for mobility schemes
    but not an absolute requirement to do so.}

    @param x: location of current ecological cell on the
    x-axis
    @type x: integer
    @param y: location of current ecological cell on the
    y-axis
    @type y: integer
    @param z: location of current ecological cell on the
    z-axis
    @type z: integer'''
    pass

def organism_location(self, x, y, z):
    '''Function to trigger organism movement from current
    ecological cell to a distant ecological cell. B{This
    function may be over-ridden by the inherited class or
    substituted to cater for mobility schemes but not an
    absolute requirement to do so.}

    @param x: location of current ecological cell on the
    x-axis
    @type x: integer
    @param y: location of current ecological cell on the
    y-axis
    @type y: integer
    @param z: location of current ecological cell on the
    z-axis
    @type z: integer'''
    pass
```

```
def update_ecology(self, x, y, z):
    '''Function to process temporary_input and
    temporary_output from the activities of the organisms
    in the current ecological cell into a local
    ecological cell condition, and update the ecosystem.
    B{This function may be over-ridden by the inherited
    class or substituted to cater for ecological schemes
    but not an absolute requirement to do so.}

    @param x: location of current ecological cell on the
    x-axis
    @type x: integer
    @param y: location of current ecological cell on the
    y-axis
    @type y: integer
    @param z: location of current ecological cell on the
    z-axis
    @type z: integer'''
    pass

def update_local(self, x, y, z):
    '''Function to update local ecological cell condition
    from the ecosystem. B{This function may be over-
    ridden by the inherited class or substituted to cater
    for ecological schemes but not an absolute
    requirement to do so.}

    @param x: location of current ecological cell on the
    x-axis
    @type x: integer
    @param y: location of current ecological cell on the
    y-axis
    @type y: integer
    @param z: location of current ecological cell on the
    z-axis
    @type z: integer'''
    pass

def report(self):
    '''Function to report the status of the world and
    ecosystem. B{This function may be over-ridden by the
    inherited class or substituted to cater for specific
    reporting schemes but not an absolute requirement to
    do so.}

    @return: dictionary of status describing the current
    generation'''
    pass
```

FILE NAME: DATABASE_CALLS.PY

'''File containing support functions for database logging of simulations.

Date created: 10th October 2013'''

```
import os, copy
import sqlite3 as s
```

```
def connect_database(dbpath, sim_parameters=None):
    '''Connects to logging database and prepares database for
    use, if database does not exist. This function can be used
    to connect to the logging database using a file path or
    using simulation parameters dictionary.
```

Simulation parameters dictionary takes precedence - if simulation parameters dictionary (sim_parameters) is not None, this function will look for database file in <simulation execution directory>/Simulations/<database file>.

In both cases, if the database is not present, it will create a SQLite3 database file and create the required database tables.

@param dbpath: File path of logging database. This parameter will only be used when sim_parameters == None.

@param sim_parameters: Dictionary of simulation parameters. Default is None.

@return: (con, cur) where

```
- con = connector
- cur = cursor'''
```

```
if sim_parameters:
    dbpath = os.sep.join([os.getcwd(),
                          'Simulations',
                          sim_parameters["database_file"]])
```

```
con = s.connect(dbpath)
```

```
cur = con.cursor()
```

```
cur.execute('''create table if not exists parameters
              (start_time text, simulation_name text,
               key text, value text)''')
```

```
cur.execute('''create table if not exists organisms
              (start_time text, pop_name text,
               org_name text, generation text,
               key text, value text)''')
```

```
cur.execute('''create table if not exists world
              (start_time text, x text, y text,
```

```

        z text, generation text,
        key text, value text)'''
cur.execute(''create table if not exists
miscellaneous
(start_time text, generation text,
key text, value text)''')
cur.execute(''create index if not exists
organisms_index1 on organisms
(pop_name, generation)''')
cur.execute(''create index if not exists
organisms_index2 on organisms
(generation, org_name)''')
con.commit()
return (con, cur)

def db_log_simulation_parameters(con, cur,
                                sim_parameters):
    '''Function to log the simulation parameters. A database
    table, parameters, had been defined for simulation
    parameters logging where the structure is (start_time text,
    simulation_name text, key text, value text). The starting
    time of current simulation (start_time) and name of the
    current simulation (simulation_name) are used as complex
    primary key to identify the current simulation. All
    parameters will be logged, except for "starting_time" and
    "simulation_name".

The following transformations of data are made:
- Population name (key = "population_names") is a
list of names. These are concatenated and delimited
by '|'. For example, ['pop_01', 'pop_02'] ==>
pop_01|pop_02
- Genetic bases (key = "chromosome_bases") is a list
of bases to make up the genetic code. These are
concatenated and delimited by '|'. For example, ['1',
'2'] ==> 1|2
- Ragaraja instructions to be used (key =
"ragaraja_instructions") is a list of 3-character
Ragaraja instructions in numbers. These are
concatenated and delimited by '|'. For example,
['000', '004', '008'] ==> 000|004|008
- Initial (ancestral) chromosome is a list of bases.
These bases are concatenated with no delimiter. For
example, [1, 2, 3] ==> 123

@param con: Database connector from connect_database()
function.
```

```

@param cur: Database cursor from connect_database()
function.
@param sim_parameters: Dictionary of parameters used in
simulate on.
@return: (con, cur) where
    - con = connector
    - cur = cursor'''
start_time = sim_parameters["starting_time"]
simulation_name = sim_parameters["simulation_name"]
for key in [k for k in list(sim_parameters.keys())
            if k not in ("simulation_name",
                        "starting_time")]:
    value = sim_parameters[key]
    if key in ("population_names",
              "chromosome_bases",
              "ragaraja_instructions"):
        value = '|'.join([str(x) for x in value])
        cur.execute('''insert into parameters \
                      values (?, ?, ?, ?)''', (str(start_time),
                                                str(simulation_name), str(key), value))
    elif key in ("initial_chromosome"):
        value = '|'.join(value)
        cur.execute('''insert into parameters \
                      values (?, ?, ?, ?)''', (str(start_time),
                                                str(simulation_name), str(key), value))
    else:
        cur.execute('''insert into parameters \
                      values (?, ?, ?, ?)''', (str(start_time),
                                                str(simulation_name), str(key),
                                                str(value)))
con.commit()
return (con, cur)

def db_report(con, cur, sim_functions, start_time,
              Populations, World, generation_count):
    '''Wrapper around user-implemented database logging which
    over-rides dose.dose_functions.database_report() function.

@param con: Database connector from connect_database()
function.
@param cur: Database cursor from connect_database()
function.
@param sim_functions: Object of simulation-specific
functions, which inherits dose.dose_functions class.
@param start_time: Starting time of current simulation in
the format of <date>-<seconds since epoch>; for example,
2013-10-11-1381480985.77.

```

```

@param Populations: A dictionary containing one or more
populations where the value is a genetic.Population object.
@param World: dose_world.World object.
@param generation_count: Current number of generations
simulated.
@return: (con, cur) where
    - con = connector
    - cur = cursor '''
sim_functions.database_report(con, cur, start_time,
                             Populations, World,
                             generation_count)

con.commit()
return (con, cur)

def db_list_simulations(cur, table='parameters'):
    '''Function to list simulations, identified by starting time
    of the simulation (and simulation name, if available), from
    logging database.

    @param cur: Database cursor from connect_database()
    function.
    @param table: Database table name to list simulations.
    Allowable values are 'parameters', 'organisms', 'world', and
    'miscellaneous'. Default value is 'parameters'.
    @return: A list containing the results.
        - If table = 'parameters', the returned list will be
        a list of list (consisting of [starting time of
        simulation, simulation name]).
        - If table is not 'parameters', the returned list
        will be a list of starting time of simulation.'''
    if table not in ('parameters', 'organisms',
                    'world', 'miscellaneous'):
        table = 'parameters'
    if table == 'parameters':
        cur.execute("""select distinct start_time,
                                simulation_name from parameters""")
        return [[str(x[0]), str(x[1])]
                for x in cur.fetchall()]
    else:
        cur.execute("select distinct start_time from %s",
                    table)
        return [str(x[0]) for x in cur.fetchall()]

def db_list_generations(cur, start_time,
                        table='organisms'):
    '''Function to list all logged generations within a
    simulation, identified by starting time of the simulation.

```

```

@param cur: Database cursor from connect_database()
function.
@param start_time: Starting time of current simulation in
the format of <date>-<seconds since epoch>; for example,
2013-10-11-1381480985.77.
@param table: Database table name to list simulations.
Allowable values are 'parameters', 'organisms', 'world', and
'miscellaneous'. Default value is 'organisms'.
@return: A list containing the results (generation
counts).'''
# query plan: SCAN TABLE organisms USING COVERING INDEX
# organisms_index2
    cur.execute("select distinct generation from %s \
        where start_time='%s'" % (str(table),
            str(start_time)))
    generations = sorted([int(str(x[0]))
        for x in cur.fetchall()])
    return [str(gen) for gen in generations]

def db_list_datafields(cur, start_time,
    table='organisms'):
    '''Function to list all logged data fields (types of data
    logged) within a simulation, identified by starting time of
    the simulation.

    @param cur: Database cursor from connect_database()
    function.
    @param start_time: Starting time of current simulation in
    the format of <date>-<seconds since epoch>; for example,
    2013-10-11-1381480985.77.
    @param table: Database table name to list simulations.
    Allowable values are 'parameters', 'organisms', 'world', and
    'miscellaneous'. Default value is 'organisms'.
    @return: A list containing the results (types of data
    logged).'''
    cur.execute("select distinct key from %s where \
        start_time='%s'" % (str(table), str(start_time)))
    return [str(x[0]) for x in cur.fetchall()]

def db_list_population_name(cur, start_time):
    '''Function to list all logged populations (as population
    names) within a simulation, identified by starting time of
    the simulation.

    @param cur: Database cursor from connect_database()
    function.

```

```

@param start_time: Starting time of current simulation in
the format of <date>-<seconds since epoch>; for example,
2013-10-11-1381480985.77.
@return: A list containing the results (types of data
logged).'''
# query plan: SCAN TABLE organisms USING COVERING INDEX
# organisms_index1
    cur.execute("select distinct pop_name from \
        organisms where start_time='%s'" % \
            str(start_time))
    return [str(x[0]) for x in cur.fetchall()]

def db_get_ecosystem(cur, start_time, datafield='all',
                    generation='all'):
    '''Analysis helper function to get a specific field of the
    ecosystem (World.ecosystem) or the entire ecosystem for one
    or more generations within a simulation (as identified by
    the starting time of the simulation).

    @param cur: Database cursor from connect_database()
    function.
    @param start_time: Starting time of current simulation in
    the format of <date>-<seconds since epoch>; for example,
    2013-10-11-1381480985.77.
    @param datafield: The specific datafield in World.ecosystem
    to extract. Predefined datafields for ecosystem are
    'local_input', 'local_output',
    'temporary_input', 'temporary_output', and 'organisms'.
    Default = 'all', which returns the entire World.ecosystem(s)
    of the generation(s). Only one or all datafield(s) in
    World.ecosystem can be extracted at a time.
    @type datafield: string
    @param generation: Generation(s) to extract World.ecosystem.
    Default = 'all', which extracts World.ecosystems for all
    logged generations within the specific simulation.
    @type generation: list
    @return: A dictionary of results - {<generation count>:
    <value>}. The value is the value of the specific datafield
    if a specific datafield is required (the data type of this
    value is dependent on that of the data type of
    World.ecosystem[datafield]) or the entire ecosystem as a
    dictionary.'''
    if generation == 'all':
        generation = db_list_generations(cur, start_time,
                                         'world')

    else:
        generation = [str(x) for x in generation]
    results = {}

```

```

    for gen in generation:
        results[gen] = {}
        ecosystem = db_reconstruct_world(cur, start_time,
                                         gen).ecosystem

        if datafield == 'all':
            results[gen] = ecosystem
        else:
            for x in list(ecosystem.keys()):
                results[gen][x] = {}
                for y in list(ecosystem[x].keys()):
                    results[gen][x][y] = {}
                    for z in list(ecosystem[x][y].
                                keys()):
                        results[gen][x][y][z] = \
                            ecosystem[x][y][z][datafield]

    return results

def db_get_organisms_status(cur, start_time,
                           population_name, datafield='all', generation='all'):
    '''Analysis helper function to get a specific field of the
    Organism.status dictionary or the entire Organism.status
    dictionary for one or more generations within a simulation
    (as identified by the starting time of the simulation).

    @param cur: Database cursor from connect_database()
    function.
    @param start_time: Starting time of current simulation in
    the format of <date>-<seconds since epoch>; for example,
    2013-10-11-1381480985.77.
    @param population_name: Name of the population.
    @param datafield: The specific key in Organism.status
    dictionary to extract. Predefined keys for Organism.status
    dictionary are 'age', 'alive', 'blood', 'death', 'deme',
    'fitness', 'gender', 'identity',
    'lifespan', 'location', 'parents', 'vitality'. Default =
    'all', which returns the entire Organism.status dictionaries
    of the generation(s). Only one or all datafield(s) in
    Organism.status dictionary can be extracted at a time.
    @type datafield: string
    @param generation: Generation(s) to extract Organism.status
    dictionary. Default = 'all', which extracts Organism.status
    dictionary for all logged generations within the specific
    simulation.
    @type generation: list
    @return: A dictionary of dictionary of results -
    {<generation count>: {<Organism identity>: <value>}}. The
    value is the value of the specific datafield if a specific

```

datafield is required (the data type of this value is dependent on that of the data type of Organism.status[datafield]) or the entire Organism.status as a dictionary. '''

```

    if generation == 'all':
        generation = db_list_generations(cur, start_time,
                                          'organisms')

    else:
        generation = [str(x) for x in generation]
    results = {}
    for gen in generation:
        results[gen] = {}
        agents = db_reconstruct_organisms(cur,
                                          start_time, population_name, gen)
        for org in agents:
            identity = org.status['identity']
            if datafield == 'all':
                results[gen][identity] = org.status
            else:
                results[gen][identity] = \
                    org.status[datafield]
    return results

def db_get_organisms_genome(cur, start_time,
                           population_name, generation='all'):
    '''Analysis helper function to get entire genome of
    organisms for one or more generations within a simulation
    (as identified by the starting time of the simulation).

    @param cur: Database cursor from connect_database()
    function.
    @param start_time: Starting time of current simulation in
    the format of <date>-<seconds since epoch>; for example,
    2013-10-11-1381480985.77.
    @param population_name: Name of the population.
    @param generation: Generation(s) to extract Organism.genome.
    Default = 'all', which extracts Organism.genome for all
    logged generations within the specific simulation.
    @type generation: list
    @return: A dictionary of dictionary of genome -
    {<generation count>: {<Organism identity>: <genome as list
    of chromosomes>}}.'''
    if generation == 'all':
        generation = db_list_generations(cur, start_time,
                                          'organisms')

    else:
        generation = [str(x) for x in generation]
    results = {}

```

```

    for gen in generation:
        results[gen] = {}
        agents = db_reconstruct_organisms(cur,
                                          start_time, population_name, gen)
        for org in agents:
            identity = org.status['identity']
            results[gen][identity] = org.genome
    return results

def db_get_organisms_chromosome_sequences(cur,
                                          start_time, population_name, generation='all'):
    '''Analysis helper function to get chromosomal sequences of
    entire genome of organisms for one or more generations
    within a simulation (as identified by the starting time of
    the simulation).

    @param cur: Database cursor from connect_database()
    function.
    @param start_time: Starting time of current simulation in
    the format of <date>-<seconds since epoch>; for example,
    2013-10-11-1381480985.77.
    @param population_name: Name of the population.
    @param generation: Generation(s) to extract Organism.genome.
    Default = 'all', which extracts Organism.genome for all
    logged generations within the specific simulation.
    @type generation: list
    @return: A dictionary of dictionary of chromosomal sequences
    - {<generation count>: {<Organism identity>: <list (all
    chromosomes) of list (individual chromosome) of chromosomal
    sequence>}}.'''
    genome_dict = db_get_organisms_genome(cur,
                                          start_time, population_name,
                                          generation)

    results = {}
    for gen in list(genome_dict.keys()):
        results[gen] = {}
        for identity in list(genome_dict[gen].keys()):
            results[gen][identity] = \
                [chromosome.sequence
                 for chromosome in
                 genome_dict[gen][identity]]
    return results

def db_reconstruct_simulation_parameters(cur,
                                         start_time):
    '''Function to reconstruct simulation parameters dictionary
    of a simulation (as identified by the starting time of the
    simulation).

```

```

@param cur: Database cursor from connect_database()
function.
@param start_time: Starting time of current simulation in
the format of <date>-<seconds since epoch>; for example,
2013-10-11-1381480985.77.
@return: Simulation parameters dictionary containing the
following keys:
    - simulation_name
    - population_names
    - population_locations
    - deployment_code
    - chromosome_bases
    - background_mutation
    - additional_mutation
    - mutation_type
    - chromosome_size
    - genome_size
    - max_tape_length
    - clean_cell
    - interpret_chromosome
    - max_codon
    - population_size
    - eco_cell_capacity
    - world_x
    - world_y
    - world_z
    - goal
    - maximum_generations
    - fossilized_ratio
    - fossilized_frequency
    - print_frequency
    - ragaraja_version
    - ragaraja_instructions
    - eco_buried_frequency
    - database_file
    - database_logging_frequency'''
parameters = {}
for key in ('simulation_name', 'population_names',
            'population_locations',
            'deployment_code', 'chromosome_bases',
            'background_mutation',
            'additional_mutation', 'mutation_type',
            'chromosome_size', 'genome_size',
            'max_tape_length', 'clean_cell',
            'interpret_chromosome', 'max_codon',
            'population_size', 'eco_cell_capacity',
            'world_x', 'world_y', 'world_z', 'goal',

```

```

        'maximum_generations',
        'fossilized_ratio',
        'fossilized_frequency',
        'print_frequency', 'ragaraja_version',
        'ragaraja_instructions',
        'eco_buried_frequency', 'database_file',
        'database_logging_frequency')):
    parameters[key] = None
start_time = str(start_time)
cur.execute("select distinct simulation_name from \
    parameters where start_time = '%s'" % start_time)
parameters["simulation_name"] = \
    str(cur.fetchone()[0])
cur.execute("select key, value from parameters \
    where start_time = '%s'" % start_time)
for r in cur.fetchall():
    if str(r[0]) == 'population_names':
        value = str(r[1]).split('|')
        exec("parameters['population_names'] = %s" % \
            str(value))
    elif str(r[0]) == 'population_locations':
        exec("parameters['population_locations'] = \
            %s" % str(r[1]))
    elif str(r[0]) == 'deployment_code':
        parameters['deployment_code'] = int(r[1])
    elif str(r[0]) == 'chromosome_bases':
        value = str(r[1]).split('|')
        exec("parameters['chromosome_bases'] = %s" % \
            str(value))
    elif str(r[0]) == 'background_mutation':
        parameters['background_mutation'] = \
            float(r[1])
    elif str(r[0]) == 'additional_mutation':
        parameters['additional_mutation'] = \
            float(r[1])
    elif str(r[0]) == 'mutation_type':
        parameters['mutation_type'] = str(r[1])
    elif str(r[0]) == 'chromosome_size':
        parameters['chromosome_size'] = int(r[1])
    elif str(r[0]) == 'genome_size':
        parameters['genome_size'] = int(r[1])
    elif str(r[0]) == 'max_tape_length':
        parameters['max_tape_length'] = int(r[1])
    elif str(r[0]) == 'clean_cell':
        exec("parameters['clean_cell'] = %s" % \
            str(r[1]))
    elif str(r[0]) == 'interpret_chromosome':
        exec("parameters['interpret_chromosome'] = \

```

```

        %s" % str(r[1]))
elif str(r[0]) == 'max_codon':
    parameters['max_codon'] = int(r[1])
elif str(r[0]) == 'population_size':
    parameters['population_size'] = int(r[1])
elif str(r[0]) == 'eco_cell_capacity':
    parameters['eco_cell_capacity'] = int(r[1])
elif str(r[0]) == 'world_x':
    parameters['world_x'] = int(r[1])
elif str(r[0]) == 'world_y':
    parameters['world_y'] = int(r[1])
elif str(r[0]) == 'world_z':
    parameters['world_z'] = int(r[1])
elif str(r[0]) == 'goal':
    try: parameters['goal'] = float(r[1])
    except ValueError:
        exec("parameters['goal'] = %s" % \
            str(r[1]))
elif str(r[0]) == 'maximum_generations':
    parameters['maximum_generations'] = int(r[1])
elif str(r[0]) == 'fossilized_ratio':
    parameters['fossilized_ratio'] = float(r[1])
elif str(r[0]) == 'fossilized_frequency':
    parameters['fossilized_frequency'] = \
        int(r[1])
elif str(r[0]) == 'print_frequency':
    parameters['print_frequency'] = int(r[1])
elif str(r[0]) == 'ragaraja_version':
    version = str(r[1])
    if version == '0.1':
        parameters['ragaraja_version'] = 0.1
    else:
        parameters['ragaraja_version'] = \
            int(r[1])
elif str(r[0]) == 'ragaraja_instructions':
    value = str(r[1]).split('|')
    exec("parameters['ragaraja_instructions'] = \
        %s" % str(value))
elif str(r[0]) == 'eco_buried_frequency':
    parameters['eco_buried_frequency'] = \
        int(r[1])
elif str(r[0]) == 'database_file':
    parameters['database_file'] = str(r[1])
elif str(r[0]) == 'database_logging_frequency':
    parameters['database_logging_frequency'] = \
        int(r[1])
return parameters

```

```

def db_reconstruct_world(cur, start_time, generation):
    '''Function to reconstruct the world object of a simulation
    (as identified by the starting time of the simulation) at a
    specific generation.

    @param cur: Database cursor from connect_database()
    function.
    @param start_time: Starting time of current simulation in
    the format of <date>-<seconds since epoch>; for example,
    2013-10-11-1381480985.77.
    @param generation: Current number of generations simulated.
    @return: dose_world.World object'''
    import dose_world
    eco_cell = {'local_input': [], 'local_output': [],
                'temporary_input': [],
                'temporary_output': [],
                'organisms': 0}
    cur.execute("select max(x), max(y), max(z) from \
world where start_time = '%s'" % start_time)
    coordinates = cur.fetchone()
    world_x = int(coordinates[0]) + 1
    world_y = int(coordinates[1]) + 1
    world_z = int(coordinates[2]) + 1
    ecosystem = {}
    for x in range(world_x):
        ecosystem[x] = {}
        for y in range(world_y):
            ecosystem[x][y] = {}
            for z in range(world_z):
                ecosystem[x][y][z] = {}
                for key in ('local_input',
                            'local_output',
                            'temporary_input',
                            'temporary_output',
                            'organisms'):
                    ecosystem[x][y][z][key] = None
    start_time = str(start_time)
    generation = str(generation)
    # query plan: SCAN TABLE world
    cur.execute("select x, y, z, key, value from \
world where start_time = '%s' and \
generation = '%s'" % (start_time, generation))
    for r in cur.fetchall():
        x = int(r[0])
        y = int(r[1])
        z = int(r[2])
        key = str(r[3])
        value = str(r[4])

```

```

        exec("ecosystem[%i][%i][%i]['%s'] = %s" % \
            (x, y, z, key, value))
    World = dose_world.World(1, 1, 1)
    World.ecosystem = ecosystem
    return World

def db_reconstruct_organisms(cur, start_time,
    population_name, generation):
    '''Function to reconstruct a list of organisms
    (genetic.Organism objects) of a population within a
    simulation (as identified by the starting time of the
    simulation) at a specific generation.

    @param cur: Database cursor from connect_database()
    function.
    @param start_time: Starting time of current simulation in
    the format of <date>-<seconds since epoch>; for example,
    2013-10-11-1381480985.77.
    @param population_name: Name of the population
    @param generation: Current number of generations simulated.
    @return: A list of Organisms (genetic.Organism objects)
    '''
    import genetic as g
    start_time = str(start_time)
    population_name = str(population_name)
    generation = str(generation)
    # query plan: SEARCH TABLE organisms USING INDEX
    #               organisms_index2 (generation=?)
    cur.execute("select distinct org_name from \
        organisms where start_time = '%s' and \
        pop_name = '%s' and generation = '%s'" % \
            (start_time, population_name, generation))
    names = [x[0] for x in cur.fetchall()]
    agents = [0] * len(names)
    for i in range(len(names)):
        org_name = str(names[i])
        org = g.Organism()
        org.genome = []
        org.status['identity'] = org_name
        # query plan: SEARCH TABLE organisms USING INDEX
        #               organisms_index2 (generation=? AND
        #               org_name=?)
        cur.execute("select key, value from organisms \
            where generation = '%s' and org_name = '%s' \
            and start_time = '%s' and pop_name = '%s'" % \
                (generation, org_name, start_time,
                population_name))
        for r in cur.fetchall():

```

```

key = str(r[0])
value = str(r[1])
if key == 'alive':
    exec("org.status['alive'] = %s" % \
        str(value))
elif key == 'vitality':
    org.status['vitality'] = float(value)
elif key == 'parents':
    if value == '': value = '[]'
    else: value = value.split('|')
    exec("org.status['parents'] = %s" % \
        str(value))
elif key == 'age':
    org.status['age'] = float(value)
elif key == 'gender':
    exec("org.status['gender'] = %s" % \
        str(value))
elif key == 'lifespan':
    org.status['lifespan'] = float(value)
elif key == 'fitness':
    exec("f = '%s'" % str(value))
    if type(f) == type(1) or \
        type(f) == type(1.0):
        org.status['fitness'] = float(value)
    else:
        exec("org.status['fitness'] = %s" % \
            str(value))
elif key == 'blood':
    if value == '': value = '[]'
    else: value = value.split('|')
    exec("org.status['blood'] = %s" % \
        str(value))
elif key == 'deme':
    org.status['deme'] = str(value)
elif key == 'location':
    value = tuple([int(x)
                    for x in value.split('|')])
    exec("org.status['location'] = %s" % \
        str(value))
elif key == 'death':
    exec("org.status['death'] = %s" % \
        str(value))
elif key.startswith('chromosome'):
    chr_position = key.split('_')[1]
    sequence = [str(x) for x in str(value)]
    cur.execute("select value from \
        parameters where \
        key='background_mutation' and \

```

```

        start_time='%s' % start_time)
    background_mutation = \
        float(cur.fetchone()[0])
    cur.execute("select value from \
        parameters where \
        key='chromosome_bases' and \
        start_time='%s' % start_time)
    bases = str(cur.fetchone()[0]).split('|')
    exec("chromosome_bases = %s" % bases)
    chromosome = g.Chromosome(sequence,
        chromosome_bases,
        background_mutation)
    org.genome.append(chromosome)
else:
    exec("org.status['%s'] = %s" % \
        (str(key), str(value)))
    agents[i] = org
return agents

def db_reconstruct_population(cur, start_time,
    population_name, generation):
    '''Function to reconstruct a population within a simulation
    (as identified by the starting time of the simulation) at a
    specific generation.

    @param cur: Database cursor from connect_database()
    function.
    @param start_time: Starting time of current simulation in
    the format of <date>-<seconds since epoch>; for example,
    2013-10-11-1381480985.77.
    @param population_name: Name of the population
    @param generation: Current number of generations simulated.
    @return: genetic.Population object'''
    import genetic
    agents = db_reconstruct_organisms(cur, start_time,
        population_name,
        generation)

    cur.execute("select value from parameters where \
        key='goal' and start_time='%s' % start_time)
    g = cur.fetchone()[0]
    try: goal = float(g)
    except ValueError: exec("goal = %s" % str(g))
    return genetic.Population(goal, 1e24, agents)

```

FILE NAME: SIMULATION_CALLS.PY

```
'''File containing support functions for running a
simulation. The functions in this file is not for public
use; all functions in this file are private functions.

Date created: 10th October 2013
'''
import random, inspect, os
import os.path
from datetime import datetime
from time import time
from copy import deepcopy
from shutil import copyfile

try:
    import cPickle as pickle
except ImportError:
    import pickle

import dose_world
import genetic
import ragaraja, register_machine

from database_calls import connect_database
from database_calls import db_log_simulation_parameters
from database_calls import db_report

def simulation_core(sim_functions, sim_parameters,
                   Populations, World):
    '''Sequential ecological cell DOSE simulator.

    Performs the following operations:
        - Creating simulation file directory for results text
          file, population freeze, and world burial storage
        - Generate a simulation start time to identify the
          current simulation
        - Define active Ragaraja instructions
        - Connecting to logging database (if needed)
        - Writing simulation parameters into results text
          file
        - Initialize World and Population
        - Deploy population(s) onto the world
        - Run the simulation and recording the results
        - Writing final results into results text file
        - Close logging database (if used)
        - Copy simulation script into simulation file
          directory
```

```

@param sim_functions: implemented simulation functions (see
sose.dose_functions)
@param sim_parameters: simulation parameters dictionary (see
Examples)
@param Populations: dictionary of population objects
@param World: dose_world.World object'''
    time_start = \
        '-'.join([str(datetime.utcnow()).split(' ')[0],
                    str(time())])
    print('Creating simulation file directories...')
    directory = \
        '-'.join([sim_parameters["simulation_name"],
                    time_start])
    directory = os.sep.join([os.getcwd(), 'Simulations',
                             directory])
    directory = directory + os.sep
    if not os.path.exists(directory):
        os.makedirs(directory)
    print('Adding simulation directory to simulation \
parameters...')
    sim_parameters["directory"] = directory
    print('Adding starting time to simulation \
parameters...')
    sim_parameters["starting_time"] = time_start
    sim_functions = sim_functions()
    if sim_parameters["ragaraja_version"] == 0:
        print('Activating ragaraja version: 0...')
        ragaraja.activate_version(
            sim_parameters["ragaraja_version"],
            sim_parameters["ragaraja_instructions"])
    else:
        print('Activating ragaraja version: ' + \
              str(sim_parameters["ragaraja_version"]) + \
              '...')
        ragaraja.activate_version(
            sim_parameters["ragaraja_version"])
    if "database_file" in sim_parameters and \
        "database_logging_frequency" in sim_parameters:
        print('Connecting to database file: ' + \
              sim_parameters["database_file"] + '...')
        (con, cur) = connect_database(None,
                                       sim_parameters)
        print('Logging simulation parameters to \
database file...')
        (con, cur) = db_log_simulation_parameters(con,
                                                    cur, sim_parameters)
    for pop_name in Populations:
        print('\nPreparing population: ' + pop_name + ' \

```

```

        for simulation...')
if 'sim_folder' in sim_parameters or \
'database_source' in sim_parameters:
    print('Calculating final generation \
count...')
    max = sim_parameters["rev_start"] \
        [sim_parameters["population_names"].
            index(pop_name)] + \
        sim_parameters["extend_gen"]
    print('Updating generation count from \
previous simulation...')
    generation_count = \
        sim_parameters["rev_start"][0]
else:
    print('Deploying population in World \
entity...')
    if sim_parameters["deployment_code"] == 0:
        print('Executing user defined \
deployment scheme...')
        deploy_0(sim_parameters, Populations,
            pop_name, World)
    elif sim_parameters["deployment_code"] == 1:
        print('Executing deployment code 1: \
Single eco-cell deployment...')
        deploy_1(sim_parameters, Populations,
            pop_name, World)
    elif sim_parameters["deployment_code"] == 2:
        print('Executing deployment code 2: \
Random eco-cell deployment...')
        deploy_2(sim_parameters, Populations,
            pop_name, World)
    elif sim_parameters["deployment_code"] == 3:
        print('Executing deployment code 3: \
Even eco-cell deployment...')
        deploy_3(sim_parameters, Populations,
            pop_name, World)
    elif sim_parameters["deployment_code"] == 4:
        print('Executing deployment code 4: \
Centralized eco-cell \
deployment...')
        deploy_4(sim_parameters, Populations,
            pop_name, World)
    print('Adding maximum generations to \
simulation parameters...')
    max = sim_parameters["maximum_generations"]
    generation_count = 0
print('Writing simulation parameters into txt \
file report...')

```

```

        write_parameters(sim_parameters, pop_name)
    print('Updating generation count...')
    Populations[pop_name].generation = \
        generation_count
print('\nSimulation preparation complete...')
while generation_count < max:
    generation_count = generation_count + 1
    sim_functions.ecoregulate(World)
    eco_cell_iterator(World, sim_parameters,
                      sim_functions.update_ecology)
    eco_cell_iterator(World, sim_parameters,
                      sim_functions.update_local)
    eco_cell_iterator(World, sim_parameters,
                      sim_functions.report)
    bury_world(sim_parameters, World,
               generation_count)
    for pop_name in Populations:
        if sim_parameters["interpret_chromosome"]:
            interpret_chromosome(sim_parameters,
                                Populations,
                                pop_name, World)
            report_generation(sim_parameters,
                             Populations, pop_name,
                             sim_functions,
                             generation_count)
            sim_functions.organism_movement(Populations,
                                           pop_name, World)
            sim_functions.organism_location(Populations,
                                           pop_name, World)
        if "database_file" in sim_parameters and \
            "database_logging_frequency" in \
                sim_parameters and \
            generation_count % \
                int(sim_parameters
                    ["database_logging_frequency"]) == 0:
            (con, cur) = db_report(con, cur,
                                   sim_functions,
                                   sim_parameters["starting_time"],
                                   Populations, World,
                                   generation_count)
    print('Generation ' + str(generation_count) + ' \
complete...')
print('\nClosing simulation results...')
for pop_name in Populations:
    close_results(sim_parameters, pop_name)
if "database_file" in sim_parameters and \
    "database_logging_frequency" in sim_parameters:
    print('Committing logged data into database \

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        file...')
    con.commit()
    print('Terminating database connection...')
    con.close()
    print('Copying simulation file script to simulation \
          results directory...')
    sim_script_basename = \
        os.path.basename(inspect.stack()[2][1])
    copyfile(inspect.stack()[2][1],
             os.path.join(sim_parameters['directory'],
                           sim_script_basename))
    print('\nSimulation ended...')

def coordinates(location):
    '''Helper function to transpose ecological cell into a
    tuple.

    @param location: location of ecological cell as 3-element
    iterable data type
    @return: location of ecological cell as (x,y,z)'''
    x = location[0]
    y = location[1]
    z = location[2]
    return (x,y,z)

def adjacent_cells(sim_parameters, location):
    '''Function to get a list of adjacent ecological cells from
    a given location.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param location: location of ecological cell as (x,y,z)
    @return: list of locations of adjacent cells.'''
    trashbin = []
    temp_cells = []
    world_size = [sim_parameters["world_x"],
                  sim_parameters["world_y"],
                  sim_parameters["world_z"]]
    for i in range(3):
        new_location = [spot for spot in location]
        new_location[0] = new_location[0] + 1
        temp_cells.append(new_location)
        new_location = [spot for spot in location]
        new_location[0] = new_location[0] - 1
        temp_cells.append(new_location)
    for i in range(2):
        new_location = [spot for spot in location]
        new_location[1] = new_location[1] - 1

```

```

        temp_cells.append(new_location)
    for i in range(0,4,3):
        temp_cells[i][1] = temp_cells[i][1] + 1
        temp_cells[i+1][1] = temp_cells[i+1][1] - 1
    temp_cells[-1][1] = temp_cells[-1][1] + 2
    for i in range(8):
        for x in range(2):
            if temp_cells[i][x] >= world_size[x] or \
               temp_cells[i][x] < 0:
                if temp_cells[i] not in trashbin:
                    trashbin.append(temp_cells[i])
    for location in trashbin:
        temp_cells.remove(location)
    return [tuple(location) for location in temp_cells]

def spawn_populations(sim_parameters):
    '''Initializing starting population(s) for a simulation.
    Each organism in each population at this stage will be
    genetic clones of each other.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @return: dictionary of population objects with population
    name as key'''
    temp_Populations = {}
    print(' - Accessing population names...')
    for pop_name in sim_parameters["population_names"]:
        print(' - Constructing population: ' + \
              pop_name + '...')
        temp_Populations[pop_name] = \
            genetic.population_constructor(sim_parameters)
        print(' - Updating organism identity and deme \
              status...')
        for individual in temp_Populations
            [pop_name].agents:
            individual.generate_name()
            individual.status['deme'] = pop_name
    return temp_Populations

def eco_cell_iterator(World, sim_parameters, function):
    '''Generic caller to call any function to be executed in
    each ecological cell in sequence.

    @param World: dose_world.World object
    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param function: function to be executed
    @return: none'''

```

```

    for x in range(sim_parameters["world_x"]):
        for y in range(sim_parameters["world_y"]):
            for z in range(sim_parameters["world_z"]):
                if len(inspect.getargspec(function)[0]) \
                    == 5:
                    function(World, x, y, z)
                else:
                    function(World)

def deploy_0(sim_parameters, Populations,
             pop_name, World):
    '''Organism deployment scheme 0 - User defined deployment
    scheme. This is called when "deployment_code" in simulation
    parameters = 0. User will have to provide a deployment
    scheme/function as "deployment_scheme" in simulation
    parameters.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param Populations: dictionary of population objects
    @param pop_name: population name
    @param World: dose_world.World object
    @return: none.'''
    sim_parameters["deployment_scheme"] (Populations,
                                         pop_name, World)

def deploy_1(sim_parameters, Populations,
             pop_name, World):
    '''Organism deployment scheme 1 - Single ecological cell
    deployment scheme. This is called when "deployment_code" in
    simulation parameters = 1. In this scheme, all organisms in
    the specified population (specified by population name) will
    be deployed in the ecological cell
    specified in "population_locations" of simulation
    parameters.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param Populations: dictionary of population objects
    @param pop_name: population name
    @param World: dose_world.World object
    @return: none.'''
    position = \
        sim_parameters["population_names"].index(pop_name)
    locations = [location
                 for location in
                 sim_parameters["population_locations"][position]]
    (x,y,z) = coordinates(locations[0])

```

```

World.ecosystem[x][y][z]['organisms'] = \
    sim_parameters["population_size"]
for individual in Populations[pop_name].agents:
    individual.status['location'] = locations[0]

def deploy_2(sim_parameters, Populations,
             pop_name, World):
    '''Organism deployment scheme 2 - Random deployment scheme.
    This is called when "deployment_code" in simulation
    parameters = 2. In this scheme, all organisms in the
    specified population (specified by population
    name) will be randomly deployed across the list of
    ecological cells specified as "population_locations" of
    simulation parameters.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param Populations: dictionary of population objects
    @param pop_name: population name
    @param World: dose_world.World object
    @return: none'''
    position = \
        sim_parameters["population_names"].index(pop_name)
    locations = [location
                 for location in
                 sim_parameters["population_locations"][position]]
    for individual in Populations[pop_name].agents:
        location = random.choice(locations)
        (x,y,z) = coordinates(location)
        while World.ecosystem[x][y][z]['organisms'] \
            >= sim_parameters["eco_cell_capacity"]:
            location = random.choice(locations)
            (x,y,z) = coordinates(location)
        World.ecosystem[x][y][z]['organisms'] = \
            World.ecosystem[x][y][z]['organisms'] + 1
        individual.status['location'] = location

def deploy_3(sim_parameters, Populations,
             pop_name, World):
    '''Organism deployment scheme 3 - Even deployment scheme.
    This is called when "deployment_code" in simulation
    parameters = 3. In this scheme, all organisms in the
    specified population (specified by population name) will be
    evenly deployed (as much as possible) across the list of
    ecological cells specified as "population_locations" of
    simulation parameters.

```

```

@param sim_parameters: simulation parameters dictionary (see
Examples)
@param Populations: dictionary of population objects
@param pop_name: population name
@param World: dose_world.World object
@return: none'''
    position = \
    sim_parameters["population_names"].index(pop_name)
    locations = [location
        for location in
            sim_parameters["population_locations"][position]]
    iterator = 0
    for i in range(sim_parameters["population_size"]):
        individual = Populations[pop_name].agents[i]
        location = locations[iterator]
        (x,y,z) = coordinates(location)
        World.ecosystem[x][y][z]['organisms'] = \
            World.ecosystem[x][y][z]['organisms'] + 1
        individual.status['location'] = location
        iterator = iterator + 1
        if iterator == len(locations):
            iterator = 0

def deploy_4(sim_parameters, Populations,
            pop_name, World):
    '''Organism deployment scheme 4 - Centralized deployment
    scheme. This is called when "deployment_code" in simulation
    parameters = 4. In this scheme, all organisms in the
    specified population (specified by population name) will be
    deployed onto the ecological cell specified
    in "population_locations" of simulation parameters, up to
    the organism capacity of the ecological cell (specified in
    "eco_cell_capacity" of simulation parameters). In event that
    the specified deployment locations are filled, undeployed
    organisms will be randomly deployed onto adjacent cells.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param Populations: dictionary of population objects
    @param pop_name: population name
    @param World: dose_world.World object
    @return: none'''
    position = \
    sim_parameters["population_names"].index(pop_name)
    locations = [location
        for location in
            sim_parameters["population_locations"][position]]
    adj_cells = adjacent_cells(sim_parameters,

```

```

                                locations[0])
for group in range((sim_parameters
    ["population_size"] // \
    sim_parameters["eco_cell_capacity"]) + 1):
    start = sim_parameters["eco_cell_capacity"] * \
        group
    end = start + sim_parameters["eco_cell_capacity"]
    for x in range(start, end):
        if x == sim_parameters["population_size"]:
            break
        individual = Populations[pop_name].agents[x]
        if x > \
            (sim_parameters["eco_cell_capacity"] - 1):
            location = random.choice(adj_cells)
            (x,y,z) = coordinates(location)
            while World.ecosystem[x][y][z]
                ['organisms'] > \
                    sim_parameters["eco_cell_capacity"]:
                location = random.choice(adj_cells)
                (x,y,z) = \
                    coordinates(random.choice(adj_cells))
            (x,y,z) = coordinates(location)
            World.ecosystem[x][y][z]['organisms'] = \
                World.ecosystem[x][y][z]['organisms'] + 1
            individual.status['location'] = location

def interpret_chromosome(sim_parameters, Populations,
    pop_name, World):
    '''Function to call Ragaraja interpreter to express /
    execute the genome for each organism in a population. The
    Turing tape (array) after execution will be logged as
    "blood" of each organism. Ragaraja interpreter will use the
    temporary_input and temporary_output lists from each
    ecological cell as the input data and output data
    respectively for genome execution - this will resemble
    consumption of environmental resources and replenishing of
    environmental resources or dumping of wastes respectively.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param Populations: dictionary of population objects
    @param pop_name: population name
    @param World: dose_world.World object
    @return: none'''
    array = [0] * sim_parameters["max_tape_length"]
    for i in range(len(Populations[pop_name].agents)):
        individual = Populations[pop_name].agents[i]
        location = individual.status['location']

```

```

(x,y,z) = coordinates(location)
if sim_parameters["clean_cell"]:
    array = [0] * \
        sim_parameters["max_tape_length"]
else:
    array = \
        Populations[pop_name].
        agents[i].status['blood']
    if array == None:
        array = [0] * \
            sim_parameters["max_tape_length"]
for chromosome_count in
    range(len(individual.genome)):
    inputdata = \
        World.ecosystem[x][y][z]['local_input']
    output = \
        World.ecosystem[x][y][z]['local_output']
    source = \
        ''.join(individual.genome
            [chromosome_count].sequence)
    array = \
        Populations[pop_name].
        agents[i].status['blood']
    try:
        (array, apointer, inputdata,
         output, source, spointer) = \
            register_machine.interpret(source,
            ragaraja.ragaraja, 3, inputdata,
            array, sim_parameters["max_tape_length"],
            sim_parameters["max_codon"])
    except Exception as e:
        error_msg = \
            ''.join(['Error at Chromosome_' + \
                str(chromosome_count), str(e)])
        Populations[pop_name].agents[i]. \
            status['chromosome_error'] = \
                error_msg
        Populations[pop_name].agents[i].
            status['blood'] = array
        Populations[pop_name].agents[i].
            status['blood'] = array
        World.ecosystem[x][y][z]['temporary_input'] \
            = inputdata
        World.ecosystem[x][y][z]['temporary_output'] \
            = output

def step(Populations, pop_name, sim_functions):
    '''Performs a generational step for a population

```

```

        - Prepopulation control
        - Mutations
        - Before mating fitness measurement
        - Mating
        - Postpopulation control
        - Generational events
        - After mating fitness measurement
        - Generate a textual report for the current
        generation

@param Populations: dictionary of population objects
@param pop_name: population name
@param sim_functions: implemented simulation functions
                    (see dose.dose_functions)
@return: report as a string'''
    if Populations[pop_name].generation > 0:
        sim_functions.prepopulation_control(Populations,
                                           pop_name)

    for organism in Populations[pop_name].agents:
        sim_functions.mutation_scheme(organism)
    sim_functions.fitness(Populations, pop_name)
    sim_functions.mating(Populations, pop_name)
    sim_functions.postpopulation_control(Populations,
                                       pop_name)

    sim_functions.generation_events(Populations,
                                   pop_name)
    Populations[pop_name].generation = \
        Populations[pop_name].generation + 1
    sim_functions.fitness(Populations, pop_name)
    return sim_functions.population_report(Populations,
                                          pop_name)

def report_generation(sim_parameters, Populations,
                    pop_name, sim_functions, generation_count):
    '''Performs a generational step (using step function) for a
    population and writes out the resulting report into results
    text file.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param Populations: dictionary of population objects
    @param pop_name: population name
    @param sim_functions: implemented simulation functions (see
    dose.dose_functions)
    @param generation_count: current generation count for
    reporting
    @return: none'''
    report = step(Populations, pop_name, sim_functions)

```

```

    if generation_count % \
        int(sim_parameters["fossilized_frequency"]) == 0:
        file = '%s%s_%s_' % (sim_parameters["directory"],
                               sim_parameters["simulation_name"], pop_name)
        freeze_population(file,
                           sim_parameters["fossilized_ratio"],
                           Populations, pop_name)
    if generation_count % \
        int(sim_parameters["print_frequency"]) == 0:
        f = open(('s%s_s%s.result.txt' % \
                  (sim_parameters["directory"],
                     sim_parameters["simulation_name"],
                     pop_name)), 'a')
        dtstamp = str(datetime.utcnow())
        f.write('\n'.join(['\n' + dtstamp,
                           'GENERATION: ' + \
                           str(generation_count),
                           str(report)]))

        f.write('\n')
        f.close()

def bury_world(sim_parameters, World, generation_count):
    '''Function to bury entire world into a file.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param World: dose_world.World object
    @param generation_count: current generation count for file
    name generation'''
    if generation_count % \
        int(sim_parameters["eco_buried_frequency"]) == 0:
        filename = '%s%s_gen%s.eco' % \
            (sim_parameters["directory"],
             sim_parameters["simulation_name"],
             str(generation_count))
        f = open(filename, 'wb')
        pickle.dump(World, f)
        f.close()

def excavate_world(eco_file):
    '''Excavate buried world from file.

    @param eco_file: buried world file generated by bury_world
    function.
    @return: excavated dose_world.World object'''
    f = open(eco_file, 'rb')
    return pickle.load(f)

```

```

def freeze_population(file, proportion,
                      Populations, pop_name):
    '''Function to freeze part or whole of the population into a
    file. If the number of organisms is less than 101, the
    entire population will be frozen.

    @param file: file name prefix
    @param proportion: proportion of the population to freeze
    @param Populations: dictionary of population objects
    @param pop_name: population name to freeze'''
    if proportion > 1.0: proportion = 1.0
    agents = Populations[pop_name].agents
    if len(agents) < 101 or \
        len(agents) * proportion < 101:
        sample = deepcopy(Populations[pop_name])
    else:
        new_agents = [agents[random.randint(0,
                                           len(agents) - 1)]
                      for x in range(int(len(agents) * \
                                         proportion))]
        sample = deepcopy(Populations[pop_name])
        sample.agents = new_agents
    name = ''.join([file, 'pop',
                    str(Populations[pop_name].generation),
                    '_', str(len(sample.agents)), '.gap'])
    f = open(name, 'wb')
    pickle.dump(sample, f)
    f.close()

def revive_population(gap_file):
    '''Revive population from a frozen population file.

    @param gap_file: frozen population file generated by
    freeze_population function.
    @return: revived population object'''
    f = open(gap_file, 'rb')
    return pickle.load(f)

def write_parameters(sim_parameters, pop_name):
    '''Function to write simulation parameters into results text
    file as header.

    @param sim_parameters: simulation parameters dictionary (see
    Examples)
    @param pop_name: population name
    @return: none'''
    f = open('%s%s_%s.result.txt' % \
              (sim_parameters["directory"],

```

```

        sim_parameters["simulation_name"],
        pop_name)), 'a')
f.write("""SIMULATION: %s
-----
""" % sim_parameters["simulation_name"])
if ('database_source' in sim_parameters) or \
    ('sim_folder' in sim_parameters):
    f.write("SIMULATION REVIVAL STARTED: %s\n\n" % \
        sim_parameters["starting_time"])
else:
    f.write("SIMULATION STARTED: %s\n\n" % \
        sim_parameters["starting_time"])
for key in sim_parameters:
    if key not in ('deployment_scheme', 'directory',
        'sim_folder'):
        f.write("%s : %s\n" % (key,
sim_parameters[key]))
f.write("""\n\nREPORT
-----
""")

def close_results(sim_parameters, pop_name):
    '''Function to write footer of results text file.

@param sim_parameters: simulation parameters dictionary (see
Examples)
@param pop_name: population name
@return: none'''
    f = open(('%%s%%s_%%s.result.txt' % \
        (sim_parameters["directory"],
        sim_parameters["simulation_name"],
        pop_name)), 'a')
    f.write('')
    f.write("""
SIMULATION ENDED: ''' + str(datetime.utcnow())
f.close()

```

FILE NAME: GENETIC.PY

```

"""Framework for Genetic Algorithm Applications.
Date created: 23rd February 2010

@see: Lim, JZR, Aw, ZQ, Goh, DJW, How, JA, Low, SXZ, Loo,
BZL, Ling, MHT. 2010. A genetic algorithm framework grounded
in biology. The Python Papers Source Codes 2: 6. """
import random, os, string
from copy import deepcopy

```

```

class Chromosome(object):
    """Representation of a linear chromosome.

    @see: Lim, JZR, Aw, ZQ, Goh, DJW, How, JA, Low, SXZ, Loo,
    BZL, Ling, MHT. 2010. A genetic algorithm framework grounded
    in biology. The Python Papers Source Codes 2: 6.

    @since: version 0.4"""
    def __init__(self, sequence, base,
                  background_mutation=0.0001):
        """Sets up a chromosome.

        @param sequence: a subscriptable object (list or
        string) representing the sequence of the chromosome.
        @param base: a subscriptable object (list or string)
        Representing allowable entities in the sequence.
        @param background_mutation: background mutation rate
        represented as the probability of number of mutations
        per base. Default = 0.0001 (0.01%).

    @since: version 0.4"""
        self.sequence = sequence
        self.base = base
        self.background_mutation = background_mutation

    def rmutate(self, type='point', rate=0.01,
                start=0, end=-1):
        """
        Random Mutation operator - to simulate random point,
        insertion, deletion, inversion, gene translocation
        and gene duplication events.

        The start and end parameters are useful for
        simulating mutational hotspots in the genome.

        @param type: type of mutation. Accepts 'point' (point
        mutation), 'insert' (insert a base), 'delete' (delete
        a base), 'invert' (invert a stretch of the
        chromosome), 'duplicate' (duplicate a stretch of the
        chromosome), 'translocate' (translocate a stretch of
        chromosome to another random position). Default =
        point.
        @param rate: probability of mutation per base above
        Background mutation rate. Default = 0.01 (1%). No
        mutation event will ever happen if (rate +
        background_mutation) is less than zero.
        @param start: starting base on the sequence for
        mutation. Default = 0, start of the genome.

```

```

@param end: last base on the sequence for mutation.
Default = -1, end of the genome.

@since: version 0.4"""
    if end > len(self.sequence) - 1:
        end = len(self.sequence) - 1
    if end == -1: end = len(self.sequence) - 1
    if start == end: start = 0
    length = int(end - start)
    mutation = int((self.background_mutation + rate) \
        * length)
    while mutation > 0:
        position = int(start) + \
            random.randrange(length - 1)
        new_base = \
            self.base[random.randrange(len(self.base))]
        if type == 'point':
            self.sequence[position] = new_base
        if type == 'delete':
            self.sequence.pop(position)
        if type == 'insert':
            self.sequence.insert(position, new_base)
        if type == 'duplicate':
            end_pos = random.randrange(position + 1,
                end)
            fragment = \
                self.sequence[position:end_pos]
            for i in range(len(fragment)):
                self.sequence.insert(end_pos + i,
                    fragment[i])
        if type == 'invert':
            end_pos = random.randrange(position + 1,
                end)
            fragment = [self.sequence.pop(position)
                for i in range(end_pos - position)]
            fragment.reverse()
            for base in fragment:
                self.sequence.insert(position, base)
        if type == 'translocate':
            end_pos = random.randrange(position + 1,
                end)
            fragment = [self.sequence.pop(position)
                for i in range(end_pos - position)]
            insertion_point = random.randint(0,
                len(self.sequence))
            for i in range(len(fragment)):
                self.sequence.insert(insertion_point
                    + i, fragment[i])

```

```

        mutation = mutation - 1

def kmutate(self, type='point',
            start=0, end=0, sequence=None, tpos=0):
    """Known Mutation operator - to simulate a known
    point, insertion, deletion, inversion, gene
    translocation or gene duplication event.

    The required parameters will be determined by the
    type of mutation:
    - point requires
    - start (the position of the base to mutate)
    - sequence (the new base)
    - delete requires
    - start (the position of the base to delete)
    - insert requires
    - start (the position of the base to begin
    insertion)
    - sequence (list of sequence or a base to insert)
    - invert requires
    - start (the position of the base to start
    inversion)
    - end (the position of the base to end inversion)
    - duplicate requires
    - start (the position of the starting base to
    duplicate)
    - end (the position of the last base to
    duplicate)
    - translocate requires
    - start (the position of the base to start
    translocation)
    - end (the position of the base to end
    translocation)
    - tpos (the position of the base insert the
    translocated sequence)

    @param type: type of mutation. Accepts 'point' (point
    mutation), 'insert' (insert a base), 'delete' (delete
    a base), 'invert' (invert a stretch of the
    chromosome), 'duplicate' (duplicate a stretch of the
    chromosome), 'translocate' (translocate a stretch of
    chromosome to another random position). Default =
    point.
    @param start: starting base on the chromosome for
    mutation. Default = 0, start of the genome.
    @param end: last base on the chromosome for mutation.
    Default = 0.
    @param sequence: list of bases to change to (point

```

```

mutation) or to insert (insertion mutation)
@param tpos: position of the chromosome to insert the
Translocated sequence.

```

```

@since: version 0.4"""
    if type == 'point':
        self.sequence[start] = sequence
    if type == 'delete':
        self.sequence.pop(start)
    if type == 'insert':
        for base in list(sequence):
            self.sequence.insert(start, base)
    if type == 'duplicate':
        fragment = self.sequence[start:end + 1]
        for i in range(len(fragment)):
            self.sequence.insert(end + i,
                                fragment[i])
    if type == 'invert':
        for base in self.sequence[start:end]:
            self.sequence.insert(start, base)
    if type == 'translocate':
        fragment = [self.sequence.pop(start)
                    for i in range(end - start)]
        for i in range(len(fragment)):
            self.sequence.insert(tpos + i,
                                fragment[i])

def replicate(self):
    """Replicates (deep copy) the chromosome.

    @return: a copy of current chromosome.

    @since: version 0.4"""
    return deepcopy(self)

```

```

class Organism(object):
    """An organism represented by a list of chromosomes and a
    status table. This class should almost never be instantiated
    on its own but acts as an ancestor class as some functions
    need to be over-ridden in the inherited class for the
    desired use.

```

Each organism is identifiable by a randomly generated 32-character 'name' as `Organism.identity`.

Methods to be over-ridden in the inherited class or substituted are

```
- fitness
- mutation_scheme
```

Pre-defined status are

1. alive - is the organism alive? True or False.
2. vitality - percentage of maximum vitality.
3. age - the current age of the organism.
4. lifespan - pre-defined maximum lifespan to be set based on scenario.
5. fitness - how fit the organism is? Set to maximum fitness.
6. death - reason of death (as death code).
7. parents - identity of parents.
8. gender - gender of organism.
9. blood - result of genomic interpretation or expression by Rajaraga interpreter.
10. identity - 32-character randomly generated name.
11. deme - defined as a sub-population or local population.
12. location - location of the organism within the ecosystem.

List of defined death codes

1. death01 - zero vitality
2. death02 - maximum age reached
3. death03 - unknown death cause

@see: Lim, JZR, Aw, ZQ, Goh, DJW, How, JA, Low, SXZ, Loo, BZL, Ling, MHT. 2010. A genetic algorithm framework grounded in biology. The Python Papers Source Codes 2: 6.

@since: version 0.4"""

```
def __init__(self, genome='dummy',
             mutation_type='point',
             additional_mutation_rate=0.01,
             gender=None):
    """Sets up a new organism with default status (age =
    0, vitality = 100, lifespan = 100, fitness = 100,
    alive = True)

    @param genome: list of chromosomes to inherit.
    Default = 'dummy'. This creates a 'dummy' chromosome
    which is basically a one-base chromosome - this is
    for applications which does not utilize the
    chromosome.
    @param mutation_type: type of mutation. Accepts
```

```

'point' (point mutation), 'insert' (insert a base),
'delete' (delete a base), 'invert' (invert a stretch
of the chromosome), 'duplicate' (duplicate a stretch
of the chromosome), 'translocate' (translocate a
stretch of chromosome to another random position).
Default = point.
@param additional_mutation_rate: probability of
mutation per base above background mutation rate.
Default = 0.01 (1%). No mutation event will ever
happen if (rate + background_mutation) is less than
zero.
@param gender: establishes the gender of the organism
which may be used for mating routines.

@since: version 0.4"""
    self.status = {'alive': True,
                    # is the organism alive?
                    'vitality': 100.0,
                    # % of vitality
                    'parents': None,
                    # identity of parent(s)
                    'age': 0.0,
                    # age of the organism
                    'gender': None,
                    # gender of organism
                    'lifespan': 100.0,
                    # maximum lifespan
                    'fitness': 100.0,
                    # % of fitness
                    'blood': None,
                    # interpreted chromosome
                    'identity': None,
                    # name of the organism
                    'deme': None,
                    # sub-population or race
                    'location': None,
                    # location of the organism
                    'death': None}

    if genome == 'dummy':
        self.genome = [Chromosome([0], [0])]
    else:
        self.genome = genome
        self.mutation_type = mutation_type
        self.additional_mutation_rate = \
            additional_mutation_rate
        self.status['gender'] = gender

def generate_name(self):

```

```

name = ''.join([random.choice(('1', '2', '3',
                                '4', '5', '6', '7', '8', '9', 'A', 'B',
                                'C', 'D', 'E', 'F', 'G', 'H', 'I', 'J',
                                'K', 'L', 'M', 'N', 'O', 'P', 'Q', 'R',
                                'S', 'T', 'U', 'V', 'W', 'X', 'Y', 'Z',
                                'a', 'b', 'c', 'd', 'e', 'f', 'g', 'h',
                                'i', 'j', 'k', 'l', 'm', 'n', 'o', 'p',
                                'q', 'r', 's', 't', 'u', 'v', 'w', 'x',
                                'y', 'z', '=', '#', '$', '%', '&', '@',
                                '<', '>', '?')) for x in range(32)])
self.status['identity'] = name

def fitness(self):
    """Function to calculate the fitness of the current
    organism. B{This function MUST be over-ridden by the
    inherited class or substituted as fitness function is
    highly dependent on utility.} The only requirement is
    that a fitness score must be returned.

    Here, the sample implementation calculates fitness as
    proportion of the genome with '1's.

    @return: fitness score or fitness list

    @since: version 0.4"""
    one_count = sum([sum(chromosome.sequence)
                     for chromosome in self.genome])
    length_of_genome = sum([len(chromosome.sequence)
                             for chromosome in self.genome])
    return float(one_count) / float(length_of_genome)

def mutation_scheme(self, type=None, rate=None):
    """Function to trigger mutation events in each
    chromosome. B{This function may be over-ridden by the
    inherited class or substituted to cater for specific
    mutation schemes but not an absolute requirement to
    do so.}

    Both type and rate must be defined at the same time,
    otherwise the initiated mutation_type and
    additional_mutation_rate will be used.

    @param type: type of mutation. Accepts 'point' (point
    mutation), 'insert' (insert a base), 'delete' (delete
    a base), 'invert' (invert a stretch of the
    chromosome), 'duplicate' (duplicate a stretch of the
    chromosome), 'translocate' (translocate a stretch of
    chromosome to another random position). Default =

```

```

None.
@param rate: probability of mutation per base above
background mutation rate. Default = None. No mutation
event will ever happen if (rate +
background_mutation) is less than zero.

@since: version 0.4"""
    for chromosome in self.genome:
        if not type and not rate:
            chromosome.rmutate(self.mutation_type,
                               self.additional_mutation_rate)
        if type and rate:
            chromosome.rmutate(type, rate)

def setStatus(self, variable, value):
    """Sets new status or change status of the organism.
    However, the following status change will result in
    death of the organism
    1. 'alive' to False
    2. 'vitality' to or below zero
    3. 'age' to or above lifespan

@param variable: name of status to change
@param value: new value of the status

@since: version 0.4"""
    if variable == 'alive' and value == True:
        self.status['alive'] = value
    if variable == 'alive' and value == False:
        self.status['alive'] = value
        self.status['death'] = 'death03'
    if variable == 'vitality':
        if float(value) > 100.1:
            self.status['vitality'] = 100.0
        if float(value) > 0.0 and \
            float(value) < 100.1:
            self.status['vitality'] = float(value)
        if float(value) == 0 or float(value) < 0:
            self.status['vitality'] = 0
            self.status['alive'] = False
            self.status['death'] = 'death01'
    elif variable == 'age':
        if float(value) < self.status['lifespan']:
            self.status['age'] = float(value)
        else:
            self.status['age'] = \
                self.status['lifespan']
            self.status['alive'] = False

```

```

        self.status['death'] = 'death02'
    else:
        self.status[variable] = value

    def getStatus(self, variable):
        """Returns a status variable of the current organism

        @param variable: name of the status variable
        @return: status or a KeyError if status is not found

        @since: version 0.4"""
        try:
            return self.status[variable]
        except:
            raise KeyError('%s not found in organism \
                           status' % str(variable))

    def __str__(self):
        """Returns the genome of the organism"""
        return str([chromosome.sequence
                    for chromosome in self.genome])

    def clone(self):
        """Clones (deep copy) the organism.

        @return: a copy of current organism.

        @since: version 0.4"""
        org = deepcopy(self)
        return org

class Population(object):
    """Representation of a population as a list of organisms.
    The entire population is in memory.

    Methods to be over-ridden in the inherited class or
    substituted are
        - prepopulation_control
        - mating
        - postpopulation_control
        - generation_events
        - report

    @see: Lim, JZR, Aw, ZQ, Goh, DJW, How, JA, Low, SXZ, Loo,
    BZL, Ling, MHT. 2010. A genetic algorithm framework grounded
    in biology. The Python Papers Source Codes 2: 6.

    @since: version 0.4"""

```

```

def __init__(self, goal,
              maxgenerations='infinite', agents=[]):
    """Establishes a population of organisms.

    @param goal: the goal to be reached by the
    population.
    @type goal: the return type of Organism.fitness()
    @param maxgenerations: maximum number of generations
    to evolve. Default = 'infinite'.
    @param agents: organisms making up the initial
    population.
    @type agents: list of Organism objects

    @since: version 0.4"""
    self.agents = agents
    self.goal = goal
    self.maxgenerations = maxgenerations
    self.generation = 0

def prepopulation_control(self):
    """Function to trigger population control events
    before mating event in each generation (For example,
    to simulate pre-puberty death). B{This function may
    be over-ridden by the inherited class or substituted
    to cater for specific events.) Although this is not
    an absolute requirement, it is extremely encouraged
    to prevent exhaustion of memory space. Without
    population control, it will seem like a reproducing
    immortal population.

    Here, the sample implementation eliminates the bottom
    half of the population based on fitness score unless
    the number of organism is less than 20. This is to
    prevent extinction. However, if the number of
    organism is more than 2000 (more than 100x initial
    population size, a random selection of 2000 will be
    used for the next generation.

    @since: version 0.4"""
    size = len(self.agents)
    sfitness = [self.agents[x].fitness()
                 for x in range(size)]
    threshold = sum(sfitness) / len(sfitness)
    temp = [self.agents[x]
             for x in range(size)
             if sfitness[x] > threshold]
    if len(temp) > 2001:

```

```

        size = len(temp)
        self.agents = [temp[random.randint(0,
                                           size - 1)]
                       for x in range(2000)]
    if len(temp) > 21:
        self.agents = temp

def mating(self):
    """Function to trigger mating events in each
    generation. B{This function may be over-ridden by the
    inherited class or substituted to cater for specific
    mating schemes but not an absolute requirement to do
    so.} Mating schemes should include
        - selection of mating partners using
          Organism.fitness() function and/or other status
        - processes and actions of mating

    Here, the sample implementation randomly selects any
    2 organisms from the culled list and generate a new
    genome for the progeny organism by single crossover.

    @since: version 0.4"""
    size = len(self.agents)
    temp = []
    for x in range(size):
        organism1 = self.agents[random.randint(0,
                                                size - 1)]
        organism2 = self.agents[random.randint(0,
                                                size - 1)]
        crossover_pt = random.randint(0,
                                      len(organism1.genome[0].sequence))
        (g1, g2) = crossover(organism1.genome[0],
                             organism2.genome[0],
                             crossover_pt)
        temp = temp + [Organism([g1])]
    self.add_organism(temp)

def postpopulation_control(self):
    """Function to trigger population control events
    after mating event in each generation(For example, to
    simulate old-age death). B{This function may be over-
    ridden by the inherited class or substituted to cater
    for specific events.} Although this is not an
    absolute requirement, it is extremely encouraged to
    prevent exhaustion of memory space. Without
    population control, it will seem like a reproducing
    immortal population.

```

```
@since: version 0.4"""
    pass

def generation_events(self):
    """Function to trigger other defined events in each
    generation. B{This function may be over-ridden by the
    inherited class or substituted to cater for specific
    events but not an absolute requirement to do so.}
    Events and controls may include
        - processes simulating disaster or other
        catastrophic events
        - changes in mutations

@since: version 0.4"""
    pass

def report(self):
    """Function to report the status of each generation.
    B{This function may be over-ridden by the inherited
    class or substituted to cater for specific reporting
    schemes but not an absolute requirement to do so.} At
    the very least, this function should report whether
    the goal is reached.

@return: dictionary of status describing the current
generation

@since: version 0.4"""
    sfitness = [self.agents[x].fitness()
                 for x in range(len(self.agents))]
    afitness = sum(sfitness) / \
                float(len(self.agents))
    return {'generation': self.generation,
            'average fitness': afitness,
            '% to goal': float(afitness - self.goal)\
                           / self.goal * 100}

def generation_step(self):
    """Function to simulate events for one generation.
    These includes
        - mating (according to the mating scheme or
        function)
        - mutating each organism in the population
        - population size control
        - other events defined under generation_events
        function
        - increment of generation count
        - reporting the population status
```

```

    @return: information returned from report function.

    @since: version 0.4"""
    if self.generation > 0:
        self.prepopulation_control()
    self.mating()
    self.postpopulation_control()
    for organism in self.agents:
        organism.mutation_scheme()
    self.generation_events()
    self.generation = self.generation + 1
    return self.report()

def add_organism(self, organism):
    """Add a new organism(s) to the population.

    @param organism: list of new Organism object(s)

    @since: version 0.4"""
    self.agents = self.agents + organism

def freeze(self, prefix='pop', proportion=0.01):
    """Preserves part or the entire population. If the
    population size or the preserved proportion is below
    100, the entire population will be preserved. The
    preserved sample will be written into a file with
    name in the following format - <prefix><generation
    count>_<sample size>.gap

    @param prefix: prefix of file name. Default = 'pop'.
    @param proportion: proportion of population to be
    preserved. Default = 0.01, preserves 1% of the
    population.

    @since: version 0.4"""
    try:
        import cPickle as pickle
    except ImportError:
        import pickle

    if proportion > 1.0: proportion = 1.0
    if len(self.agents) < 101 or \
        len(self.agents) * proportion < 101:
        sample = self.agents
    else:
        size = len(self.agents)
        sample = [self.agents[random.randint(0,
```

```

        size - 1)]
    for x in range(int(len(self.agents)
        * proportion))]
    name = ''.join([prefix, str(self.generation),
        '_', str(len(sample)), '.gap'])
    f = open(name, 'wb')
    pickle.dump(sample, f)
    f.close()

def revive(self, filename, type='replace'):
    """Revives a frozen population.

    @param filename: file name of frozen population
    (generated by Population.freeze() function)
    @param type: type of revival. Allows 'replace'
    (replace the current population with the revived
    population) or 'add' (add the revived population to
    the current population). Default = 'replace'.

    @since: version 0.4"""
    try:
        import cPickle as pickle
    except ImportError:
        import pickle
    if type == 'replace':
        self.agents = \
            pickle.load(open(filename, 'rb'))
    if type == 'add':
        self.agents = self.agents + \
            pickle.load(open(filename, 'rb'))

def crossover(chromosome1, chromosome2, position):
    """Cross-over operator - swaps the data on the 2 given
    chromosomes after a given position.

    @param chromosome1: Chromosome object
    @param chromosome2: Chromosome object
    @param position: base position of the swap over
    @type position: integer
    @return: (resulting chromosome1, resulting chromosome2)

    New chromosomes inherit their parent's crossed sequences,
    bases and background mutation rate.

    @since: version 0.4"""
    seq1 = chromosome1.sequence
    seq2 = chromosome2.sequence
    position = int(position)

```

```

    if len(seq1) > position and len(seq2) > position:
        new1 = Chromosome(seq1[:position] + \
                           seq2[position:],
                           chromosome1.base,
                           chromosome1.background_mutation)
        new2 = Chromosome(seq2[:position] + \
                           seq1[position:],
                           chromosome2.base,
                           chromosome2.background_mutation)
        return (new1, new2)
    elif len(seq1) > position:
        new1 = Chromosome(seq1[:position],
                           chromosome1.base,
                           chromosome1.background_mutation)
        new2 = Chromosome(chromosome2.sequence + \
                           seq1[position:],
                           chromosome2.base,
                           chromosome2.background_mutation)
        return (new1, new2)
    elif len(seq2) > position:
        new1 = Chromosome(chromosome1.sequence + \
                           seq2[position:],
                           chromosome1.base,
                           chromosome1.background_mutation)
        new2 = Chromosome(seq2[:position],
                           chromosome2.base,
                           chromosome2.background_mutation)
        return (new1, new2)
    else:
        return (chromosome1, chromosome2)

population_data = \
{
    'chromosome_bases' : [1, 2, 3, 4],
    'chromosome_length' : 200,
    'chromosome_type' : 'defined',
    'initial_chromosome' : [1] * 200,
    'background_mutation' : 0.0001,
    'genome_size' : 1,
    'population_size' : 200,
    'fitness_function' : 'default',
    'mutation_scheme' : 'default',
    'additional_mutation' : 0.01,
    'mutation_type' : 'point',
    'goal' : 4,
    'maximum_generations' : 'infinite',
    'prepopulation_control' : 'default',
    'mating' : 'default',

```

```

    'postpopulation_control' : 'default',
    'generation_events' : 'default',
    'population_report' : 'default'
}

def population_constructor(data=population_data):
    """Function to construct a population based on a dictionary
    of population data.

    Population data contains the following keys:
    - 'chromosome_bases' = List of allowable bases.
      Default = [1, 2, 3, 4].
    - 'chromosome_length' = Length of a chromosome.
      Default = 200.
    - 'chromosome_type' = Type of chromosome. Default =
      'defined'.
    - 'initial_chromosome' = Initial chromosome. Default
      = [1] * 200.
    - 'background_mutation' = Background mutation rate.
      Default = 0.0001 (0.01%).
    - 'genome_size' = Number of chromosomes per organism.
      Default = 1.
    - 'population_size' = Size of initial population
      (number of organisms). Default = 200.
    - 'fitness_function' = Fitness evaluation function.
      Accepts 'default' or a function. Please refer to
      Organism. Default = 'default'.
    - 'mutation_scheme' = Function simulating the
      mutation scheme of an organism. Accepts 'default' or
      a function. Please refer to Organism. Default =
      'default'.
    - 'additional_mutation' = Mutation rate on top of
      background mutation rate. Default = 0.01 (1%).
    - 'mutation_type' = Type of default mutation. Default
      = 'point'.
    - 'goal' = Goal of the population, as evaluated by
      fitness function. Default = 4.
    - 'maximum_generations' = Number of generations to
      simulate. Accepts an integer or 'infinite'. Default =
      'infinite'.
    - 'prepopulation_control' = Function simulating pre-
      mating population control. Please refer to
      Population. Default = 'default'.
    - 'mating' = Function simulating mating procedure
      (mate selection and act of mating control. Please
      refer to Population. Default = 'default'.
    - 'postpopulation_control' = Function simulating
      post-mating population control. Please refer to

```

```

        Population.Default = 'default'.
        - 'generation_events' = Function simulating other
        possible (usually rare or random) events in the
        generation. Please refer to Population.Default =
        'default'.
        - 'population_report' = Function to generate the
        status report of the generation. Please refer to
        Population.Default = 'default'.

@param data: population data
@type data: dictionary

@return: Population object

@see: Lim, JZR, Aw, ZQ, Goh, DJW, How, JA, Low, SXZ, Loo,
BZL, Ling, MHT. 2010. A genetic algorithm framework grounded
in biology. The Python Papers Source Codes 2: 6.

@since: version 0.4"""
    chr = Chromosome(data['initial_chromosome'],
                     data['chromosome_bases'],
                     data['background_mutation'])
    org = Organism([chr]*data['genome_size'],
                  data['mutation_type'],
                  data['additional_mutation'])
    org_set = [org.clone()
               for x in range(data['population_size'])]
    pop = Population(data['goal'],
                    int(data['maximum_generations']),
                    org_set)

    return pop

def population_simulate(population,
                       printfreq=100,
                       freezefreq='never',
                       freezefile='pop',
                       freezeportion=0.01,
                       resultfile='result.txt'):
    """Function to simulate the population - start the GA.

    @param population: Population to run.
    @type population: Population object
    @param printfreq: Reporting intervals on screen. Default =
    100.
    @param freezefreq: Generation intervals to freeze population
    into a file. See Population.freeze method. Accepts an
    integer or 'never'. Default = 'never'.

```

@param freezefile: Prefix of file name of frozen population. See Population.freeze method. Default = 'pop'.
 @param freezeportion: Proportion of population to be preserved. See Population.freeze method. Default = 0.01, preserves 1% of the population.
 @param resultfile: Name of file to print out results of each generation. Format of output is dependent on reporting method of the population (Population.report). If 'None', it will print out the results into a file. Default = 'result.txt'.

@see: Lim, JZR, Aw, ZQ, Goh, DJW, How, JA, Low, SXZ, Loo, BZL, Ling, MHT. 2010. A genetic algorithm framework grounded in biology. The Python Papers Source Codes 2: 6.

@since: version 0.4

```
"""
    if freezefreq == 'never': freezefreq = int(12e14)
    if resultfile != 'None':
        result = open(resultfile, 'w')
    report = population.generation_step()
    reportitems = ['|'.join([str(key), str(report[key])])
                  for key in list(report.keys())]
    result.writelines('|'.join(reportitems))
    result.writelines(os.linesep)
    while population.generation < \
        population.maxgenerations:
        if resultfile != 'None':
            report = population.generation_step()
            reportitems = ['|'.join([str(key),
                                    str(report[key])])
                          for key in list(report.keys())]
            result.writelines('|'.join(reportitems))
        if population.generation % int(printfreq) == 0:
            print('|'.join(reportitems))
        result.writelines(os.linesep)
        if population.generation % int(freezefreq) == 0:
            population.freeze(freezefile,
                             freezeportion)
    population.freeze(freezefile, 1.0)
    result.close()
```

EXAMPLES OF SIMULATION FILE FROM CASTILLO AND LING (2014A)

FILE NAME: RUN_EXAMPLES_WITHOUT_INSTALLATION.PY

```
'''This file contains inserts the path for DOSE into system
path to allow for any examples importing this file to
execute without needing DOSE to be installed into into
Python site-packages.'''
```

```
import sys
import os
```

```
cwd = os.getcwd().split(os.sep)
#cwd[-1] = 'dose'
cwd = os.sep.join(cwd[:-1])
```

```
sys.path.append(cwd)
```

```
import dose
```

FILE NAME: 09_NO_MIGRATION_ISOLATED_MATING.PY

```
'''Example 09a: Examining the effects of no migration on
genetic distance differences from an initially identical
population (development of sub-populations or demes which
may lead to speciation)
```

```
This example is identical to Example 03, except background
mutation rate is changed from 10% in Example 03 to 0.1% in
this example.
```

```
In this simulation,
- 1 populations of 1250 organisms
- each organism will have 1 chromosome of only 2
bases (1 and 0)
- Evenly deployed across 25 eco-cells (50 organism
per eco-cell)
- 0.1% background point mutation on chromosome of 50
bases
- no organism movement throughout the simulation
- no Ragaraja interpretation of genome
- 1000 generations to be simulated'''
```

```
# needed to run this example without prior
# installation of DOSE into Python site-packages
```

```
try:
    import run_examples_without_installation
except ImportError: pass
```

```
# Example codes starts from here
```

```
import dose, random
```

```
parameters = \
```

```
{
    "simulation_name": "09_no_migration_isolated_mating",
    "population_names": ['pop_01'],
    "population_locations": [[(x,y,z)
                               for x in range(5)
                               for y in range(5)
                               for z in range(1)]]],

    "deployment_code": 3,
    "chromosome_bases": ['0','1'],
    "background_mutation": 0.001,
    "additional_mutation": 0,
    "mutation_type": 'point',
    "chromosome_size": 5000,
    "genome_size": 1,
    "max_tape_length": 50,
    "clean_cell": True,
    "interpret_chromosome": False,
    "max_codon": 2000,
    "population_size": 1250,
    "eco_cell_capacity": 50,
    "world_x": 5,
    "world_y": 5,
    "world_z": 1,
    "goal": 0,
    "maximum_generations": 1000,
    "fossilized_ratio": 0.01,
    "fossilized_frequency": 100,
    "print_frequency": 10,
    "ragaraja_version": 0,
    "ragaraja_instructions": ['000', '001', '010',
                              '011', '100', '101'],
    "eco_buried_frequency": 1250,
    "database_file": "sim09_no_migration.db",
    "database_logging_frequency": 1}
}
```

```
class simulation_functions(dose.dose_functions):

    def organism_movement(self, Populations,
                          pop_name, World): pass
    def organism_location(self, Populations,
                          pop_name, World): pass
    def ecoregulate(self, World): pass
    def update_ecology(self, World, x, y, z): pass
    def update_local(self, World, x, y, z): pass
    def report(self, World): pass
    def fitness(self, Populations, pop_name): pass

    def mutation_scheme(self, organism):
        organism.genome[0].rmutate(
```

```

        parameters["mutation_type"],
        parameters["additional_mutation"])

def prepopulation_control(self, Populations,
    pop_name): pass

def mating(self, Populations, pop_name):
    for location in parameters
        ["population_locations"][0]:
        group = dose.filter_location(location,
            Populations[pop_name].agents)
        for x in range(len(group)//2):
            parents = []
            for i in range(2):
                parents.append(random.choice(
                    Populations[pop_name].agents))
                while parents[i] not in group:
                    parents[i] = \
                        random.choice(
                            Populations[pop_name].agents)
                Populations[pop_name].agents.

                remove(parents[i])
            crossover_pt = \
                random.randint(0,
                    len(parents[0].genome[0].sequence))
            (new_chromol, new_chromo2) = \
                dose.genetic.crossover(
                    parents[0].genome[0],
                    parents[1].genome[0], crossover_pt)
            children = \
                [dose.genetic.Organism([new_chromol],
                    parameters["mutation_type"],
                    parameters["additional_mutation"]),
                    dose.genetic.Organism([new_chromo2],
                    parameters["mutation_type"],
                    parameters["additional_mutation"])]
            for child in children:
                child.status['parents'] = \
                    [parents[0].status['identity'],
                    parents[1].status['identity']]
                child.status['location'] = location
                child.generate_name()
                child.status['deme'] = pop_name
                Populations[pop_name].
                    agents.append(child)

def postpopulation_control(self, Populations,
```

```

        pop_name): pass
def generation_events(self, Populations,
    pop_name): pass

def population_report(self, Populations, pop_name):
    report_list = []
    for organism in Populations[pop_name].agents:
        identity = str(organism.status['identity'])
        report_list.append(identity)
    return '\n'.join(report_list)

def database_report(self, con, cur, start_time,
    Populations, World,
    generation_count):

    try:
        dose.database_report_populations(con, cur,
            start_time, Populations, generation_count)
    except: pass
    try:
        dose.database_report_world(con, cur,
            start_time, World, generation_count)
    except: pass

def deployment_scheme(self, Populations,
    pop_name, World): pass

dose.simulate(parameters, simulation_functions)

```

FILE NAME: 09_ADJACENT_MIGRATION_ISOLATED_MATING.PY

'''Example 09b: Extending from Example 09a (as baseline) to examine the effects of short migration (adjacent cell migration) on genetic distance differences from an initially identical population (development of sub-populations or demes which may lead to speciation)

This example is identical to Example 04, except background mutation rate is changed from 10% in Example 03 to 0.1% in this example.

In this simulation,

- 1 populations of 1250 organisms
- each organism will have 1 chromosome of only 2 bases (1 and 0)
- Evenly deployed across 25 eco-cells (50 organism per eco-cell)
- 0.1% background point mutation on chromosome of 50 bases

```

- 10% organism movement per eco-cell per generation
throughout the simulation
- the same organism may move twice due to sequential
evaluation of the eco-cells but the probability of
such event will be 10% x 10% = 1%; similarly, 3 or
more movement by the same organism may happen with
reducing probabilities
- no Ragaraja interpretation of genome
- 1000 generations to be simulated'''
# needed to run this example without prior
# installation of DOSE into Python site-packages
try:
    import run_examples_without_installation
except ImportError: pass

# Example codes starts from here
import dose, random

parameters = \
{
    "simulation_name": \
        "09_adjacent_migration_isolated_mating",
    "population_names": ['pop_01'],
    "population_locations": [[(x,y,z)
                                for x in range(5)
                                for y in range(5)
                                for z in range(1)]],
    "deployment_code": 3,
    "chromosome_bases": ['0','1'],
    "background_mutation": 0.001,
    "additional_mutation": 0,
    "mutation_type": 'point',
    "chromosome_size": 5000,
    "genome_size": 1,
    "max_tape_length": 50,
    "clean_cell": True,
    "interpret_chromosome": False,
    "max_codon": 2000,
    "population_size": 1250,
    "eco_cell_capacity": 50,
    "world_x": 5,
    "world_y": 5,
    "world_z": 1,
    "goal": 0,
    "maximum_generations": 1000,
    "fossilized_ratio": 0.01,
    "fossilized_frequency": 100,
    "print_frequency": 10,
    "ragaraja_version": 0,

```

```

"ragaraja_instructions": ['000', '001', '010',
                           '011', '100', '101'],
"eco_buried_frequency": 1250,
"database_file": "sim09_adjacent.db",
"database_logging_frequency": 1}

class simulation_functions(dose.dose_functions):

    def organism_movement(self, Populations,
                          pop_name, World):
        for location in parameters
            ["population_locations"][0]:
                group = dose.filter_location(location,
                    Populations[pop_name].agents)
                adj_cells = dose.simulation_calls.
                    adjacent_cells(parameters, location)
                for i in range(int(round((len(group) \
                    * 0.1)))):
                    (x,y,z) = dose.simulation_calls. \
                        coordinates(location)
                    World.ecosystem[x][y][z]['organisms'] =\
                    World.ecosystem[x][y][z]['organisms'] - 1
                    immigrant = random.choice(
                        Populations[pop_name].agents)
                    while immigrant not in group:
                        immigrant = random.choice(
                            Populations[pop_name].agents)
                    new_location = random.choice(adj_cells)
                    immigrant.status['location'] = \
                        new_location
                    (x,y,z) = dose.simulation_calls. \
                        coordinates(new_location)
                    World.ecosystem[x][y][z]['organisms'] =\
                    World.ecosystem[x][y][z]['organisms'] + 1

    def organism_location(self, Populations,
                          pop_name, World): pass
    def ecoregulate(self, World): pass
    def update_ecology(self, World, x, y, z): pass
    def update_local(self, World, x, y, z): pass
    def report(self, World): pass
    def fitness(self, Populations, pop_name): pass

    def mutation_scheme(self, organism):
        organism.genome[0].rmutate(
            parameters["mutation_type"],
            parameters["additional_mutation"])

```

```

def prepopulation_control(self, Populations,
    pop_name): pass

def mating(self, Populations, pop_name):
    for location in parameters
        ["population_locations"][0]:
        group = dose.filter_location(location,
            Populations[pop_name].agents)
        for x in range(len(group)//2):
            parents = []
            for i in range(2):
                parents.append(
                    random.choice(
                        Populations[pop_name].agents))
                while parents[i] not in group:
                    parents[i] = random.choice(
                        Populations[pop_name].agents)
                Populations[pop_name].agents.
                    remove(parents[i])
            crossover_pt = random.randint(0,
                len(parents[0].genome[0].sequence))
            (new_chromol, new_chromo2) = \
                dose.genetic.crossover(
                    parents[0].genome[0],
                    parents[1].genome[0], crossover_pt)
            children = [dose.genetic.Organism(
                [new_chromol],
                parameters["mutation_type"],
                parameters["additional_mutation"]),
                dose.genetic.Organism(
                    [new_chromo2],
                    parameters["mutation_type"],
                    parameters["additional_mutation"])]
            for child in children:
                child.status['parents'] = \
                    [parents[0].status['identity'],
                     parents[1].status['identity']]
                child.status['location'] = location
                child.generate_name()
                child.status['deme'] = pop_name
                Populations[pop_name].
                    agents.append(child)

def postpopulation_control(self, Populations,
    pop_name): pass
def generation_events(self, Populations,
    pop_name): pass

```

```

def population_report(self, Populations, pop_name):
    report_list = []
    for organism in Populations[pop_name].agents:
        identity = str(organism.status['identity'])
        report_list.append(identity)
    return '\n'.join(report_list)

def database_report(self, con, cur, start_time,
                   Populations, World,
                   generation_count):
    try:
        dose.database_report_populations(con, cur,
                                         start_time, Populations, generation_count)
    except: pass
    try:
        dose.database_report_world(con, cur,
                                   start_time, World, generation_count)
    except: pass

def deployment_scheme(self, Populations,
                      pop_name, World): pass

dose.simulate(parameters, simulation_functions)

```

FILE NAME: 09_LONG_MIGRATION_ISOLATED_MATING.PY

'''Example 09c: Extending from Example 09a (as baseline) and Example 09b (short distance migration) to examine the effects of long distance migration (across one or more eco-cells) on genetic distance differences from an initially identical population (development of sub-populations or demes which may lead to speciation)

This example is identical to Example 05, except background mutation rate is changed from 10% in Example 03 to 0.1% in this example.

In this simulation,

- 1 populations of 1250 organisms
- each organism will have 1 chromosome of only 2 bases (1 and 0)
- Evenly deployed across 25 eco-cells (50 organism per eco-cell)
- 20% background point mutation on chromosome of 50 bases
- 10% organism movement per eco-cell per generation throughout the simulation
- the same organism may move twice due to sequential

```

        evaluation of the eco-cells but the probability of
        such event will be 10% x 10% = 1%; similarly, 3 or
        more movement by the same organism may happen with
        reducing probabilities
        - the destination eco-cell for movement is random;
        thus, from 25 possible eco-cells, there is a
        probability of 4% chance of relocating back to the
        same (original) eco-cell, 16% chance of relocating to
        one of the 4-neighbour eco-cells and 32% chance of
        relocating to one of the 8-neighbour eco-cells
        - no Ragaraja interpretation of genome
        - 1000 generations to be simulated'''
# needed to run this example without prior
# installation of DOSE into Python site-packages
try:
    import run_examples_without_installation
except ImportError: pass

# Example codes starts from here
import dose, random

parameters = \
{"simulation_name": "09_long_migration_isolated_mating",
 "population_names": ['pop_01'],
 "population_locations": [[(x,y,z)
                            for x in range(5)
                            for y in range(5)
                            for z in range(1)]],

 "deployment_code": 3,
 "chromosome_bases": ['0','1'],
 "background_mutation": 0.001,
 "additional_mutation": 0,
 "mutation_type": 'point',
 "chromosome_size": 5000,
 "genome_size": 1,
 "max_tape_length": 50,
 "clean_cell": True,
 "interpret_chromosome": False,
 "max_codon": 2000,
 "population_size": 1250,
 "eco_cell_capacity": 50,
 "world_x": 5,
 "world_y": 5,
 "world_z": 1,
 "goal": 0,
 "maximum_generations": 1000,
 "fossilized_ratio": 0.01,
 "fossilized_frequency": 100,

```

```

"print_frequency": 10,
"ragaraja_version": 0,
"ragaraja_instructions": ['000', '001', '010',
                          '011', '100', '101'],
"eco_buried_frequency": 1250,
"database_file": "sim09_long_migration.db",
"database_logging_frequency": 1}

class simulation_functions(dose.dose_functions):

    def organism_movement(self, Populations,
                          pop_name, World): pass

    def organism_location(self, Populations,
                          pop_name, World):
        for location in parameters
            ["population_locations"][0]:
                group = dose.filter_location(location,
                    Populations[pop_name].agents)
                for i in range(int(round((len(group) * \
                    0.1)))):
                    (x,y,z) = dose.simulation_calls.
                        coordinates(location)
                    World.ecosystem[x][y][z]['organisms'] =\
                    World.ecosystem[x][y][z]['organisms'] - 1
                    immigrant = random.choice(
                        Populations[pop_name].agents)
                    while immigrant not in group:
                        immigrant = random.choice(
                            Populations[pop_name].agents)
                    new_location = \
                        random.choice(parameters
                            ["population_locations"][0])
                    while new_location == location:
                        new_location = \
                            random.choice(parameters
                                ["population_locations"][0])
                    immigrant.status['location'] = \
                        new_location
                    (x,y,z) = dose.simulation_calls.
                        coordinates(new_location)
                    World.ecosystem[x][y][z]['organisms'] =\
                    World.ecosystem[x][y][z]['organisms'] + 1

    def ecoregulate(self, World): pass
    def update_ecology(self, World, x, y, z): pass
    def update_local(self, World, x, y, z): pass
    def report(self, World): pass

```

```

def fitness(self, Populations, pop_name): pass

def mutation_scheme(self, organism):
    organism.genome[0].rmutate
    (parameters["mutation_type"],
     parameters["additional_mutation"])

def prepopulation_control(self, Populations,
    pop_name): pass

def mating(self, Populations, pop_name):
    for location in parameters
    ["population_locations"][0]:
        group = dose.filter_location(location,
            Populations[pop_name].agents)
        for x in range(len(group)//2):
            parents = []
            for i in range(2):
                parents.append(random.choice(
                    Populations[pop_name].agents))
                while parents[i] not in group:
                    parents[i] = \
                        random.choice(
                            Populations[pop_name].agents)
            Populations[pop_name].agents.
            remove(parents[i])
            crossover_pt = random.randint(0,
                len(parents[0].genome[0].sequence))
            (new_chromol, new_chromo2) = \
                dose.genetic.crossover(
                    parents[0].genome[0],
                    parents[1].genome[0],
                    crossover_pt)
            children = [dose.genetic.Organism(
                [new_chromol],
                parameters["mutation_type"],
                parameters["additional_mutation"]),
                dose.genetic.Organism(
                    [new_chromo2],
                    parameters["mutation_type"],
                    parameters["additional_mutation"])]
            for child in children:
                child.status['parents'] = \
                    [parents[0].status['identity'],
                     parents[1].status['identity']]
                child.status['location'] = location
                child.generate_name()
                child.status['deme'] = pop_name

```

```

        Populations[pop_name].agents.
            append(child)

def postpopulation_control(self, Populations,
    pop_name): pass
def generation_events(self, Populations,
    pop_name): pass

def population_report(self, Populations, pop_name):
    report_list = []
    for organism in Populations[pop_name].agents:
        identity = str(organism.status['identity'])
        report_list.append(identity)
    return '\n'.join(report_list)

def database_report(self, con, cur, start_time,
    Populations, World,
    generation_count):

    try:
        dose.database_report_populations(con, cur,
            start_time, Populations, generation_count)
    except: pass
    try:
        dose.database_report_world(con, cur,
            start_time, World, generation_count)
    except: pass

def deployment_scheme(self, Populations,
    pop_name, World): pass

dose.simulate(parameters, simulation_functions)

```

EXAMPLE FOR DEMONSTRATING REVIVAL OF SIMULATION FROM DATABASE

*'''Reviving a simulation from simulation logging database
and run the simulation for another 200 generations.*

*Example 09 ran the simulation for 1000 generations and
logged the events into case_study_01.db file. This example
revived the logged simulation from case_study_01.db file
using the simulation start time as identifier (2013-10-19-
1382200534.1) and run the simulation from generation 1001 to
1200. The events of generation 1001 to 1200 are logged in
case_study_01.db
file.'''*

try:

```

    import run_examples_without_installation
except ImportError: pass

import dose, genetic
import os, random

'''
Needs pre-existing 09_no_migration_isolated_mating.py.
Change simulation_time below to the corresponding starting
time of the said simulation.
'''
rev_parameters = \
{ "database_source" : "case_study_01.db",
  "simulation_time": "2013-10-19-1382200534.1",
  "population_locations": [[(x,y,z)
                             for x in xrange(5)
                             for y in xrange(5)
                             for z in xrange(1)]] ,

  "rev_start" : [1000],
  "extend_gen" : 200,
  "simulation_name": "08_revive_simulation_03",
  "chromosome_bases": ['0', '1'],
  "background_mutation": 0.1,
  "additional_mutation": 0,
  "mutation_type": 'point',
  "max_tape_length": 50,
  "clean_cell": True,
  "interpret_chromosome": True,
  "max_codon": 2000,
  "goal": 0,
  "eco_cell_capacity": 100,
  "fossilized_ratio": 0.01,
  "fossilized_frequency": 20,
  "print_frequency": 10,
  "ragaraja_version": 0,
  "ragaraja_instructions": ['000', '001', '010',
                             '011', '100', '101'],
  "eco_buried_frequency": 100,
  "database_file": "case_study_01.db",
  "database_logging_frequency": 1}

class simulation_functions(dose.dose_functions):

    def organism_movement(self, Populations,
                          pop_name, World): pass
    def organism_location(self, Populations,
                          pop_name, World): pass
    def.ecoregulate(self, World): pass

```

```

def update_ecology(self, World, x, y, z): pass
def update_local(self, World, x, y, z): pass
def report(self, World): pass
def fitness(self, Populations, pop_name): pass

def mutation_scheme(self, organism):
    organism.genome[0].rmutate(
        rev_parameters["mutation_type"],
        rev_parameters["additional_mutation"])

def prepopulation_control(self, Populations,
    pop_name): pass

def mating(self, Populations, pop_name):
    for location in rev_parameters
        ["population_locations"][0]:
        group = dose.filter_location(location,
            Populations[pop_name].agents)
        for x in xrange(len(group)/2):
            parents = []
            for i in xrange(2):
                parents.append(random.choice(
                    Populations[pop_name].agents))
                while parents[i] not in group:
                    parents[i] = random.choice(
                        Populations[pop_name].agents)
            Populations[pop_name].agents.
                remove(parents[i])
            crossover_pt = random.randint(0,
                len(parents[0].genome[0].sequence))
            (new_chromol, new_chromo2) = \
                genetic.crossover(
                    parents[0].genome[0],
                    parents[1].genome[0],
                    crossover_pt)
            children = [genetic.Organism(
                [new_chromol],
                rev_parameters["mutation_type"],
                rev_parameters["additional_mutation"]),
                genetic.Organism(
                    [new_chromo2],
                    rev_parameters["mutation_type"],
                    rev_parameters["additional_mutation"])]
            for child in children:
                child.status['location'] = location
                child.generate_name()
                child.status['deme'] = pop_name
                Populations[pop_name].agents.

```

```

        append(child)

    def postpopulation_control(self, Populations,
                              pop_name): pass
    def generation_events(self, Populations,
                          pop_name): pass

    def population_report(self, Populations, pop_name):
        report_list = []
        for organism in Populations[pop_name].agents:
            chromosome = \
                ''.join(organism.genome[0].sequence)
            location = str(organism.status['location'])
            report_list.append(chromosome + ' ' + \
                               location)
        return '\n'.join(report_list)

    def database_report(self, con, cur, start_time,
                       Populations, World,
                       generation_count):
        try:
            dose.database_report_populations(con, cur,
                                             start_time, Populations, generation_count)
        except: pass
        try:
            dose.database_report_world(con, cur,
                                       start_time, World, generation_count)
        except: pass

    def deployment_scheme(self, Populations,
                          pop_name, World): pass

dose.revive_simulation(rev_parameters,
                      simulation_functions)

```

HOW TO WRITE A SIMULATION

A simulation script is a Python code with 4 sections:

1. Import dose library by “import dose”
2. A dictionary of simulation parameters
3. A class containing simulation methods/functions
4. Run simulation by calling DOSE simulation function with the simulation parameters and simulation methods as parameters: `dose.simulate(parameters, simulation_functions)`

Since (1) and (4) are standard, only (2) and (3) are elaborated here.

A dictionary of simulation parameters provides for the set of parameters needed to construct the simulation and store the simulation results.

```
parameters = \
{"simulation_name": \
    "01_basic_functions_one_cell_deployment",
 "population_names": ['pop_01'],
 "population_locations": [[(0,0,0)]],
 "deployment_code": 1,
 "chromosome_bases": ['0','1'],
 "background_mutation": 0.1,
 "additional_mutation": 0,
 "mutation_type": 'point',
 "chromosome_size": 30,
 "genome_size": 1,
 "max_tape_length": 50,
 "clean_cell": True,
 "interpret_chromosome": True,
 "max_codon": 2000,
 "population_size": 100,
 "eco_cell_capacity": 100,
 "world_x": 5,
 "world_y": 5,
 "world_z": 5,
 "goal": 0,
 "maximum_generations": 100,
 "fossilized_ratio": 0.01,
 "fossilized_frequency": 20,
 "print_frequency": 10,
 "ragaraja_version": 0,
 "ragaraja_instructions": ['000', '001', '010',
                           '011', '100', '101'],
 "eco_buried_frequency": 100,
 "database_file": "simulation.db",
 "database_logging_frequency": 1}
```

A class containing simulation methods/functions provides a set of simulation-specific functions.

```
class simulation_functions(dose.dose_functions):

    def organism_movement(self, Populations,
                          pop_name, World): pass
    def organism_location(self, Populations,
                          pop_name, World): pass
    def ecoregulate(self, World): pass
    def update_ecology(self, World, x, y, z): pass
```

```

def update_local(self, World, x, y, z): pass
def report(self, World): pass
def fitness(self, Populations, pop_name): pass

def mutation_scheme(self, organism):
    organism.genome[0].rmutate(
        parameters["mutation_type"],
        parameters["additional_mutation"])

def prepopulation_control(self, Populations,
    pop_name): pass
def mating(self, Populations, pop_name): pass
def postpopulation_control(self, Populations,
    pop_name): pass
def generation_events(self, Populations,
    pop_name): pass

def population_report(self, Populations, pop_name):
    sequences = [''.join(org.genome[0].sequence)
        for org in Populations[pop_name].agents]
    identities = [org.status['identity']
        for org in Populations[pop_name].agents]
    locations = [str(org.status['location'])
        for org in Populations[pop_name].agents]
    demes = [org.status['deme']
        for org in Populations[pop_name].agents]
    return '\n'.join(sequences)

def database_report(self, con, cur, start_time,
    Populations, World,
    generation_count):
    try:
        dose.database_report_populations(con, cur,
            start_time, Populations, generation_count)
    except: pass
    try:
        dose.database_report_world(con, cur,
            start_time, World, generation_count)
    except: pass

def deployment_scheme(self, Populations,
    pop_name, World): pass

```

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