Package ‘SITH’

January 5, 2021

Type Package

Title A Spatial Model of Intra-Tumor Heterogeneity

Version 1.1.0

Date 2021-01-03

Author Phillip B. Nicol

Maintainer Phillip B. Nicol <philnicol740@gmail.com>

Description Implements a three-dimensional stochastic model of cancer growth and mutation similar to the one described in Waclaw et al. (2015) <doi:10.1038/nature14971>. Allows for interactive 3D visualizations of the simulated tumor. Provides a comprehensive summary of the spatial distribution of mutants within the tumor. Contains functions which create synthetic sequencing datasets from the generated tumor.

License GPL (>= 2)

Depends R (>= 3.6.0)

Imports Rcpp (>= 1.0.4), scatterplot3d, stats, graphics, grDevices

Suggests rgl, igraph, knitr, rmarkdown, testthat

LinkingTo Rcpp

RoxygenNote 7.1.1

VignetteBuilder knitr

Encoding UTF-8

URL https://github.com/phillipnicol/SITH

BugReports https://github.com/phillipnicol/SITH/issues

NeedsCompilation yes

Repository CRAN

Date/Publication 2021-01-05 15:10:02 UTC
Description

The SITH (spatial model of intratumor heterogeneity) package implements a lattice based spatial model of tumor growth and mutation. Interactive 3D visualization of the tumor are possible using rgl. Additional functions for visualization and investigating the spatial distribution of mutants are provided. SITH also provides functions to simulate single cell sequencing and bulk sampling data sets from the simulated tumor.

Background

On-lattice models of tumor growth and mutation are computationally efficient and provide a simple setting to study how spatially constrained growth impacts intratumor heterogeneity. While this model has been studied extensively in literature (see Waclaw (2015), Chkhaidze (2019), Opasic (2019)), existing software is either not publicly available or inconvenient to use with R.

The motivation for creating the SITH package was to provide a spatial simulator that is both easy to use and can be used entirely with R. The core function in the package is simulateTumor(), which wraps a C++ implementation of the model into R using the Rcpp package. Once the results of the simulation are saved as an R object, SITH provides several other useful functions for studying this model.

See the package vignette for more information on the model and the algorithm used.

Author(s)

Phillip B. Nicol
bulkSample

References


Examples

```r
#Simulate tumor growth
out <- simulateTumor()

#3d interactive visualization using rgl
visualizeTumor(out)
#or see regions with lots of mutants
visualizeTumor(out, plot.type = "heat")

#get a summary on the spatial dist. of mutants
sp <- spatialDistribution(out)

#simulate single cell sequencing
Scs <- randomSingleCells(tumor = out, ncells = 5, fnr = 0.1)

#simulate bulk sampling
Bulks <- randomBulkSamples(tumor = out, nsamples = 5)
```

---

**bulkSample**

*Simulate bulk sampling*

**Description**

Simulate bulk sequencing data by taking a local sample from the tumor and computing the variant allele frequencies of the various mutations.

**Usage**

```r
bulkSample(tumor, pos, cube.length = 5, threshold = 0.05, coverage = 0)
```
Arguments

tumor A list which is the output of `simulateTumor()`.
pos The center point of the sample.
cube.length The side length of the cube of cells to be sampled.
threshold Only mutations with an allele frequency greater than the threshold will be included in the sample.
coverage If nonzero then deep sequencing with specified coverage is performed.

Details

A local region of the tumor is sampled by constructing a cube with side length `cube.length` around the center point `pos`. Each cell within the cube is sampled, and the reported quantity is variant (or mutation) allele frequency. Lattice sites without cells are assumed to be normal tissue, and thus the reported MAF may be less than 1.0 even if the mutation is present in all cancerous cells.

If `coverage` is non-zero then deep sequencing can be simulated. For a chosen coverage $C$, it is known that the number of times the base is read follows a $\text{Pois}(C)$ distribution (see Illumina’s website). Let $d$ be the true coverage sampled from this distribution. Then the estimated VAF is drawn from a $\text{Bin}(d, p)/d$ distribution.

Note that `cube.length` is required to be an odd integer (in order to have a well-defined center point).

Value

A data frame with 1 row and columns corresponding to the mutations. The entries are the mutation allele frequency.

Author(s)

Phillip B. Nicol

References


Examples

```r
set.seed(416776544, kind = "Mersenne-Twister", normal.kind = "Inversion")
out <- simulateTumor(max_pop = 1000)
df <- bulkSample(tumor = out, pos = c(0,0,0))
```
**plotSlice**  

2D cross section of the simulated tumor.

**Description**

2D cross section of the simulated tumor.

**Usage**

plotSlice(tumor, slice.dim = "x", level = 0, plot.type = "normal")

**Arguments**

- **tumor**: A list which is the output of `simulateTumor()`.
- **slice.dim**: One of "x", "y" or "z", which denotes the dimension which will be fixed to obtain a 2D cross section.
- **level**: Which value will the dimension given in `slice.dim` be fixed at?
- **plot.type**: Which type of plot to draw. "Normal" assigns a random rgb value to each genotype while "heat" colors cells with more mutations red and cells with fewer mutations blue. This is exactly the same as `plot.type` in `visualizeTumor`.

**Value**

None.

**Author(s)**

Phillip B. Nicol

---

**progressionChain**  

Create a linear chain graph to describe the order of mutations

**Description**

A helper function for `simulateTumor()` which returns to the user the edge list for a linear chain.

**Usage**

progressionChain(n)

**Arguments**

- **n**: Number of vertices in the chain
progressionDAG_from_igraph

Value

A matrix with 4 columns and n-1 rows which will be accepted as input to simulateTumor().

Author(s)

Phillip B. Nicol <philnicol740@gmail.com>

Examples

G <- progressionChain(3)

progressionDAG_from_igraph

Define the progression of mutations from an igraph object

Description

A helper function for simulateTumor() which returns to the user the edge list for a DAG which is defined as an igraph object.

Usage

progressionDAG_from_igraph(iG)

Arguments

iG An igraph object for a directed acyclic graph.

Value

A matrix with 4 columns which contains the edges of the graph as well as the rate of crossing each edge and the selective advantage/disadvantage obtained by crossing each edge.

Author(s)

Phillip B. Nicol <philnicol740@gmail.com>
randomBulkSamples

Simulate multi-region bulk sampling

Description
Simulate bulk sequencing data by taking a local sample from the tumor and computing the variant allele frequencies of the various mutations.

Usage
randomBulkSamples(
  tumor, nsamples, cube.length = 5, threshold = 0.05, coverage = 0
)

Arguments
tumor A list which is the output of simulateTumor().
nsamples The number of bulk samples to take.
cube.length The side length of the cube of cells to be sampled.
threshold Only mutations with an allele frequency greater than the threshold will be included in the sample.
coverage If nonzero then deep sequencing with specified coverage is performed.

Details
This is the same as bulkSample(), except multiple samples are taken with random center points.

Value
A data frame with nsamples rows and columns corresponding to the mutations. The entries are the mutation allele frequency.

Author(s)
Phillip B. Nicol

Examples
out <- simulateTumor(max_pop = 1000)
df <- randomBulkSamples(tumor = out, nsamples = 5, cube.length = 5, threshold = 0.05)
randomNeedles

Simulate fine needle aspiration

Description

Simulate a sampling procedure which takes a fine needle through the simulated tumor and reports the mutation allele frequency of the sampled cells.

Usage

randomNeedles(tumor, nsamples, threshold = 0.05, coverage = 0)

Arguments

tumor A list which is the output of simulateTumor().
nsamples The number of samples to take.
threshold Only mutations with an allele frequency greater than the threshold will be included in the sample.
coverage If nonzero then deep sequencing with specified coverage is performed.

Details

This sampling procedure is inspired by Chkhaidze et. al. (2019) and simulates fine needle aspiration. A random one-dimensional cross-section of the tumor is chosen, and the cells within this cross section are sampled, reporting mutation allele frequency.

Author(s)

Phillip B. Nicol

References


Examples

c <- simulateTumor(max_pop = 1000)
df <- randomNeedles(tumor = c, nsamples = 5)
randomSingleCells  

Simulate single cell sequencing data

Description

Simulate single cell sequencing data by random selecting cells from the tumor.

Usage

randomSingleCells(tumor, ncells, fpr = 0, fnr = 0)

Arguments

tumor  A list which is the output of simulateTumor().

ncells  The number of cells to sample.

fpr  The false positive rate

fnr  The false negative rate

Details

The procedure is exactly the same as singleCell() except that it allows multiple cells to be sequenced at once (chosen randomly throughout the entire tumor).

Value

A data frame with sample names on the row and mutation ID on the column. A 1 indicates that the mutation is present in the cell and a 0 indicates the mutation is not present.

Author(s)

Phillip B. Nicol <philnicol740@gmail.com>

Examples

out <- simulateTumor(max_pop = 1000)
df <- randomSingleCells(tumor = out, ncells = 5, fnr = 0.1)
simulateTumor

Spatial simulation of tumor growth

Description

Simulate the spatial growth of a tumor with a multi-type branching process on the three-dimensional integer lattice.

Usage

```r
simulateTumor(
  max_pop = 250000,
  div_rate = 0.25,
  death_rate = 0.18,
  mut_rate = 0.01,
  driver_prob = 0.003,
  selective_adv = 1.05,
  disease_model = NULL,
  verbose = TRUE
)
```

Arguments

- `max_pop`: Number of cells in the tumor.
- `div_rate`: Cell division rate.
- `death_rate`: Cell death rate.
- `mut_rate`: Mutation rate. When a cell divides, both daughter cell acquire \( \text{Pois}(u) \) genetic alterations.
- `driver_prob`: The probability that a genetic alteration is a driver mutation.
- `selective_adv`: The selective advantage conferred to a driver mutation. A cell with \( k \) driver mutations is given birth rate \( b_s^k \).
- `disease_model`: Edge list for a directed acyclic graph describing possible transitions between states. See `progressionChain()` for an example of a valid input matrix.
- `verbose`: Whether or not to print simulation details to the R console.

Details

The model is based upon Waclaw et al. (2015), although the simulation algorithm used is different. A growth of a cancerous tumor is modeled using an exponential birth-death process on the three-dimensional integer lattice. Each cell is given a birth rate \( b \) and a death rate \( d \) such that the time until cell division or cell death is exponentially distributed with parameters \( b \) and \( d \), respectively. A cell can replicate if at least one of the six sites adjacent to it is unoccupied. Each time cell replication occurs, both daughter cells receive \( \text{Pois}(u) \) genetic alterations. Each alteration is a driver mutation with some probability \( du \). A cell with \( k \) driver mutations is given birth rate \( b_s^k \). The simulation begins with a single cell at the origin at time \( t = 0 \).
simulateTumor

The model is simulated using a Gillespie algorithm. See the package vignette for details on how the algorithm is implemented.

Value

A list with components

- **cell_ids** - A data frame containing the information for the simulated cells. (x,y,z) position, allele ID number (note that 0 is the wild-type allele), number of genetic alterations, and Euclidean distance from origin are included.

- **muts** - A data frame consisting of the mutation ID number, the count of the mutation within the population, and the mutation allele frequency (which is the count divided by N).

- **phylo_tree** - A data frame giving all of the information necessary to determine the order of mutations. The parent of a mutation is defined to be the most recent mutation that precedes it. Since the ID 0 corresponds to the initial mutation, 0 does not have any parents and is thus the root of the tree.

- **genotypes** - A data frame containing the information about the mutations that make up each allele. The i-th row of this data frame corresponds to the allele ID i-1. The positive numbers in each row correspond to the IDs of the mutations present in that allele, while a -1 is simply a placeholder and indicates no mutation. The count column gives the number of cells which have the specific allele.

- **color_scheme** - A vector containing an assignment of a color to each allele.

- **drivers** - A vector containing the ID numbers for the driver mutations.

- **time** - The simulated time (in days).

- **params** - The parameters used for the simulation.

Author(s)

Phillip B. Nicol <philnicol740@gmail.com>

References


Examples

```r
out <- simulateTumor(max_pop = 1000)
#Take a look at mutants in order of decreasing MAF
sig_muts <- out$muts[order(out$muts$MAF, decreasing = TRUE),]

#Specify the disease model
out <- simulateTumor(max_pop = 1000, disease_model = progressionChain(3))
```
SingleCell

Simulate single cell sequencing data

Description

Simulate single cell sequencing data by selecting a cell at a specified position.

Usage

singleCell(tumor, pos, noise = 0)

Arguments

tumor
A list which is the output of `simulateTumor()`.
pos
A vector of length 3 giving the (x,y,z) coordinates of the cell to sample.
noise
The false negative rate.

Details

This function selects the cell at pos (error if no cell at specified position exists) and returns the list of mutations present in the cell. Due to technological artifacts, the false negative rate can be quite higher (10-20 percent). To account for this, the noise parameter introduces false negatives into the data set at the specified rate.

Value

A data frame with 1 row and columns corresponding to the mutations present in the cell. A 1 indicates that the mutation is detected while a 0 indicates the mutation is not detected.

Author(s)

Phillip B. Nicol <philnicol740@gmail.com>

References


Examples

```r
set.seed(1126490984)
out <- simulateTumor(max_pop = 1000)
df <- singleCell(tumor = out, pos = c(0,0,0), noise = 0.1)
```
spatialDistribution

Quantify the spatial distribution of mutants

Description

Provides a summary the spatial distribution of mutants within the simulated tumor.

Usage

spatialDistribution(tumor, N = 500, cutoff = 0.01, make.plot = TRUE)

Arguments

tumor A list which is the output of simulateTumor().
N The number of pairs to sample.
cutoff For a plot of clone sizes, all mutations with a MAF below cutoff are ignored.
make.plot Whether or not to make plots.

Details

The genotype of a cell can be interpreted as a binary vector where the $i$-th component is 1 if mutation $i$ is present in the cell and is 0 otherwise. Then a natural comparison of the similarity between two cells is the Jaccard index $J(A, B) = |I(A, B)|/|U(A, B)|$, where $I(A, B)$ is the intersection of $A$ and $B$ and $U(A, B)$ is the union. This function estimates the Jaccard index as a function of Euclidean distance between the cells by randomly sampling $N$ pairs of cells.

Value

A list with the following components

- mean_mutant - A data frame with 2 columns giving the mean number of mutants as a function of Euclidean distance from the lattice origin (Euclid. distance rounded to nearest integer).
- mean_driver - The same as mean_mutant except for driver mutations only. Will be NULL if no drivers are present in the simulated tumor.
- jaccard A data frame with two columns giving mean jaccard index as a function of Euclidean distance between pairs of cells (rounded to nearest integer).

Author(s)

Phillip B. Nicol

Examples

set.seed(1126490984)
out <- simulateTumor(max_pop = 1000, driver_prob = 0.1)
sp <- spatialDistribution(tumor = out, make.plot = FALSE)
visualizeTumor

Interactive visualization of the simulated tumor

Description

Interactive visualization of the simulated tumor using the rgl package (if available).

Usage

visualizeTumor(tumor, plot.type = "normal", background = "black", axes = FALSE)

Arguments

tumor A list which is the output of simulateTumor().
plot.type Which type of plot to draw. "Normal" assigns a random rgb value to each genotype while "heat" colors cells with more mutations red and cells with fewer mutations blue.
background If rgl is installed, this will set the color of the background
axes Will include axes (rgl only).

Details

If rgl is installed, then the plots will be interactive. If rgl is unavailable, static plots will be created with scatterplot3d. Since plotting performance with scatterplot3d is reduced, it is strongly recommended that rgl is installed for optimal use of this function.

Value

None.

Author(s)

Phillip B. Nicol
Index

bulkSample, 3, 7

plotSlice, 5
progressionChain, 5, 10
progressionDAG_from_igraph, 6

randomBulkSamples, 7
randomNeedles, 8
randomSingleCells, 9

simulateTumor, 2, 4–9, 10, 12–14
singleCell, 9, 12
SITH(SITH-package), 2
SITH-package, 2
spatialDistribution, 13

visualizeTumor, 14