Package 'dyngen'

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Type Package

Title A Multi-Modal Simulator for Spearheading Single-Cell Omics Analyses

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Description A novel, multi-modal simulation engine for studying dynamic cellular processes at single-cell resolution. 'dyngen' is more flexible than current single-cell simulation engines. It allows better method development and benchmarking, thereby stimulating development and testing of novel computational methods. Cannoodt et al. (2020) <doi:10.1101/2020.02.06.936971>.

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URL https://github.com/dynverse/dyngen

BugReports https://github.com/dynverse/dyngen/issues

Depends R (>= 3.5.0)

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backbone *Backbone of the simulation model*

Description

A module is a group of genes which, to some extent, shows the same expression behaviour. Several modules are connected together such that one or more genes from one module will regulate the expression of another module. By creating chains of modules, a dynamic behaviour in gene regulation can be created.

Usage

```
backbone(module_info, module_network, expression_patterns)
```

Arguments

`module_info` A tibble containing meta information on the modules themselves.

- module_id (character): the name of the module
- basal (numeric): basal expression level of genes in this module, must be between [0, 1]
- burn (logical): whether or not outgoing edges of this module will be active during the burn in phase
- independence (numeric): the independence factor between regulators of this module, must be between [0, 1]

`module_network` A tibble describing which modules regulate which other modules.

- from (character): the regulating module
- to (character): the target module
- effect (integer): `1L` if the regulating module upregulates the target module, `-1L` if it downregulates
- strength (numeric): the strength of the interaction
- hill (numeric): hill coefficient, larger than 1 for positive cooperativity, between 0 and 1 for negative cooperativity

`expression_patterns`
A tibble describing the expected expression pattern changes when a cell is simulated by dyngen. Each row represents one transition between two cell states.

- from (character): name of a cell state
- to (character): name of a cell state
- module_progression (character): differences in module expression between the two states. Example: `"+4,-1|+9|-4"` means the expression of module 4 will go up at the same time as module 1 goes down; afterwards module 9 expression will go up, and afterwards module 4 expression will go down again.
- start (logical): Whether or not this from cell state is the start of the trajectory
- burn (logical): Whether these cell states are part of the burn in phase. Cells will not get sampled from these cell states.
- time (numeric): The duration of an transition.

Value

A dyngen backbone.

See Also

`list_backbones()` for a list of all backbone methods.

Examples

```
library(tibble)
backbone <- backbone(
  module_info = tribble(
    ~module_id, ~basal, ~burn, ~independence,
```

```
    "M1",       1,      TRUE,  1,
    "M2",       0,      FALSE, 1,
    "M3",       0,      FALSE, 1
  ),
  module_network = tribble(
    ~from, ~to,  ~effect, ~strength, ~hill,
    "M1",  "M2", 1L,      1,         2,
    "M2",  "M3", 1L,      1,         2
  ),
  expression_patterns = tribble(
    ~from, ~to,  ~module_progression, ~start, ~burn, ~time,
    "s0",  "s1", "+M1",               TRUE,   TRUE,  15,
    "s1",  "s2", "+M2,+M3",           FALSE,  FALSE, 30
  )
)


model <-
  initialise_model(backbone = backbone) %>%
  generate_tf_network() %>%
  generate_feature_network() %>%
  generate_kinetics() %>%
  generate_gold_standard() %>%
  generate_cells() %>%
  generate_experiment()

dataset <- wrap_dataset(model)
```

---

bblego *Design your own custom backbone easily*

Description

You can use the bblego functions in order to create custom backbones using various components. Please note that the bblego functions currently only allow you to create tree-like backbones.

Usage

```
bblego(..., .list = NULL)

bblego_linear(
  from,
  to,
  type = sample(c("simple", "doublerep1", "doublerep2"), 1),
  num_modules = sample(4:6, 1),
  burn = FALSE
)
```

```
bblego_branching(
  from,
  to,
  type = "simple",
  num_steps = 3,
  num_modules = 2 + length(to) * (3 + num_steps),
  burn = FALSE
)

bblego_start(
  to,
  type = sample(c("simple", "doublerep1", "doublerep2"), 1),
  num_modules = sample(4:6, 1)
)

bblego_end(
  from,
  type = sample(c("simple", "doublerep1", "doublerep2"), 1),
  num_modules = sample(4:6, 1)
)
```

Arguments

`..., .list`	bblego components, either as separate args or as a list.
`from`	The begin state of this component.
`to`	The end state of this component.
`type`	Some components have alternative module regulatory networks. `bblego_start()`, `bblego_linear()`, `bblego_end()`: • `"simple"`: a sequence of modules in which every module upregulates the next module. • `"doublerep1"`: a sequence of modules in which every module downregulates the next module, and each module has positive basal expression. • `"doublerep2"`: a sequence of modules in which every module upregulates the next module, but downregulates the one after that. • `"flipflop"`: a sequence of modules in which every module upregulates the next module. In addition, the last module upregulates itself and strongly downregulates the first module. `bblego_branching()`: • `"simple"`: a set of `n` modules (with `n = length(to)`) which all downregulate one another and upregulate themselves. This causes a branching to occur in the trajectory.
`num_modules`	The number of modules this component is allowed to use. Various components might require a minimum number of components in order to work properly.
`burn`	Whether or not these components are part of the warm-up simulation.
`num_steps`	The number of branching steps to reduce the odds of double positive cells occurring.

Details

A backbone always needs to start with a single `bblego_start()` state and needs to end with one or more `bblego_end()` states. The order of the mentioned states needs to be such that a state is never specified in the first argument (except for `bblego_start()`) before having been specified as the second argument.

Value

A dyngen backbone.

Examples

```
backbone <- bblego(
  bblego_start("A", type = "simple", num_modules = 2),
  bblego_linear("A", "B", type = "simple", num_modules = 3),
  bblego_branching("B", c("C", "D"), type = "simple", num_steps = 3),
  bblego_end("C", type = "flipflop", num_modules = 4),
  bblego_end("D", type = "doublerep1", num_modules = 7)
)
```

---

dyngen *dyngen: A multi-modal simulator for spearheading single-cell omics analyses*

Description

A toolkit for generating synthetic single cell data.

---

example_model *A (very!) small toy dyngen model*

Description

Used for showcasing examples of functions.

Usage

```
example_model
```

Format

An object of class `list` (inherits from `dyngen::init`) of length 19.

`generate_cells` *Simulate the cells*

Description

`generate_cells()` runs simulations in order to determine the gold standard of the simulations. `simulation_default()` is used to configure parameters pertaining this process.

Usage

```
generate_cells(model)

simulation_default(
  burn_time = NULL,
  total_time = NULL,
  ssa_algorithm = ssa_etl(tau = 30/3600),
  census_interval = 4,
  experiment_params = bind_rows(simulation_type_wild_type(num_simulations = 32),
    simulation_type_knockdown(num_simulations = 0)),
  store_reaction_firings = FALSE,
  store_reaction_propensities = FALSE,
  compute_cellwise_grn = FALSE,
  compute_dimred = TRUE,
  compute_rna_velocity = FALSE,
  kinetics_noise_function = kinetics_noise_simple(mean = 1, sd = 0.005)
)

simulation_type_wild_type(
  num_simulations,
  seed = sample.int(10 * num_simulations, num_simulations)
)

simulation_type_knockdown(
  num_simulations,
  timepoint = runif(num_simulations),
  genes = "*",
  num_genes = sample(1:5, num_simulations, replace = TRUE, prob = 0.25^(1:5)),
  multiplier = runif(num_simulations, 0, 1),
  seed = sample.int(10 * num_simulations, num_simulations)
)
```

Arguments

`model`	A dyngen intermediary model for which the gold standard been generated with `generate_gold_standard()`.
`burn_time`	The burn in time of the system, used to determine an initial state vector. If `NULL`, the burn time will be inferred from the backbone.

`total_time` The total simulation time of the system. If `NULL`, the simulation time will be inferred from the backbone.

`ssa_algorithm` Which SSA algorithm to use for simulating the cells with `GillespieSSA2::ssa()`

`census_interval`
A granularity parameter for the outputted simulation.

`experiment_params`
A tibble generated by rbinding multiple calls of `simulation_type_wild_type()` and `simulation_type_knockdown()`.

`store_reaction_firings`
Whether or not to store the number of reaction firings.

`store_reaction_propensities`
Whether or not to store the propensity values of the reactions.

`compute_cellwise_grn`
Whether or not to compute the cellwise GRN activation values.

`compute_dimred` Whether to perform a dimensionality reduction after simulation.

`compute_rna_velocity`
Whether or not to compute the propensity ratios after simulation.

`kinetics_noise_function`
A function that will generate noise to the kinetics of each simulation. It takes the `feature_info` and `feature_network` as input parameters, modifies them, and returns them as a list. See `kinetics_noise_none()` and `kinetics_noise_simple()`.

`num_simulations`
The number of simulations to run.

`seed` A set of seeds for each of the simulations.

`timepoint` The relative time point of the knockdown

`genes` Which genes to sample from. `"*"` for all genes.

`num_genes` The number of genes to knockdown.

`multiplier` The strength of the knockdown. Use 0 for a full knockout, 0<x<1 for a knockdown, and >1 for an overexpression.

Value

A dyngen model.

Examples

```
library(dplyr)
model <-
  initialise_model(
    backbone = backbone_bifurcating(),
    simulation = simulation_default(
      ssa_algorithm = ssa_etl(tau = .1),
      experiment_params = bind_rows(
        simulation_type_wild_type(num_simulations = 4),
        simulation_type_knockdown(num_simulations = 4)
      )
```

```
    )
  )


model <- model %>%
  generate_tf_network() %>%
  generate_feature_network() %>%
  generate_kinetics() %>%
  generate_gold_standard() %>%
  generate_cells()

plot_simulations(model)
plot_gold_mappings(model)
plot_simulation_expression(model)

model <- model %>%
  generate_experiment()

dataset <- wrap_dataset(model)
```

`generate_dataset` *Generate a dataset*

Description

This function contains the complete pipeline for generating a dataset with **dyngen**. In order to have more control over how the dataset is generated, run each of the steps in this function separately.

Usage

```
generate_dataset(
  model,
  output_dir = NULL,
  make_plots = FALSE,
  store_dimred = model$simulation_params$compute_dimred,
  store_cellwise_grn = model$simulation_params$compute_cellwise_grn,
  store_rna_velocity = model$simulation_params$compute_rna_velocity
)
```

Arguments

`model`	A dyngen initial model created with `initialise_model()`.
`output_dir`	If not `NULL`, then the generated model and dynwrap dataset will be written to files in this directory.
`make_plots`	Whether or not to generate an overview of the dataset.
`store_dimred`	Whether or not to store the dimensionality reduction constructed on the true counts.

store_cellwise_grn
: Whether or not to also store cellwise GRN information.

store_rna_velocity
: WHether or not to store the log propensity ratios.

Value

A list containing a dyngen model (`li$model`) and a dynwrap dataset (`li$dataset`).

Examples

```
out <-
  initialise_model(
    backbone = backbone_bifurcating()
  ) %>%
  generate_dataset()

model <- out$model
dataset <- out$dataset
```

---

generate_experiment *Sample cells from the simulations*

Description

`generate_experiment()` runs samples cells along the different simulations. `experiment_snapshot()` assumes that cells are sampled from a heterogeneous pool of cells. Cells will thus be sampled uniformily from the trajectory. `experiment_synchronised()` assumes that all the cells are synchronised and are sampled at different timepoints.

Usage

```
generate_experiment(model)

list_experiment_samplers()

experiment_snapshot(
  realcount = NULL,
  sample_capture_rate = function(n) rnorm(n, 1, 0.05) %>% pmax(0),
  weight_bw = 0.1
)

experiment_synchronised(
  realcount = NULL,
  sample_capture_rate = function(n) rnorm(n, 1, 0.05) %>% pmax(0),
  num_timepoints = 8,
  pct_between = 0.75
)
```

Arguments

model	A dyngen intermediary model for which the simulations have been run with generate_cells().
realcount	The name of a dataset in realcounts. If NULL, a random dataset will be sampled from realcounts.
sample_capture_rate	A function that samples values for the simulated capture rates of genes.
weight_bw	[snapshot] A bandwidth parameter for determining the distribution of cells along each edge in order to perform weighted sampling.
num_timepoints	[synchronised] The number of time points used in the experiment.
pct_between	[synchronised] The percentage of 'unused' simulation time.

Value

A dyngen model.

Examples

```
names(list_experiment_samplers())

model <-
  initialise_model(
    backbone = backbone_bifurcating(),
    experiment = experiment_synchronised()
  )


model <-
  model %>%
  generate_tf_network() %>%
  generate_feature_network() %>%
  generate_kinetics() %>%
  generate_gold_standard() %>%
  generate_cells() %>%
  generate_experiment()

dataset <- wrap_dataset(model)
```

---

generate_feature_network

Generate a target network

Description

generate_feature_network() generates a network of target genes that are regulated by the previously generated TFs, and also a separate network of housekeeping genes (HKs). feature_network_default() is used to configure parameters pertaining this process.

Usage

```
generate_feature_network(model)

feature_network_default(
  realnet = NULL,
  damping = 0.01,
  target_resampling = Inf,
  max_in_degree = 5
)
```

Arguments

model	A dyngen intermediary model for which the transcription network has been generated with generate_tf_network().
realnet	The name of a gene regulatory network (GRN) in realnets. If NULL, a random network will be sampled from realnets. Alternatively, a custom GRN can be used by passing a weighted sparse matrix.
damping	A damping factor used for the page rank algorithm used to subsample the realnet.
target_resampling	How many targets / HKs to sample from the realnet per iteration.
max_in_degree	The maximum in-degree of a target / HK.

Value

A dyngen model.

Examples

```
model <-
  initialise_model(
    backbone = backbone_bifurcating(),
    feature_network = feature_network_default(damping = 0.1)
  )


model <- model %>%
  generate_tf_network() %>%
  generate_feature_network()

plot_feature_network(model)

model <- model %>%
  generate_kinetics() %>%
  generate_gold_standard() %>%
  generate_cells() %>%
  generate_experiment()
dataset <- wrap_dataset(model)
```

generate_gold_standard *Simulate the gold standard*

Description

generate_gold_standard() runs simulations in order to determine the gold standard of the simulations. gold_standard_default() is used to configure parameters pertaining this process.

Usage

```
generate_gold_standard(model)

gold_standard_default(tau = 30/3600, census_interval = 10/60)
```

Arguments

model	A dyngen intermediary model for which the kinetics of the feature network has been generated with generate_kinetics().
tau	The time step of the ODE algorithm used to generate the gold standard.
census_interval	A granularity parameter of the gold standard time steps. Should be larger than or equal to tau.

Value

A dyngen model.

Examples

```
model <-
  initialise_model(
    backbone = backbone_bifurcating(),
    gold_standard = gold_standard_default(tau = .01, census_interval = 1)
  )


model <- model %>%
  generate_tf_network() %>%
  generate_feature_network() %>%
  generate_kinetics() %>%
  generate_gold_standard()

plot_gold_simulations(model)
plot_gold_mappings(model)
plot_gold_expression(model)

model <- model %>%
```

```
  generate_cells() %>%
  generate_experiment

dataset <- wrap_dataset(model)
```

`generate_kinetics` *Determine the kinetics of the feature network*

Description

`generate_kinetics()` samples the kinetics of genes in the feature network for which the kinetics have not yet been defined. `kinetics_default()` is used to configure parameters pertaining this process.

Usage

```
generate_kinetics(model)

kinetics_default()
```

Arguments

`model`	A dyngen intermediary model for which the feature network has been generated with `generate_feature_network()`.

Details

To write different kinetics settings, you need to write three functions with interface function(feature_info, feature_network, cache_dir, verbose). Described below are the default kinetics samplers.

`sampler_tfs()` mutates the `feature_info` data frame by adding the following columns:

- `transcription_rate`: the rate at which pre-mRNAs are transcribed, in pre-mRNA / hour. Default distribution: U(1, 2).
- `translation_rate`: the rate at which mRNAs are translated into proteins, in protein / mRNA / hour. Default distribution: U(100, 150).
- `mrna_halflife`: the half-life of (pre-)mRNA molecules, in hours. Default distribution: U(2.5, 5).
- `protein_halflife`: the half-life of proteins, in hours. Default distribution: U(5, 10).
- `splicing_rate`: the rate at which pre-mRNAs are spliced into mRNAs, in reactions / hour. Default value: log(2) / (10/60), which corresponds to a half-life of 10 minutes.
- `independence`: the degree to which all regulators need to be bound for transcription to occur (0), or whether transcription can occur if only one of the regulators is bound (1).

`sampler_nontfs()` samples the `transcription_rate`, `translation_rate`, `mrna_halflife` and `protein_halflife` from a supplementary file of Schwannhäusser et al., 2011, doi.org/10.1038/nature10098. `splicing_rate` is by default the same as in `sampler_tfs()`. `independence` is sampled from U(0, 1).

`sampler_interactions()` mutates the `feature_network` data frame by adding the following columns.

- `effect`: the effect of the interaction; upregulating = +1, downregulating = -1. By default, sampled from -1, 1 with probabilities .25, .75.
- `strength`: the strength of the interaction. Default distribution: 10^U(0, 2).
- `hill`: the hill coefficient. Default distribution: N(2, 2) with a minimum of 1 and a maximum of 10.

Value

A dyngen model.

Examples

```
model <-
  initialise_model(
    backbone = backbone_bifurcating(),
    kinetics = kinetics_default()
  )


model <- model %>%
  generate_tf_network() %>%
  generate_feature_network() %>%
  generate_kinetics() %>%
  generate_gold_standard() %>%
  generate_cells() %>%
  generate_experiment()

dataset <- wrap_dataset(model)
```

---

generate_tf_network *Generate a transcription factor network from the backbone*

Description

`generate_tf_network()` generates the transcription factors (TFs) that drive the dynamic process a cell undergoes. `tf_network_default()` is used to configure parameters pertaining this process.

Usage

```
generate_tf_network(model)

tf_network_default(
  min_tfs_per_module = 1L,
  sample_num_regulators = function() 2,
  weighted_sampling = FALSE
)
```

Arguments

model	A dyngen initial model created with initialise_model().
min_tfs_per_module	The number of TFs to generate per module in the backbone.
sample_num_regulators	A function to generate the number of TFs per module each TF will be regulated by.
weighted_sampling	When determining what TFs another TF is regulated by, whether to perform weighted sampling (by rank) or not.

Value

A dyngen model.

Examples

```
model <-
  initialise_model(
    backbone = backbone_bifurcating(),
    tf_network = tf_network_default(min_tfs_per_module = 1L)
  ) %>%
  generate_tf_network()

plot_feature_network(model)


model <- model %>%
  generate_feature_network() %>%
  generate_kinetics() %>%
  generate_gold_standard() %>%
  generate_cells() %>%
  generate_experiment()

dataset <- wrap_dataset(model)
```

`initialise_model` *Initial settings for simulating a dyngen dataset*

Description

Initial settings for simulating a dyngen dataset

Usage

```
initialise_model(
  backbone,
  num_cells = 1000,
  num_tfs = nrow(backbone$module_info),
  num_targets = 100,
  num_hks = 50,
  distance_metric = c("pearson", "spearman", "cosine", "euclidean", "manhattan"),
  tf_network_params = tf_network_default(),
  feature_network_params = feature_network_default(),
  kinetics_params = kinetics_default(),
  gold_standard_params = gold_standard_default(),
  simulation_params = simulation_default(),
  experiment_params = experiment_snapshot(),
  verbose = TRUE,
  download_cache_dir = NULL,
  num_cores = 1,
  id = NULL
)
```

Arguments

`backbone`	The gene module configuration that determines the type of dynamic process being simulated. See `list_backbones()` for a full list of different backbones available in this package.
`num_cells`	The number of cells to sample.
`num_tfs`	The number of transcription factors (TFs) to generate. TFs are the main drivers of the changes that occur in a cell. TFs are regulated only by other TFs.
`num_targets`	The number of target genes to generate. Target genes are regulated by TFs and sometimes by other target genes.
`num_hks`	The number of housekeeping genes (HKs) to generate. HKs are typically highly expressed, and are not regulated by the TFs or targets.
`distance_metric`	The distance metric to be used to calculate the distance between cells. See `dynutils::calculate_distance()` for a list of possible distance metrics.
`tf_network_params`	Settings for generating the TF network with `generate_tf_network()`, see `tf_network_default()`.

feature_network_params
: Settings for generating the feature network with generate_feature_network(), see feature_network_default().

kinetics_params
: Settings for determining the kinetics of the feature network with generate_kinetics(), see kinetics_default().

gold_standard_params
: Settings pertaining simulating the gold standard with generate_gold_standard(), see gold_standard_default().

simulation_params
: Settings pertaining the simulation itself with generate_cells(), see simulation_default().

experiment_params
: Settings related to how the experiment is simulated with generate_experiment(), see experiment_snapshot() or experiment_synchronised().

verbose
: Whether or not to print messages during the simulation.

download_cache_dir
: If not NULL, temporary downloaded files will be cached in this directory.

num_cores
: Parallellisation parameter for various steps in the pipeline.

id
: An identifier for the model.

Value

A dyngen model.

Examples

```
model <- initialise_model(
  backbone = backbone_bifurcating()
)
plot_backbone_modulenet(model)
plot_backbone_statenet(model)


model <-
  model %>%
  generate_tf_network() %>%
  generate_feature_network() %>%
  generate_kinetics() %>%
  generate_gold_standard() %>%
  generate_cells() %>%
  generate_experiment()

dataset <- wrap_dataset(model)
```

kinetics_noise_none *Add small noise to the kinetics of each simulation*

Description

Add small noise to the kinetics of each simulation

Usage

```
kinetics_noise_none()

kinetics_noise_simple(mean = 1, sd = 0.005)
```

Arguments

mean	The mean level of noise (should be 1)
sd	The sd of the noise (should be a relatively small value)

Value

A list of noise generators for the kinetics.

list_backbones *List of all predefined backbone models*

Description

A module is a group of genes which, to some extent, shows the same expression behaviour. Several modules are connected together such that one or more genes from one module will regulate the expression of another module. By creating chains of modules, a dynamic behaviour in gene regulation can be created.

Usage

```
list_backbones()

backbone_bifurcating()

backbone_bifurcating_converging()

backbone_bifurcating_cycle()

backbone_bifurcating_loop()

backbone_branching(
```

```
  num_modifications = rbinom(1, size = 6, 0.25) + 1,
  min_degree = 3,
  max_degree = sample(min_degree:5, 1)
)

backbone_binary_tree(num_modifications = rbinom(1, size = 6, 0.25) + 1)

backbone_consecutive_bifurcating()

backbone_trifurcating()

backbone_converging()

backbone_cycle()

backbone_cycle_simple()

backbone_linear()

backbone_linear_simple()

backbone_disconnected(
  left_backbone = list_backbones() %>% keep(., names(.) != "disconnected") %>%
    sample(1) %>% first(),
  right_backbone = list_backbones() %>% keep(., names(.) != "disconnected") %>%
    sample(1) %>% first(),
  num_common_modules = 10
)
```

Arguments

num_modifications	The number of branch points in the generated backbone.
min_degree	The minimum degree of each node in the backbone.
max_degree	The maximum degree of each node in the backbone.
left_backbone	A backbone (other than a disconnected backbone), see list_backbones().
right_backbone	A backbone (other than a disconnected backbone), see list_backbones().
num_common_modules	The number of modules which are regulated by either backbone.

Value

A list of all the available backbone generators.

See Also

backbone() for more information on the data structures that define the backbone.

Examples

```
names(list_backbones())

bb <- backbone_bifurcating()
bb <- backbone_bifurcating_converging()
bb <- backbone_bifurcating_cycle()
bb <- backbone_bifurcating_loop()
bb <- backbone_binary_tree()
bb <- backbone_branching()
bb <- backbone_consecutive_bifurcating()
bb <- backbone_converging()
bb <- backbone_cycle()
bb <- backbone_cycle_simple()
bb <- backbone_disconnected()
bb <- backbone_linear()
bb <- backbone_linear_simple()
bb <- backbone_trifurcating()

model <- initialise_model(
  backbone = bb
)


out <- generate_dataset(model)
```

---

plot_backbone_modulenet

Visualise the backbone of a model

Description

Visualise the backbone of a model

Usage

```
plot_backbone_modulenet(model)
```

Arguments

model	A dyngen initial model created with initialise_model().

Value

A ggplot2 object.

Examples

```
data("example_model")
plot_backbone_modulenet(example_model)
```

plot_backbone_statenet *Visualise the backbone state network of a model*

Description

Visualise the backbone state network of a model

Usage

```
plot_backbone_statenet(model, detailed = FALSE)
```

Arguments

model	A dyngen initial model created with initialise_model().
detailed	Whether or not to also plot the substates of transitions.

Value

A ggplot2 object.

Examples

```
data("example_model")
plot_backbone_statenet(example_model)
```

plot_feature_network *Visualise the feature network of a model*

Description

Visualise the feature network of a model

Usage

```
plot_feature_network(
  model,
  show_tfs = TRUE,
  show_targets = TRUE,
  show_hks = FALSE
)
```

Arguments

model	A dyngen intermediary model for which the feature network has been generated with generate_feature_network().
show_tfs	Whether or not to show the transcription factors.
show_targets	Whether or not to show the targets.
show_hks	Whether or not to show the housekeeping genes.

Value

A ggplot2 object.

Examples

```
data("example_model")
plot_feature_network(example_model)
```

---

plot_gold_expression *Visualise the expression of the gold standard over simulation time*

Description

Visualise the expression of the gold standard over simulation time

Usage

```
plot_gold_expression(
  model,
  what = c("mol_premrna", "mol_mrna", "mol_protein"),
  label_changing = TRUE
)
```

Arguments

model	A dyngen intermediary model for which the simulations have been run with generate_gold_standard().
what	Which molecule types to visualise.
label_changing	Whether or not to add a label next to changing molecules.

Value

A ggplot2 object.

Examples

```
data("example_model")
plot_gold_expression(example_model, what = "mol_mrna", label_changing = FALSE)
```

plot_gold_mappings *Visualise the mapping of the simulations to the gold standard*

Description

Visualise the mapping of the simulations to the gold standard

Usage

```
plot_gold_mappings(
  model,
  selected_simulations = NULL,
  do_facet = TRUE,
  mapping = aes(comp_1, comp_2)
)
```

Arguments

model	A dyngen intermediary model for which the simulations have been run with generate_cells().
selected_simulations	Which simulation indices to visualise.
do_facet	Whether or not to facet according to simulation index.
mapping	Which components to plot.

Value

A ggplot2 object.

Examples

```
data("example_model")
plot_gold_mappings(example_model)
```

plot_gold_simulations *Visualise the simulations using the dimred*

Description

Visualise the simulations using the dimred

Usage

```
plot_gold_simulations(
  model,
  detailed = FALSE,
  mapping = aes(comp_1, comp_2),
  highlight = 0
)
```

Arguments

model	A dyngen intermediary model for which the simulations have been run with generate_cells().
detailed	Whether or not to colour according to each separate sub-edge in the gold standard.
mapping	Which components to plot.
highlight	Which simulation to highlight. If highlight == 0 then the gold simulation will be highlighted.

Value

A ggplot2 object.

Examples

```
data("example_model")
plot_gold_simulations(example_model)
```

---

plot_simulations *Visualise the simulations using the dimred*

Description

Visualise the simulations using the dimred

Usage

```
plot_simulations(model, mapping = aes(comp_1, comp_2))
```

Arguments

model	A dyngen intermediary model for which the simulations have been run with generate_cells().
mapping	Which components to plot.

Value

A ggplot2 object.

Examples

```
data("example_model")
plot_simulations(example_model)
```

`plot_simulation_expression` *Visualise the expression of the simulations over simulation time*

Description

Visualise the expression of the simulations over simulation time

Usage

```
plot_simulation_expression(
  model,
  simulation_i = 1:4,
  what = c("mol_premrna", "mol_mrna", "mol_protein"),
  facet = c("simulation", "module_group", "module_id", "none"),
  label_nonzero = FALSE
)
```

Arguments

`model`	A dyngen intermediary model for which the simulations have been run with `generate_cells()`.
`simulation_i`	Which simulation to visualise.
`what`	Which molecule types to visualise.
`facet`	What to facet on.
`label_nonzero`	Plot labels for non-zero molecules.

Value

A ggplot2 object.

Examples

```
data("example_model")
plot_simulation_expression(example_model)
```

realcounts *A set of real single cell expression datasets*

Description

Statistics are derived from these datasets in order to simulate single cell experiments.

Usage

```
realcounts
```

Format

An object of class `tbl_df` (inherits from `tbl`, `data.frame`) with 110 rows and 2 columns.

realnets *A set of gold standard gene regulatory networks*

Description

These networks are subsampled in order to generate realistic feature and housekeeping networks.

Usage

```
realnets
```

Format

An object of class `tbl_df` (inherits from `tbl`, `data.frame`) with 32 rows and 2 columns.

rnorm_bounded *A bounded version of rnorm*

Description

A bounded version of rnorm

Usage

```
rnorm_bounded(n, mean = 0, sd = 1, min = -Inf, max = Inf)
```

Arguments

n	number of observations. If length(n) > 1, the length is taken to be the number required.
mean	vector of means.
sd	vector of standard deviations.
min	lower limits of the distribution.
max	upper limits of the distribution.

Value

Generates values with rnorm, bounded by [min, max]

Examples

```
rnorm_bounded(10)
```

---

simtime_from_backbone *Determine simulation time from backbone*

Description

Determine simulation time from backbone

Usage

```
simtime_from_backbone(backbone, burn = FALSE)
```

Arguments

backbone	A valid dyngen backbone object
burn	Whether or not to compute the simtime for only the burn phase

Value

An estimation of the required simulation time

Examples

```
backbone <- backbone_linear()

simtime_from_backbone(backbone)

model <- initialise_model(
  backbone = backbone,
  simulation_params = simulation_default(
    burn_time = simtime_from_backbone(backbone, burn = TRUE),
    total_time = simtime_from_backbone(backbone, burn = FALSE)
  )
)
```

`wrap_dataset` *Wrap a simulation into a dynwrap object*

Description

The output of this object can be used with **dyno**.

Usage

```
wrap_dataset(
  model,
  store_cellwise_grn = !is.null(model$experiment$cellwise_grn),
  store_dimred = !is.null(model$experiment$dimred),
  store_rna_velocity = !is.null(model$experiment$rna_velocity)
)
```

Arguments

`model`	A dyngen output model for which the experiment has been emulated with `generate_experiment()`.
`store_cellwise_grn`	Whether or not to also store cellwise GRN information.
`store_dimred`	Whether or not to store the dimensionality reduction constructed on the true counts.
`store_rna_velocity`	WHether or not to store the log propensity ratios.

Value

A dynwrap object.

Examples

```
data("example_model")
dataset <- wrap_dataset(example_model)

# dynplot::plot_dimred(dataset)
```

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