

# Package ‘CausalQueries’

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**Type** Package

**Title** Make, Update, and Query Binary Causal Models

**Version** 0.1.0

**Description**

Users can declare binary causal models, update beliefs about causal types given data and calculate arbitrary estimands. Model definition makes use of 'dagitty' functionality. Updating is implemented in 'stan'. The approach used in 'CausalQueries' is a generalization of the 'biqq' models described in ``Mixing Methods: A Bayesian Approach'' (Humphreys and Jacobs, 2015, <[DOI:10.1017/S0003055415000453](https://doi.org/10.1017/S0003055415000453)>). The conceptual extension makes use of work on probabilistic causal models described in Pearl's Causality (Pearl, 2009, <[DOI:10.1017/CBO9780511803161](https://doi.org/10.1017/CBO9780511803161)>).

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CausalQueries-package 'CausalQueries'

**Description**

'CausalQueries' is a package that lets you generate binary causal models, update over models given data and calculate arbitrary causal queries. Model definition makes use of dagitty syntax. Updating is implemented in 'stan'.

all\_data\_types      *All data types***Description**

Creates dataframe with all data types (including NA types) that are possible from a model.

**Usage**

```
all_data_types(
  model,
  complete_data = FALSE,
  possible_data = FALSE,
  given = NULL
)
```

## Arguments

model	A causal_model. A model object generated by <a href="#">make_model</a> .
complete_data	Logical. If ‘TRUE‘ returns only complete data types (no NAs). Defaults to ‘FALSE‘.
possible_data	Logical. If ‘TRUE‘ returns only complete data types (no NAs) that are *possible* given model restrictions. Note that in principle an intervention could make observationally impossible data types arise. Defaults to ‘FALSE‘.
given	A character. A quoted statement that evaluates to logical. Data conditional on specific values.

## Value

A `data.frame` with all data types (including NA types) that are possible from a model.

## Examples

```
all_data_types(make_model('X -> Y'))
model <- make_model('X -> Y') %>% set_restrictions(labels = list(Y = '00'), keep = TRUE)
all_data_types(model)
all_data_types(model, complete_data = TRUE)
all_data_types(model, possible_data = TRUE)
all_data_types(model, given = 'X==1')
all_data_types(model, given = 'X==1 & Y==1')
```

collapse\_data      *Make compact data with data strategies*

## Description

Take a ‘data.frame‘ and return compact ‘data.frame‘ of event types and strategies.

## Usage

```
collapse_data(
  data,
  model,
  drop_NA = TRUE,
  drop_family = FALSE,
  summary = FALSE
)
```

## Arguments

data	A <code>data.frame</code> . Data of nodes that can take three values: 0, 1, and NA. In long form as generated by <code>make_events</code>
model	A <code>causal_model</code> . A model object generated by <code>make_model</code> .
drop_NA	Logical. Whether to exclude strategy families that contain no observed data. Exceptionally if no data is provided, minimal data on data on first node is returned. Defaults to ‘TRUE’
drop_family	Logical. Whether to remove column <code>strategy</code> from the output. Defaults to ‘FALSE’.
summary	Logical. Whether to return summary of the data. See details. Defaults to ‘FALSE’.

## Value

A vector of data events

If `summary = TRUE` ‘collapse\_data’ returns a list containing the following components:

<code>data_events</code>	A compact <code>data.frame</code> of event types and strategies.
<code>observed_events</code>	A vector of character strings specifying the events observed in the data
<code>unobserved_events</code>	A vector of character strings specifying the events not observed in the data

## Examples

```
model <- make_model('X -> Y')

df <- data.frame(X = c(0,1,NA), Y = c(0,0,1))

df %>% collapse_data(model)

collapse_data(df, model, drop_NA = FALSE)

collapse_data(df, model, drop_family = TRUE)

collapse_data(df, model, summary = TRUE)

data <- make_data(model, n = 0)
collapse_data(data, model)

model <- make_model('X -> Y') %>% set_restrictions('X[]==1')
df <- simulate_data(model, n = 10)
df[1,1] <- ''
collapse_data(df, model)
data <- data.frame(X= 0:1)
collapse_data(data, model)
```

---

complements

---

*Make statement for complements*

---

## Description

Generate a statement for X1, X1 complement each other in the production of Y

## Usage

```
complements(X1, X2, Y)
```

## Arguments

- |    |   |
|----|---|
| X1 | A character. The quoted name of the input node 1. |
| X2 | A character. The quoted name of the input node 2. |
| Y  | A character. The quoted name of the outcome node. |

## Value

A character statement of class statement

## See Also

Other statements: [decreasing\(\)](#), [increasing\(\)](#), [interacts\(\)](#), [non\\_decreasing\(\)](#), [non\\_increasing\(\)](#), [substitutes\(\)](#), [te\(\)](#)

## Examples

```
complements('A', 'B', 'W')
```

---

<code>data_type_names</code>	<i>Data type names</i>
------------------------------	------------------------

---

**Description**

Provides names to data types

**Usage**

```
data_type_names(model, data)
```

**Arguments**

<code>model</code>	A <code>causal_model</code> . A model object generated by <a href="#">make_model</a> .
<code>data</code>	A <code>data.frame</code> . Data of nodes that can take three values: 0, 1, and NA. In long form as generated by <a href="#">make_events</a>

**Value**

A vector of strings of data types

**Examples**

```
model <- make_model('X -> Y')
data <- simulate_data(model, n = 2)
data_type_names(model, data)
```

---

<code>decreasing</code>	<i>Make monotonicity statement (negative)</i>
-------------------------	---

---

**Description**

Generate a statement for Y monotonic (decreasing) in X

**Usage**

```
decreasing(X, Y)
```

**Arguments**

<code>X</code>	A character. The quoted name of the input node
<code>Y</code>	A character. The quoted name of the outcome node

**Value**

A character statement of class `statement`

## See Also

Other statements: `complements()`, `increasing()`, `interacts()`, `non_decreasing()`, `non_increasing()`, `substitutes()`, `te()`

## Examples

```
decreasing('A', 'B')
```

`democracy_data`

*Democracy Data*

## Description

A dataset containing information on inequality, democracy, mobilization, and international pressure.  
Made by `devtools::use_data(democracy_data, CausalQueries)`

## Usage

```
democracy_data
```

## Format

A data frame with 84 rows and 5 nodes:

**C**ase Case

**D**emocracy Democracy

**I**nequality Inequality

**P**International Pressure International Pressure

**M**obilization Mobilization

## Source

<https://www.cambridge.org/core/journals/american-political-science-review/article/inequality-and-regime-change-democratic-transitions-and-the-stability-of-democratic-rule/C39AAF4CF274445555FF41F7CC896AE3#fndtn-supplementary-materials/>

---

draw\_causal\_type      *Draw a single causal type given a parameter vector*

---

## Description

Output is a parameter dataframe recording both parameters (case level priors) and the case level causal type.

## Usage

```
draw_causal_type(model, ...)
```

## Arguments

model	A causal_model. A model object generated by <a href="#">make_model</a> .
...	Arguments passed to ‘set_parameters’

## Examples

```
# Simple draw using model's parameter vector
make_model("X -> M -> Y") %>%
  draw_causal_type()

# Draw parameters from priors and draw type from parameters
make_model("X -> M -> Y") %>%
  draw_causal_type(., param_type = "prior_draw")

# Draw type given specified parameters
make_model("X -> M -> Y") %>%
  draw_causal_type(., parameters = 1:10)

# Define a causal type and reveal data
model <- make_model("X -> Y; X <-> Y")
type <- model %>% draw_causal_type()
make_data(model, parameters = type$causal_type)
```

---

expand\_data      *Expand compact data object to data frame*

---

## Description

Expand compact data object to data frame

## Usage

```
expand_data(data_events = NULL, model)
```

**Arguments**

- `data_events` A `data.frame`. It must be compatible with nodes in `model`. The default columns are `event`, `strategy` and `count`.
- `model` A `causal_model`. A model object generated by [make\\_model](#).

**Value**

A `data.frame` with rows as data observation

**Examples**

```
model <- make_model('X->M->Y')
make_events(model, n = 5) %>%
  expand_data(model)
make_events(model, n = 0) %>%
  expand_data(model)
```

<code>expand_wildcard</code>	<i>Expand wildcard</i>
------------------------------	------------------------

**Description**

Expand statement containing wildcard

**Usage**

```
expand_wildcard(to_expand, join_by = "|", verbose = TRUE)
```

**Arguments**

- `to_expand` A character vector of length 1L.
- `join_by` A logical operator. Used to connect causal statements: *AND* ('&') or *OR* ('|'). Defaults to '|'.
- `verbose` Logical. Whether to print expanded query on the console.

**Value**

A character string with the expanded expression. Wildcard '.' is replaced by 0 and 1.

## Examples

```
# Position of parentheses matters for type of expansion
# In the "global expansion" versions of the entire statement are joined
expand_wildcard('Y[X=1, M=.] > Y[X=1, M=.]')
# In the "local expansion" versions of indicated parts are joined
expand_wildcard('Y[X=1, M=.] > (Y[X=1, M=.]')

# If parentheses are missing global expansion used.
expand_wildcard('Y[X=1, M=.] > Y[X=1, M=.]')

# Expressions not requiring expansion are allowed
expand_wildcard('Y[X=1]')
```

## get\_ambiguities\_matrix

*Get ambiguities matrix*

## Description

Return ambiguities matrix if it exists; otherwise calculate it assuming no confounding. The ambiguities matrix maps from causal types into data types.

## Usage

```
get_ambiguities_matrix(model)
```

## Arguments

model	A causal_model. A model object generated by <a href="#">make_model</a> .
-------	--

## Value

A `data.frame`. Causal types (rows) corresponding to possible data realizations (columns).

## Examples

```
model <- make_model('X -> Y')
get_ambiguities_matrix(model = model)
```

`get_causal_types`      *Get causal types*

### Description

Return data frame with types produced from all combinations of possible data produced by a DAG.

### Usage

```
get_causal_types(model)
```

### Arguments

`model`      A `causal_model`. A model object generated by [make\\_model](#).

### Value

A `data.frame` indicating causal types of a model

### Examples

```
get_causal_types(make_model('X -> Y'))
```

`get_event_prob`      *Draw event probabilities*

### Description

Draw event probabilities

### Usage

```
get_event_prob(model, parameters = NULL, A = NULL, P = NULL, given = NULL)
```

### Arguments

`model`      A `causal_model`. A model object generated by [make\\_model](#).

`parameters`      A vector of real numbers in [0,1]. Values of parameters to specify (optional). By default, parameters is drawn from `model$parameters_df`.

`A`      A `data.frame`. Ambiguity matrix. Not required but may be provided to avoid repeated computation for simulations.

`P`      A `data.frame`. Parameter matrix. Not required but may be provided to avoid repeated computation for simulations.

`given`      A string specifying known values on nodes, e.g. "X==1 & Y==1"

**Value**

An array of event probabilities

**Examples**

```
model <- make_model('X -> Y')
get_event_prob(model = model)
get_event_prob(model = model, parameters = rep(1, 6))
get_event_prob(model = model, parameters = 1:6)
```

get\_nodal\_types

*Get list of types for nodes in a DAG*

**Description**

As type labels are hard to interpret for large models, the type list includes an attribute to help interpret them. See `attr(types, interpret)`

**Usage**

```
get_nodal_types(model, collapse = TRUE)
```

**Arguments**

- |          |   |
|----------|---|
| model    | A causal_model. A model object generated by <a href="#">make_model</a> .  |
| collapse | Logical. If ‘TRUE’, shows unique nodal types for each node. If ‘FALSE’, shows for each node a matrix with nodal types as rows and parent types as columns, if applicable. Defaults to ‘TRUE’. |

**Value**

A named list of nodal types for each parent in a DAG

**Examples**

```
model <- make_model('X -> K -> Y')
get_nodal_types(model)

model <- make_model('X -> K -> Y') %>%
  set_restrictions(statement = 'K[X=1]>K[X=0']) %>%
  set_confound(list(K = 'Y[K=1]>Y[K=0'])) 
get_nodal_types(model)
```

`get_parameter_matrix` *Get parameter matrix*

### Description

Return parameter matrix if it exists; otherwise calculate it assuming no confounding. The parameter matrix maps from parameters into causal types. In models without confounding parameters correspond to nodal types.

### Usage

```
get_parameter_matrix(model)
```

### Arguments

<code>model</code>	A model created by <code>make_model()</code>
--------------------	--

### Value

A `data.frame`, the parameter matrix, mapping from parameters to causal types

### Examples

```
model <- make_model('X -> Y')
get_parameter_matrix(model)
```

`get_parameter_names` *Get parameter names*

### Description

Parameter names taken from P matrix or model if no P matrix provided

### Usage

```
get_parameter_names(model, include_paramset = TRUE)
```

### Arguments

<code>model</code>	A <code>causal_model</code> . A model object generated by <code>make_model</code> .
<code>include_paramset</code>	Logical. Whether to include the param set prefix as part of the name.

### Value

A character vector with the names of the parameters in the model

## Examples

```
get_parameter_names(make_model('X->Y'))
```

---

get_param_dist	<i>Get a distribution of model parameters</i>
----------------	---

---

## Description

Using parameters, priors, or posteriors

## Usage

```
get_param_dist(model, using, n_draws = 4000)
```

## Arguments

model	A causal_model. A model object generated by <a href="#">make_model</a> .
using	A character string. It indicates whether to use ‘priors’, ‘posteriors’ or ‘parameters’.
n_draws	An integer. If no prior distribution is provided, generate prior distribution with n_draws number of draws.

## Value

A matrix with the distribution of the parameters in the model

## Examples

```
get_param_dist(model = make_model('X->Y'), using = 'priors', n_draws = 4)
get_param_dist(model = make_model('X->Y'), using = 'parameters')
```

---

get_parents	<i>Get list of parents of all nodes in a model</i>
-------------	--

---

## Description

Get list of parents of all nodes in a model

## Usage

```
get_parents(model)
```

**Arguments**

`model` A `causal_model`. A model object generated by [make\\_model](#).

**Value**

A list of parents in a DAG

**Examples**

```
model <- make_model('X -> K -> Y')
get_parents(model)
```

`get_parmap`

*Get parmap: a matrix mapping from parameters to data types*

**Description**

Gets parmap from a model, or generates if not available.

**Usage**

```
get_parmap(model, A = NULL, P = NULL)
```

**Arguments**

`model` A `causal_model`. A model object generated by [make\\_model](#).

`A` A `data.frame`. Ambiguity matrix. Not required but may be provided to avoid repeated computation for simulations.

`P` A `data.frame`. Parameter matrix. Not required but may be provided to avoid repeated computation for simulations.

**Value**

A matrix

**Examples**

```
get_parmap(model = make_model('X->Y'))
```

---

```
get_prior_distribution
```

*Get a prior distribution from priors*

---

## Description

Add to the model a ‘n\_draws x n\_param‘ matrix of possible parameters.

## Usage

```
get_prior_distribution(model, n_draws = 4000)
```

## Arguments

- |         |  |
|---------|--|
| model   | A causal_model. A model object generated by <a href="#">make_model</a> . |
| n_draws | A scalar. Number of draws.   |

## Value

A ‘data.frame‘ with dimension ‘n\_param‘x ‘n\_draws‘ of possible lambda draws

## See Also

Other prior\_distribution: [make\\_prior\\_distribution\(\)](#), [set\\_prior\\_distribution\(\)](#)

## Examples

```
make_model('X -> Y') %>% set_prior_distribution(n_draws = 5) %>% get_prior_distribution()  
make_model('X -> Y') %>% get_prior_distribution(3)
```

---

```
get_query_types
```

*Look up query types*

---

## Description

Find which nodal or causal types are satisfied by a query.

## Usage

```
get_query_types(model, query, map = "causal_type", join_by = "|")
```

## Arguments

model	A causal_model. A model object generated by <a href="#">make_model</a> .
query	A character string. An expression defining nodal types to interrogate <a href="#">realise_outcomes</a>
map	Types in query. Either nodal_type or causal_type. Default is causal_type.
join_by	A logical operator. Used to connect causal statements: AND ('&') or OR (' '). Defaults to ' '.

## Value

A list containing some of the following elements

types	A named vector with logical values indicating whether a nodal_type or a causal_type satisfy 'query'
query	A character string as specified by the user
expanded_query	A character string with the expanded query. Only differs from 'query' if this contains wildcard '.'.
evaluated_nodes	Value that the nodes take given a query
node	A character string of the node whose nodal types are being queried
type_list	List of causal types satisfied by a query

## Examples

```

model <- make_model('X -> M -> Y; X->Y')
query <- '(Y[X=0] > Y[X=1])'

get_query_types(model, query, map="nodal_type")
get_query_types(model, query, map="causal_type")
get_query_types(model, query)

# Examples with map = "nodal_type"

query <- '(Y[X=0, M = .] > Y[X=1, M = 0])'
get_query_types(model, query, map="nodal_type")

query <- '(Y[] == 1)'
get_query_types(model, query, map="nodal_type")
get_query_types(model, query, map="nodal_type", join_by = '&')

# Root nodes specified with []
get_query_types(model, '(X[] == 1)', map="nodal_type")

query <- '(M[X=1] == M[X=0])'
get_query_types(model, query, map="nodal_type")

# Helpers
model <- make_model('M->Y; X->Y')
query <- complements('X', 'M', 'Y')
get_query_types(model, query, map="nodal_type")

```

```
# Examples with map = "causal_type"

model <- make_model('X -> M -> Y; X->Y')
query <- 'Y[M=M[X=0], X=1]==1'
get_query_types(model, query, map= "causal_type")

query <- '(Y[X=1, M = 1] > Y[X=0, M = 1]) & (Y[X=1, M = 0] > Y[X=0, M = 0])'
get_query_types(model, query, "causal_type")

query <- 'Y[X=1] == Y[X=0]'
get_query_types(model, query, "causal_type")

query <- '(X == 1) & (M==1) & (Y ==1) & (Y[X=0] ==1)'
get_query_types(model, query, "causal_type")

query <- '(Y[X = .]==1)'
get_query_types(model, query, "causal_type")
```

**get\_type\_prob***Get type probabilities***Description**

Gets probability of vector of causal types given a single realization of parameters, possibly drawn from model priors.

**Usage**

```
get_type_prob(model, P = NULL, parameters = NULL)
```

**Arguments**

- |                         |   |
|-------------------------|---|
| <code>model</code>      | A <code>causal_model</code> . A model object generated by <a href="#">make_model</a> .  |
| <code>P</code>          | A <code>data.frame</code> . Parameter matrix. Not required but may be provided to avoid repeated computation for simulations.                           |
| <code>parameters</code> | A vector of real numbers in [0,1]. Values of parameters to specify (optional). By default, parameters is drawn from <code>model\$parameters_df</code> . |

**Details**

By default, parameters is drawn from ‘using‘ argument (either from priors, posteriors, or from `model$parameters`)

**Value**

A vector with probabilities of vector of causal types

## Examples

```
get_type_prob(model = make_model('X->Y'))
get_type_prob(model = make_model('X->Y'), parameters = 1:6)
```

### get\_type\_prob\_multiple

*Draw matrix of type probabilities, before or after estimation*

## Description

Draw matrix of type probabilities, before or after estimation

## Usage

```
get_type_prob_multiple(
  model,
  using = "priors",
  parameters = NULL,
  n_draws = 4000,
  param_dist = NULL,
  P = NULL
)
```

## Arguments

model	A causal_model. A model object generated by <a href="#">make_model</a> .
using	A character. It indicates whether to use 'priors', 'posteriors' or 'parameters'.
parameters	A vector of real numbers in [0,1]. Values of parameters to specify (optional). By default, parameters is drawn from <code>model\$parameters_df</code> .
n_draws	An integer. If no prior distribution is provided, generate prior distribution with <code>n_draws</code> number of draws.
param_dist	A matrix. Distribution of parameters. Optional for speed.
P	A data.frame. Parameter matrix. Not required but may be provided to avoid repeated computation for simulations.

## Value

A matrix of type probabilities.

## Examples

```
model <- make_model('X -> Y')
get_type_prob_multiple(model, using = 'priors', n_draws = 3)
get_type_prob_multiple(model, using = 'parameters', n_draws = 3)
```

---

increasing	<i>Make monotonicity statement (positive)</i>
------------	---

---

### Description

Generate a statement for Y monotonic (increasing) in X

### Usage

```
increasing(X, Y)
```

### Arguments

X	A character. The quoted name of the input node
Y	A character. The quoted name of the outcome node

### Value

A character statement of class `statement`

### See Also

Other statements: [complements\(\)](#), [decreasing\(\)](#), [interacts\(\)](#), [non\\_decreasing\(\)](#), [non\\_increasing\(\)](#), [substitutes\(\)](#), [te\(\)](#)

### Examples

```
increasing('A', 'B')
```

---

---

interacts	<i>Make statement for any interaction</i>
-----------	---

---

### Description

Generate a statement for X1, X2 interact in the production of Y

### Usage

```
interacts(X1, X2, Y)
```

### Arguments

X1	A character. The quoted name of the input node 1.
X2	A character. The quoted name of the input node 2.
Y	A character. The quoted name of the outcome node.

**Value**

A character statement of class `statement`

**See Also**

Other statements: `complements()`, `decreasing()`, `increasing()`, `non_decreasing()`, `non_increasing()`, `substitutes()`, `te()`

**Examples**

```
interacts('A', 'B', 'W')
get_query_types(model = make_model('X-> Y <- W'),
                 query = interacts('X', 'W', 'Y'), map = "causal_type")
```

`interpret_type`

*Interpret or find position in nodal type*

**Description**

Interprets the position of one or more digits (specified by `position`) in a nodal type. Alternatively returns nodal type digit positions that correspond to one or more given condition.

**Usage**

```
interpret_type(model, condition = NULL, position = NULL)
```

**Arguments**

- |                        |  |
|------------------------|--|
| <code>model</code>     | A <code>causal_model</code> . A model object generated by <code>make_model</code> .  |
| <code>condition</code> | A vector of characters. Strings specifying the child node, followed by 'l' (given) and the values of its parent nodes in <code>model</code> .  |
| <code>position</code>  | A named list of integers. The name is the name of the child node in <code>model</code> , and its value a vector of digit positions in that node's nodal type to be interpreted. See 'Details'. |

**Details**

A node for a child node X with k parents has a nodal type represented by X followed by  $2^k$  digits. Argument `position` allows user to interpret the meaning of one or more digit positions in any nodal type. For example `position = list(X = 1:3)` will return the interpretation of the first three digits in causal types for X. Argument `condition` allows users to query the digit position in the nodal type by providing instead the values of the parent nodes of a given child. For example, `condition = 'X | Z=0 & R=1'` returns the digit position that corresponds to values X takes when Z = 0 and R = 1.

**Value**

A named list with interpretation of positions of the digits in a nodal type

**Examples**

```
model <- make_model('R -> X; Z -> X; X -> Y')
#Example using digit position
interpret_type(model, position = list(X = c(3,4), Y = 1))
#Example using condition
interpret_type(model, condition = c('X | Z=0 & R=1', 'X | Z=0 & R=0'))
#Return interpretation of all digit positions of all nodes
interpret_type(model)
```

make\_data

*Make data***Description**

Make data

**Usage**

```
make_data(
  model,
  n = NULL,
  parameters = NULL,
  param_type = NULL,
  nodes = NULL,
  n_steps = NULL,
  probs = NULL,
  subsets = TRUE,
  complete_data = NULL,
  given = NULL,
  verbose = TRUE,
  ...
)
```

**Arguments**

<code>model</code>	A causal_model. A model object generated by <a href="#">make_model</a> .
<code>n</code>	Non negative integer. Number of observations. If not provided it is inferred from the largest n_step.
<code>parameters</code>	A vector of real numbers in [0,1]. Values of parameters to specify (optional). By default, parameters is drawn from model\$parameters_df.

<code>param_type</code>	A character. String specifying type of parameters to make ("flat", "prior_mean", "posterior_mean", "prior_draw", "posterior_draw", "define"). With <code>param_type</code> set to define use arguments to be passed to <code>make_priors</code> ; otherwise flat sets equal probabilities on each nodal type in each parameter set; <code>prior_mean</code> , <code>prior_draw</code> , <code>posterior_mean</code> , <code>posterior_draw</code> take parameters as the means or as draws from the prior or posterior.
<code>nodes</code>	A list. Which nodes to be observed at each step. If NULL all nodes are observed.
<code>n_steps</code>	A list. Number of observations to be observed at each step
<code>probs</code>	A list. Observation probabilities at each step
<code>subsets</code>	A list. Strata within which observations are to be observed at each step. TRUE for all, otherwise an expression that evaluates to a logical condition.
<code>complete_data</code>	A <code>data.frame</code> . Dataset with complete observations. Optional.
<code>given</code>	A string specifying known values on nodes, e.g. "X==1 & Y==1"
<code>verbose</code>	Logical. If TRUE prints step schedule.
<code>...</code>	additional arguments that can be passed to <code>link{make_parameters}</code>

## Details

Note that default behavior is not to take account of whether a node has already been observed when determining whether to select or not. One can however specifically request observation of nodes that have not been previously observed.

## Value

A `data.frame` with simulated data.

## Examples

```
# Simple draws
model <- make_model("X -> M -> Y")
make_data(model)
make_data(model, n = 3, nodes = c("X", "Y"))
make_data(model, n = 3, param_type = "prior_draw")
make_data(model, n = 10, param_type = "define", parameters = 0:9)

# Data Strategies
# A strategy in which X, Y are observed for sure and M is observed
# with 50% probability for X=1, Y=0 cases

model <- make_model("X -> M -> Y")
make_data(
  model,
  n = 8,
  nodes = list(c("X", "Y"), "M"),
  probs = list(1, .5),
  subsets = list(TRUE, "X==1 & Y==0"))
```

```

# n not provided but inferred from largest n_step (not from sum of n_steps)
make_data(
  model,
  nodes = list(c("X", "Y"), "M"),
  n_steps = list(5, 2))

# Wide then deep
make_data(
  model,
  n = 8,
  nodes = list(c("X", "Y"), "M"),
  subsets = list(TRUE, "!is.na(X) & !is.na(Y)"),
  n_steps = list(6, 2))

make_data(
  model,
  n = 8,
  nodes = list(c("X", "Y"), c("X", "M")),
  subsets = list(TRUE, "is.na(X)"),
  n_steps = list(3, 2))

# Example with probabilities at each step

make_data(
  model,
  n = 8,
  nodes = list(c("X", "Y"), c("X", "M")),
  subsets = list(TRUE, "is.na(X)"),
  probs = list(.5, .2))

# Example with given data
make_data(model, given = "X==1 & Y==1", n = 5)

```

**make\_events***Make data in compact form***Description**

Draw n events given event probabilities. Draws full data only. For incomplete data see [make\\_data](#).

**Usage**

```
make_events(
  model,
  n = 1,
  w = NULL,
  P = NULL,
  A = NULL,
```

```

parameters = NULL,
param_type = NULL,
include_strategy = FALSE,
...
)

```

## Arguments

<code>model</code>	A <code>causal_model</code> . A model object generated by <a href="#">make_model</a> .
<code>n</code>	An integer. Number of observations.
<code>w</code>	A numeric matrix. A ‘n_parameters x 1‘ matrix of event probabilities with named rows.
<code>P</code>	A <code>data.frame</code> . Parameter matrix. Not required but may be provided to avoid repeated computation for simulations.
<code>A</code>	A <code>data.frame</code> . Ambiguity matrix. Not required but may be provided to avoid repeated computation for simulations.
<code>parameters</code>	A vector of real numbers in [0,1]. Values of parameters to specify (optional). By default, <code>parameters</code> is drawn from <code>model\$parameters_df</code> .
<code>param_type</code>	A character. String specifying type of parameters to make ('flat', 'prior_mean', 'posterior_mean', 'prior_draw', 'posterior_draw', 'define'). With <code>param_type</code> set to <code>define</code> use arguments to be passed to <code>make_priors</code> ; otherwise <code>flat</code> sets equal probabilities on each nodal type in each parameter set; <code>prior_mean</code> , <code>prior_draw</code> , <code>posterior_mean</code> , <code>posterior_draw</code> take parameters as the means or as draws from the prior or posterior.
<code>include_strategy</code>	Logical. Whether to include a 'strategy' vector. Defaults to FALSE. Strategy vector does not vary with full data but expected by some functions.
<code>...</code>	Arguments to be passed to <code>make_priors</code> if <code>param_type == define</code>

## Value

A `data.frame` of events

## Examples

```

model <- make_model('X -> Y')
make_events(model = model)
make_events(model = model, param_type = 'prior_draw')
make_events(model = model, include_strategy = TRUE)

```

---

make\_model*Make a model*

---

## Description

make\_model uses [dagitty](#) syntax and functionality to specify nodes and edges of a graph. Implied causal types are calculated and default priors are provided under the assumption of no confounding. Models can be updated with specification of a parameter matrix,  $P$ , by providing restrictions on causal types, and/or by providing informative priors on parameters. The default setting for a causal model have flat (uniform) priors and parameters putting equal weight on each parameter within each parameter set. These can be adjust with `set_priors` and `set_parameters`

## Usage

```
make_model(statement, add_causal_types = TRUE)
```

## Arguments

- |                               |   |
|-------------------------------|---|
| <code>statement</code>        | A character. Statement describing causal relations using <a href="#">dagitty</a> syntax. Only directed relations are permitted. For instance "X -> Y" or "X1 -> Y <- X2; X1 -> X2". |
| <code>add_causal_types</code> | Logical. Whether to create and attach causal types to <code>model</code> . Defaults to 'TRUE'.  |

## Value

An object of class `causal_model`.

An object of class "`causal_model`" is a list containing at least the following components:

- |                            |   |
|----------------------------|---|
| <code>dag</code>           | A <code>data.frame</code> with columns 'parent' and 'children' indicating how nodes relate to each other. |
| <code>node</code>          | A named list with the nodes in the model  |
| <code>statement</code>     | A character vector of the statement that defines the model  |
| <code>nodal_types</code>   | A named list with the nodal types in the model  |
| <code>parameters_df</code> | A <code>data.frame</code> with descriptive information of the parameters in the model                     |

## Examples

```
make_model(statement = "X -> Y")
modelXKY <- make_model("X -> K -> Y; X -> Y")

# Example where cyclically dag attempted
## Not run:
modelXKX <- make_model("X -> K -> X")

## End(Not run)
```

```

# Examples with confounding
model <- make_model("X->Y; X <-> Y")
model$P
model <- make_model("Y2 <- X -> Y1; X <-> Y1; X <-> Y2")
dim(model$P)
model$P
model <- make_model("X1 -> Y <- X2; X1 <-> Y; X2 <-> Y")
dim(model$P)
model$parameters_df

# A single node graph is also possible
model <- make_model("X")

# Unconnected nodes cannot
## Not run:
model <- make_model("X <-> Y")

## End(Not run)

```

### **make\_parameter\_matrix *Make parameter matrix***

## Description

Calculate parameter matrix assuming no confounding. The parameter matrix maps from parameters into causal types. In models without confounding parameters correspond to nodal types.

## Usage

```
make_parameter_matrix(model)
```

## Arguments

model	A causal_model. A model object generated by <a href="#">make_model</a> .
-------	--

## Value

A data.frame, the parameter matrix, mapping from parameters to causal types

## Examples

```

model <- make_model('X -> Y')
make_parameter_matrix(model)

```

---

make\_parmap*Make parmap: a matrix mapping from parameters to data types*

---

## Description

Generates a matrix with a row per parameter and a column per data type.

## Usage

```
make_parmap(model, A = NULL, P = NULL)
```

## Arguments

- model            A causal\_model. A model object generated by [make\\_model](#).
- A                A data.frame. Ambiguity matrix. Not required but may be provided to avoid repeated computation for simulations.
- P                A data.frame. Parameter matrix. Not required but may be provided to avoid repeated computation for simulations.

## Value

A matrix

## Examples

```
make_parmap(model = make_model('X->Y'))
make_parmap(model = make_model('X->Y; X<->Y'))
make_parmap(model = make_model('X->Y; X<->Y')) %>% attr("map")
make_parmap(model = make_model('X -> M -> Y; X <-> Y'))
make_parmap(model = make_model('X -> M -> Y; M <-> Y'))
model <- make_model('X -> M -> Y; M <-> Y; X <-> M')
make_parmap(model)
make_parmap(model) %>% attr("map")
# Any ways (without paths splits)
make_parmap(model) %*% (make_parmap(model) %>% attr("map"))

## Not run:
# X1 and X2 are confounded and jointly determine Y1, Y2.
# For instance for models in which X and Y take on four values rather than 2.
model <- make_model("Y2 <- X1 -> Y1; Y2 <- X2 -> Y1; X1 <-> X2; Y1 <-> Y2")
data <- CausalQueries:::minimal_event_data(model)
check <- CausalQueries:::prep_stan_data(model, data, keep_transformed = TRUE)
check$n_params
a <- update_model(model)
make_parmap(model) %>% dim

## End(Not run)
```

**make\_prior\_distribution***Make a prior distribution from priors***Description**

Create a ‘n\_param‘x ‘n\_draws‘ database of possible lambda draws to be attached to the model.

**Usage**

```
make_prior_distribution(model, n_draws = 4000)
```

**Arguments**

- |                      |  |
|----------------------|--|
| <code>model</code>   | A causal_model. A model object generated by <a href="#">make_model</a> . |
| <code>n_draws</code> | A scalar. Number of draws.   |

**Value**

A ‘data.frame‘ with dimension ‘n\_param‘x ‘n\_draws‘ of possible lambda draws

**See Also**

Other prior\_distribution: [get\\_prior\\_distribution\(\)](#), [set\\_prior\\_distribution\(\)](#)

**Examples**

```
make_model('X -> Y') %>% make_prior_distribution(n_draws = 5)
```

**non\_decreasing***Make monotonicity statement (non negative)***Description**

Generate a statement for Y weakly monotonic (increasing) in X

**Usage**

```
non_decreasing(X, Y)
```

**Arguments**

- |                |  |
|----------------|--|
| <code>X</code> | A character. The quoted name of the input node   |
| <code>Y</code> | A character. The quoted name of the outcome node |

**Value**

A character statement of class `statement`

**See Also**

Other statements: `complements()`, `decreasing()`, `increasing()`, `interacts()`, `non_increasing()`, `substitutes()`, `te()`

**Examples**

```
non_decreasing('A', 'B')
```

---

non_increasing	<i>Make monotonicity statement (non positive)</i>
----------------	---

---

**Description**

Generate a statement for Y weakly monotonic (not increasing) in X

**Usage**

```
non_increasing(X, Y)
```

**Arguments**

X	A character. The quoted name of the input node
Y	A character. The quoted name of the outcome node

**Value**

A character statement of class `statement`

**See Also**

Other statements: `complements()`, `decreasing()`, `increasing()`, `interacts()`, `non_decreasing()`, `substitutes()`, `te()`

**Examples**

```
non_increasing('A', 'B')
```

observe_data	<i>Observe data, given a strategy</i>
--------------	---------------------------------------

## Description

Observe data, given a strategy

## Usage

```
observe_data(
  complete_data,
  observed = NULL,
  nodes_to_observe = NULL,
  prob = 1,
  m = NULL,
  subset = TRUE
)
```

## Arguments

complete_data	A <code>data.frame</code> . Data observed and unobserved.
observed	A <code>data.frame</code> . Data observed.
nodes_to_observe	A list. Nodes to observe.
prob	A scalar. Observation probability.
m	A integer. Number of units to observe; if specified, <code>m</code> overrides <code>prob</code> .
subset	A character. Logical statement that can be applied to rows of complete data. For instance observation for some nodes might depend on observed values of other nodes; or observation may only be sought if data not already observed!

## Value

A `data.frame` with logical values indicating which nodes to observe in each row of ‘`complete_data`’.

## Examples

```
model <- make_model("X -> Y")
df <- simulate_data(model, n = 8)
# Observe X values only
observe_data(complete_data = df, nodes_to_observe = "X")
# Observe half the Y values for cases with observed X = 1
observe_data(complete_data = df,
            observed = observe_data(complete_data = df, nodes_to_observe = "X"),
            nodes_to_observe = "Y", prob = .5,
            subset = "X==1")
```

---

parameter_setting	<i>Setting parameters</i>
-------------------	---------------------------

---

## Description

Functionality for altering parameters:

A vector of 'true' parameters; possibly drawn from prior or posterior.

Add a true parameter vector to a model. Parameters can be created using arguments passed to `make_parameters` and `make_priors`.

Extracts parameters as a named vector

## Usage

```
make_parameters(
  model,
  parameters = NULL,
  param_type = NULL,
  warning = TRUE,
  normalize = TRUE,
  ...
)

set_parameters(
  model,
  parameters = NULL,
  param_type = NULL,
  warning = FALSE,
  ...
)

get_parameters(model, param_type = NULL)
```

## Arguments

<code>model</code>	A causal_model. A model object generated by <code>make_model</code> .
<code>parameters</code>	A vector of real numbers in [0,1]. Values of parameters to specify (optional). By default, parameters is drawn from <code>model\$parameters_df</code> .
<code>param_type</code>	A character. String specifying type of parameters to make ("flat", "prior_mean", "posterior_mean", "prior_draw", "posterior_draw", "define). With <code>param_type</code> set to <code>define</code> use arguments to be passed to <code>make_priors</code> ; otherwise <code>flat</code> sets equal probabilities on each nodal type in each parameter set; <code>prior_mean</code> , <code>prior_draw</code> , <code>posterior_mean</code> , <code>posterior_draw</code> take parameters as the means or as draws from the prior or posterior.
<code>warning</code>	Logical. Whether to warn about parameter renormalization.

normalize	Logical. If parameter given for a subset of a family the residual elements are normalized so that parameters in param_set sum to 1 and provided params are unaltered.
...	Options passed onto <a href="#">make_priors</a> .

**Value**

A vector of draws from the prior or distribution of parameters

An object of class causal\_model. It essentially returns a list containing the elements comprising a model (e.g. 'statement', 'nodal\_types' and 'DAG') with true vector of parameters attached to it.

A vector of draws from the prior or distribution of parameters

**Examples**

```
# make_parameters examples:

# Simple examples
model <- make_model('X -> Y')
data  <- simulate_data(model, n = 2)
model <- update_model(model, data)
make_parameters(model, parameters = c(.25, .75, 1.25,.25, .25, .25))
make_parameters(model, param_type = 'flat')
make_parameters(model, param_type = 'prior_draw')
make_parameters(model, param_type = 'prior_mean')
make_parameters(model, param_type = 'posterior_draw')
make_parameters(model, param_type = 'posterior_mean')

#altering values using \code{alter_at}
make_model("X -> Y") %>% make_parameters(parameters = c(0.5,0.25),
alter_at = "node == 'Y' & nodal_type %in% c('00','01')")

#altering values using \code{param_names}
make_model("X -> Y") %>% make_parameters(parameters = c(0.5,0.25),
param_names = c("Y.10","Y.01"))

#altering values using \code{statement}
make_model("X -> Y") %>% make_parameters(parameters = c(0.5),
statement = "Y[X=1] > Y[X=0]")

#altering values using a combination of other arguments
make_model("X -> Y") %>% make_parameters(parameters = c(0.5,0.25),
node = "Y", nodal_type = c("00","01"))

# Normalize renormalizes values not set so that value set is not renormalized
make_parameters(make_model('X -> Y'),
               statement = 'Y[X=1]>Y[X=0]', parameters = .5)
make_parameters(make_model('X -> Y'),
```

```

statement = 'Y[X=1]>Y[X=0]', parameters = .5, normalize = FALSE)

# set_parameters examples:

make_model('X->Y') %>% set_parameters(1:6) %>% get_parameters()

# Simple examples
model <- make_model('X -> Y')
data  <- simulate_data(model, n = 2)
model <- update_model(model, data)
set_parameters(model, parameters = c(.25, .75, 1.25,.25, .25, .25))
set_parameters(model, param_type = 'flat')
set_parameters(model, param_type = 'prior_draw')
set_parameters(model, param_type = 'prior_mean')
set_parameters(model, param_type = 'posterior_draw')
set_parameters(model, param_type = 'posterior_mean')

#altering values using \code{alter_at}
make_model("X -> Y") %>% set_parameters(parameters = c(0.5,0.25),
alter_at = "node == 'Y' & nodal_type %in% c('00','01')")

#altering values using \code{param_names}
make_model("X -> Y") %>% set_parameters(parameters = c(0.5,0.25),
param_names = c("Y.10","Y.01"))

#altering values using \code{statement}
make_model("X -> Y") %>% set_parameters(parameters = c(0.5),
statement = "Y[X=1] > Y[X=0]")

#altering values using a combination of other arguments
make_model("X -> Y") %>% set_parameters(parameters = c(0.5,0.25),
node = "Y", nodal_type = c("00","01"))

# get_parameters examples:

get_parameters(make_model('X -> Y'))

```

**Description**

Functionality for altering priors:

`make_priors` Generates priors for a model.  
`set_priors` Adds priors to a model.  
`Extracts` priors as a named vector

## Usage

```
make_priors(
  model,
  alphas = NA,
  distribution = NA,
  alter_at = NA,
  node = NA,
  nodal_type = NA,
  label = NA,
  param_set = NA,
  given = NA,
  statement = NA,
  join_by = "|",
  param_names = NA
)

set_priors(
  model,
  alphas = NA,
  distribution = NA,
  alter_at = NA,
  node = NA,
  nodal_type = NA,
  label = NA,
  param_set = NA,
  given = NA,
  statement = NA,
  join_by = "|",
  param_names = NA
)

get_priors(model)
```

## Arguments

<code>model</code>	A model object generated by <code>make_model()</code> .
<code>alphas</code>	Real positive numbers giving hyperparameters of the Dirichlet distribution
<code>distribution</code>	string indicating a common prior distribution (uniform, jeffreys or certainty)
<code>alter_at</code>	string specifying filtering operations to be applied to <code>parameters_df</code> , yielding a logical vector indicating parameters for which values should be altered. (see examples)
<code>node</code>	string indicating nodes which are to be altered

<code>nodal_type</code>	string. Label for nodal type indicating nodal types for which values are to be altered
<code>label</code>	string. Label for nodal type indicating nodal types for which values are to be altered. Equivalent to <code>nodal_type</code> .
<code>param_set</code>	string indicating the name of the set of parameters to be altered
<code>given</code>	string indicates the node on which the parameter to be altered depends
<code>statement</code>	causal query that determines nodal types for which values are to be altered
<code>join_by</code>	string specifying the logical operator joining expanded types when <code>statement</code> contains wildcards. Can take values '&' (logical AND) or ' ' (logical OR).
<code>param_names</code>	string. The name of specific parameter in the form of, for example, 'X.1', 'Y.01'

## Details

Seven arguments govern which parameters should be altered. The default is 'all' but this can be reduced by specifying

- \* `alter_at` String specifying filtering operations to be applied to `parameters_df`, yielding a logical vector indicating parameters for which values should be altered. "node == 'X' & nodal\_type
- \* `node`, which restricts for example to parameters associated with node 'X'
- \* `label` or `nodal_type` The label of a particular nodal type, written either in the form Y0000 or Y.Y0000
- \* `param_set` The `param_set` of a parameter.
- \* `given` Given parameter set of a parameter.
- \* `statement`, which restricts for example to nodal types that satisfy the statement 'Y[X=1] > Y[X=0]'
- \* `param_set, given`, which are useful when setting confound statements that produces several sets of parameters

Two arguments govern what values to apply:

- \* `alphas` is one or more non negative numbers and
- \* `distribution` indicates one of a common class: uniform, jeffreys, or 'certain'

Forbidden statements include:

- Setting `distribution` and `values` at the same time.
- Setting a `distribution` other than uniform, jeffreys or certainty.
- Setting negative values.
- specifying `alter_at` with any of `node`, `nodal_type`, `param_set`, `given`, `statement` or `param_names`
- specifying `param_names` with any of `node`, `nodal_type`, `param_set`, `given`, `statement` or `alter_at`
- specifying `statement` with any of `node` or `nodal_type`

## Value

A vector indicating the hyperparameters of the prior distribution of the nodal types.

An object of class `causal_model`. It essentially returns a list containing the elements comprising a model (e.g. 'statement', 'nodal\_types' and 'DAG') with the 'priors' attached to it.

A vector indicating the hyperparameters of the prior distribution of the nodal types.

## Examples

```
# make_priors examples:

# Pass all nodal types
model <- make_model("Y <- X")
make_priors(model, alphas = .4)
make_priors(model, distribution = "jeffreys")

model <- CausalQueries::make_model("X -> M -> Y; X <-> Y")

#altering values using \code{alter_at}
make_priors(model = model, alphas = c(0.5,0.25),
alter_at = "node == 'Y' & nodal_type %in% c('00','01') & given == 'X.0'")

#altering values using \code{param_names}
make_priors(model = model, alphas = c(0.5,0.25),
param_names = c("Y.10_X.0","Y.10_X.1"))

#altering values using \code{statement}
make_priors(model = model, alphas = c(0.5,0.25),
statement = "Y[M=1] > Y[M=0]")

#altering values using a combination of other arguments
make_priors(model = model, alphas = c(0.5,0.25),
node = "Y", nodal_type = c("00","01"), given = "X.0")

# set_priors examples:

# Pass all nodal types
model <- make_model("Y <- X")
set_priors(model, alphas = .4)
set_priors(model, distribution = "jeffreys")

model <- CausalQueries::make_model("X -> M -> Y; X <-> Y")

#altering values using \code{alter_at}
set_priors(model = model, alphas = c(0.5,0.25),
alter_at = "node == 'Y' & nodal_type %in% c('00','01') & given == 'X.0'")

#altering values using \code{param_names}
set_priors(model = model, alphas = c(0.5,0.25),
param_names = c("Y.10_X.0","Y.10_X.1"))

#altering values using \code{statement}
set_priors(model = model, alphas = c(0.5,0.25),
statement = "Y[M=1] > Y[M=0]")

#altering values using a combination of other arguments
set_priors(model = model, alphas = c(0.5,0.25), node = "Y",
nodal_type = c("00","01"), given = "X.0")
```

```
# get_priors examples:  
get_priors(make_model('X -> Y'))
```

query\_distribution     *Calculate query distribution*

## Description

Calculated distribution of a query from a prior or posterior distribution of parameters

## Usage

```
query_distribution(  
  model,  
  query,  
  given = TRUE,  
  using = "parameters",  
  parameters = NULL,  
  type_distribution = NULL,  
  verbose = FALSE,  
  join_by = "|",  
  case_level = FALSE  
)
```

## Arguments

model	A causal_model. A model object generated by <a href="#">make_model</a> .
query	A character. A query on potential outcomes such as "Y[X=1] - Y[X=0]"
given	A character. A quoted expression evaluates to logical statement. given allows the query to be conditioned on *observational* distribution. A value of TRUE is interpreted as no conditioning.
using	A character. Whether to use priors, posteriors or parameters
parameters	A vector of real numbers in [0,1]. A true parameter vector to be used instead of parameters attached to the model in case using specifies parameters
type_distribution	A numeric vector. If provided saves calculation, otherwise calculated from model; may be based on prior or posterior
verbose	Logical. Whether to print mean and standard deviation of the estimand on the console.
join_by	A character. The logical operator joining expanded types when query contains wildcard (.). Can take values "&" (logical AND) or " " (logical OR). When restriction contains wildcard (.) and join_by is not specified, it defaults to " ", otherwise it defaults to NULL.
case_level	Logical. If TRUE estimates the probability of the query for a case.

**Value**

A vector of draws from the distribution of the potential outcomes specified in query

**Examples**

```
model <- make_model("X -> Y") %>%
  set_prior_distribution() %>%
  set_parameter_matrix()

distribution <- query_distribution(model, query = "(Y[X=1] - Y[X=0])")

distribution <- query_distribution(model, query = "(Y[X=1] - Y[X=0])", given = "X==1")
distribution <- query_distribution(model, query = "(Y[X=1] - Y[X=0])", given = "Y[X=1]==0")
distribution <- query_distribution(model, query = "(Y[X=1] - Y[X=0])", given = "Y[X=1]==1")
distribution <- query_distribution(model, query = "(Y[X=1] > Y[X=0])")
distribution <- query_distribution(model,
  query = "(Y[X=1] > Y[X=0])",
  given = "X==1 & Y==1",
  verbose = TRUE)
distribution <- query_distribution(model,
  query = "(Y[X=1] > Y[X=0])",
  given = "X==1 & Y==1",
  case_level = TRUE,
  verbose = TRUE)
distribution <- query_distribution(model, query = "(Y[X=.] == 1)", join_by = "&")
distribution <- query_distribution(model, query = "(Y[X=1] - Y[X=0])", using = "parameters")
df   <- simulate_data(model, n = 3)
updated_model <- update_model(model, df)
query_distribution( updated_model , query = "(Y[X=1] - Y[X=0])", using = "posteriors")
updated_model <- update_model(model, df, keep_transformed = TRUE)
query_distribution( updated_model , query = "(Y[X=1] - Y[X=0])", using = "posteriors")
```

**query\_model**

*Generate estimands dataframe*

**Description**

Calculated from a parameter vector, from a prior or from a posterior distribution

**Usage**

```
query_model(
  model,
  queries = NULL,
  given = NULL,
  using = list("parameters"),
  parameters = NULL,
  stats = NULL,
```

```

n_draws = 4000,
expand_grid = FALSE,
case_level = FALSE,
query = NULL
)

```

### Arguments

model	A causal_model. A model object generated by <a href="#">make_model</a> .
queries	A vector of characters. Query on potential outcomes such as "Y[X=1] - Y[X=0]".
given	A character. A quoted expression that evaluates to a logical statement. Allows estimand to be conditioned on *observational* (or counterfactual) distribution.
using	A character. Whether to use priors, posteriors or parameters.
parameters	A vector of real numbers in [0,1]. Values of parameters to specify (optional). By default, parameters is drawn from model\$parameters_df.
stats	Functions to be applied to estimand distribution. If NULL, defaults to mean, standard deviation, and 95% confidence interval.
n_draws	An integer. Number of draws.
expand_grid	Logical. If TRUE then all combinations of provided lists are examined. If not then each list is cycled through separately. Defaults to FALSE.
case_level	Logical. If TRUE estimates the probability of the query for a case.
query	alias for queries

### Value

A `data.frame` with columns Query, Given and Using defined by corresponding input values. Further columns are generated as specified in stats.

### Examples

```

model <- make_model("X -> Y") %>% set_prior_distribution(n_draws = 10000)

estimands_df <- query_model(
  model,
  query = list(ATE = "Y[X=1] - Y[X=0]", Share_positive = "Y[X=1] > Y[X=0]"),
  using = c("parameters", "priors"),
  expand_grid = TRUE)

estimands_df <- query_model(
  model,
  query = list(ATE = "Y[X=1] - Y[X=0]", Share_positive = "Y[X=1] > Y[X=0]"),
  using = c("parameters", "priors"),
  expand_grid = FALSE)

estimands_df <- query_model(
  model,
  using = list("parameters", "priors"),

```

```

query = list(ATE = "Y[X=1] - Y[X=0]", Is_B = "Y[X=1] > Y[X=0]"),
given = list(TRUE, "Y==0 & X==1"),
expand_grid = TRUE)

# An example: a stat representing uncertainty of token causation
token_var <- function(x) mean(x)*(1-mean(x))
estimands_df <- query_model(
  model,
  using = list("parameters", "priors"),
  query = "Y[X=1] > Y[X=0]",
  stats = c(mean = mean, sd = sd, token_var = token_var))

```

**realise\_outcomes**      *Realise outcomes*

## Description

Realise outcomes for all causal types. Calculated by sequentially calculating endogenous nodes. If a do operator is applied to any node then it takes the given value and all its descendants are generated accordingly.

## Usage

```
realise_outcomes(model, dos = NULL, node = NULL)
```

## Arguments

- |       |   |
|-------|---|
| model | A causal_model. A model object generated by <a href="#">make_model</a> .  |
| dos   | A named list. Do actions defining node values, e.g., list(X = 0, M = 1).  |
| node  | A character. An optional quoted name of the node whose outcome should be revealed. If specified all values of parents need to be specified via dos. |

## Details

`realise_outcomes` starts off by creating types (via [get\\_nodal\\_types](#)). It then takes types of endogenous and reveals their outcome based on the value that their parents took. Exogenous nodes outcomes correspond to their type.

## Value

A data.frame object of revealed data for each node (columns) given causal / nodal type (rows) .

## Examples

```
model <- make_model("X -> Y")
realise_outcomes(model)

model <- make_model("X1->Y;X2->M;M->Y")
realise_outcomes(model, dos = list(X1 = 1, M = 0))

model <- make_model("X->M->Y")
realise_outcomes(model, dos = list(M = 1), node = "Y")
```

## set\_ambiguities\_matrix

*Set ambiguity matrix*

## Description

Add an ambiguities matrix to a model

## Usage

```
set_ambiguities_matrix(model, A = NULL)
```

## Arguments

- model            A causal\_model. A model object generated by [make\\_model](#).
- A                A data.frame. Ambiguity matrix. Not required but may be provided to avoid repeated computation for simulations.

## Value

An object of type causal\_model with the ambiguities matrix attached

## Examples

```
model <- make_model('X -> Y') %>%
  set_ambiguities_matrix()
model$A
```

---

**set\_confound***Set confound*

---

## Description

Adjust parameter matrix to allow confounding.

## Usage

```
set_confound(model, confound = NULL)
```

## Arguments

- model            A `causal_model`. A model object generated by [make\\_model](#).
- confound        A list of statements indicating pairs of nodes whose types are jointly distributed (e.g. `list("A <-> B", "C <-> D")`).

## Details

Confounding between X and Y arises when the nodal types for X and Y are not independently distributed. In the  $X \rightarrow Y$  graph, for instance, there are 2 nodal types for X and 4 for Y. There are thus 8 joint nodal types:

		t^X				
		0	1			Sum
t^Y	00	$\Pr(t^X=0 \& t^Y=00)$	$\Pr(t^X=1 \& t^Y=00)$	$\Pr(t^Y=00)$		
10	.	.	.	.	.	.
01	.	.	.	.	.	.
11	.	.	.	.	.	.
Sum		$\Pr(t^X=0)$	$\Pr(t^X=1)$			1

This table has 8 interior elements and so an unconstrained joint distribution would have 7 degrees of freedom. A no confounding assumption means that  $\Pr(t^X | t^Y) = \Pr(t^X)$ , or  $\Pr(t^X, t^Y) = \Pr(t^X)\Pr(t^Y)$ . In this case there would be 3 degrees of freedom for Y and 1 for X, totaling 4 rather than 7.

`set_confound` lets you relax this assumption by increasing the number of parameters characterizing the joint distribution. Using the fact that  $P(A,B) = P(A)P(B|A)$  new parameters are introduced to capture  $P(B|A=a)$  rather than simply  $P(B)$ . For instance here two parameters (and one degree of freedom) govern the distribution of types X and four parameters (with 3 degrees of freedom) govern the types for Y given the type of X for a total of  $1+3+3 = 7$  degrees of freedom.

## Value

An object of class `causal_model` with updated `parameters_df` and parameter matrix.

## Examples

```

make_model('X -> Y; X <-> Y') %>%
get_parameters()

make_model('X -> M -> Y; X <-> Y') %>%
get_parameters()

model <- make_model('X -> M -> Y; X <-> Y; M <-> Y')
model$parameters_df

# Example where set_confound is implemented after restrictions
make_model("A -> B -> C") %>%
set_restrictions(increasing("A", "B")) %>%
set_confound("B <-> C") %>%
get_parameters()

# Example where two parents are confounded
make_model('A -> B <- C; A <-> C') %>%
set_parameters(node = "C", c(0.05, .95, .95, 0.05)) %>%
make_data(n = 50) %>%
cor()

# Example with two confounds, added sequentially
model <- make_model('A -> B -> C') %>%
set_confound(list("A <-> B", "B <-> C"))
model$statement
# plot(model)

```

`set_parameter_matrix` *Set parameter matrix*

## Description

Add a parameter matrix to a model

## Usage

```
set_parameter_matrix(model, P = NULL)
```

## Arguments

- model            A causal\_model. A model object generated by [make\\_model](#).
- P                A data.frame. Parameter matrix. Not required but may be provided to avoid repeated computation for simulations.

## Value

An object of class causal\_model. It essentially returns a list containing the elements comprising a model (e.g. 'statement', 'nodal\_types' and 'DAG') with the parameter matrix attached to it.

**Examples**

```
model <- make_model('X -> Y')
P <- diag(8)
colnames(P) <- rownames(model$causal_types)
model <- set_parameter_matrix(model, P = P)
```

**set\_parmap***Set parmap: a matrix mapping from parameters to data types***Description**

Generates and adds parmap to a model

**Usage**

```
set_parmap(model)
```

**Arguments**

**model**            A causal\_model. A model object generated by [make\\_model](#).

**Value**

A matrix

**Examples**

```
set_parmap(model = make_model('X->Y'))
```

**set\_prior\_distribution***Add prior distribution draws***Description**

Add ‘n\_param x n\_draws‘ database of possible lambda draws to the model.

**Usage**

```
set_prior_distribution(model, n_draws = 4000)
```

**Arguments**

**model**            A causal\_model. A model object generated by [make\\_model](#).  
**n\_draws**          A scalar. Number of draws.

**Value**

An object of class `causal_model`. It essentially returns a list containing the elements comprising a model (e.g. 'statement', 'nodal\_types' and 'DAG') with the 'prior\_distribution' attached to it.

**See Also**

Other prior\_distribution: `get_prior_distribution()`, `make_prior_distribution()`

**Examples**

```
make_model('X -> Y') %>% set_prior_distribution(n_draws = 5) %>% get_prior_distribution()
```

set_restrictions	<i>Restrict a model</i>
------------------	-------------------------

**Description**

Restrict a model's parameter space. This reduces the number of nodal types and in consequence the number of unit causal types.

**Usage**

```
set_restrictions(
  model,
  statement = NULL,
  join_by = "|",
  labels = NULL,
  given = NULL,
  keep = FALSE,
  update_types = TRUE,
  wildcard = FALSE
)
```

**Arguments**

<code>model</code>	A <code>causal_model</code> . A model object generated by <code>make_model</code> .
<code>statement</code>	A quoted expressions defining the restriction. If values for some parents are not specified, statements should be surrounded by parentheses, for instance <code>(Y[A = 1] &gt; Y[A = 0])</code> will be interpreted for all combinations of other parents of Y set at possible levels they might take.
<code>join_by</code>	A string. The logical operator joining expanded types when <code>statement</code> contains wildcard <code>(.)</code> . Can take values ' <code>&amp;</code> ' (logical AND) or ' <code> </code> ' (logical OR). When restriction contains wildcard <code>(.)</code> and <code>join_by</code> is not specified, it defaults to ' <code> </code> ', otherwise it defaults to <code>NULL</code> . Note that <code>join_by</code> joins within statements, not across statements.

<b>labels</b>	A list of character vectors specifying nodal types to be kept or removed from the model. Use <code>get_nodal_types</code> to see syntax. Note that <code>labels</code> gets overwritten by <code>statement</code> if <code>statement</code> is not NULL.
<b>given</b>	A character vector or list of character vectors specifying nodes on which the parameter set to be restricted depends. When restricting by <code>statement</code> , <code>given</code> must either be NULL or of the same length as <code>statement</code> . When mixing statements that are further restricted by <code>given</code> and ones that are not, statements without <code>given</code> restrictions should have <code>given</code> specified as one of NULL, NA, "" or "".
<b>keep</b>	Logical. If 'FALSE', removes and if 'TRUE' keeps only causal types specified by <code>statement</code> or <code>labels</code> .
<b>update_types</b>	Logical. If 'TRUE' the 'causal_types' matrix gets updated after application of restrictions.
<b>wildcard</b>	Logical. If 'TRUE' allows for use of wildcards in restriction string. Default 'FALSE'.

## Details

Restrictions are made to nodal types, not to unit causal types. Thus for instance in a model  $X \rightarrow M \rightarrow Y$ , one cannot apply a simple restriction so that  $Y$  is nondecreasing in  $X$ , however one can restrict so that  $M$  is nondecreasing in  $X$  and  $Y$  nondecreasing in  $M$ . To have a restriction that  $Y$  be nondecreasing in  $X$  would otherwise require restrictions on causal types, not nodal types, which implies a form of undeclared confounding (i.e. that in cases in which  $M$  is decreasing in  $X$ ,  $Y$  is decreasing in  $M$ ).

Since restrictions are to nodal types, all parents of a node are implicitly fixed. Thus for model `make_model('X -> Y <- W')` the request `set_restrictions(`(Y[X=1] == 0)`)` is interpreted as `set_restrictions(`(Y[X=1, W=0] == 0 | Y[X=1, W=1] == 0)`)`.

Statements with implicitly controlled nodes should be surrounded by parentheses, as in these examples.

Note that prior probabilities are redistributed over remaining types.

## Value

An object of class `model`. The causal types and nodal types in the model are reduced according to the stated restriction.

## See Also

Other restrictions: [restrict\\_by\\_labels\(\)](#), [restrict\\_by\\_query\(\)](#)

## Examples

```
# 1. Restrict parameter space using statements
model <- make_model('X->Y') %>%
  set_restrictions(statement = c('X[] == 0'))

model <- make_model('X->Y') %>%
  set_restrictions(non_increasing('X', 'Y'))
```

```

model <- make_model('X -> Y <- W') %>%
  set_restrictions(c(decreasing('X', 'Y'), substitutes('X', 'W', 'Y')))

model$parameters_df

model <- make_model('X-> Y <- W') %>%
  set_restrictions(statement = decreasing('X', 'Y'))
model$parameters_df

model <- make_model('X->Y') %>%
  set_restrictions(decreasing('X', 'Y'))
model$parameters_df

model <- make_model('X->Y') %>%
  set_restrictions(c(increasing('X', 'Y'), decreasing('X', 'Y')))
model$parameters_df

# Restrict to define a model with monotonicity
model <- make_model('X->Y') %>%
  set_restrictions(statement = c('Y[X=1] < Y[X=0]'))
get_parameter_matrix(model)

# Restrict to a single type in endogenous node
model <- make_model('X->Y') %>%
  set_restrictions(statement = '(Y[X = 1] == 1)', join_by = '&', keep = TRUE)
get_parameter_matrix(model)

# Use of | and &
# Keep node if *for some value of B* Y[A = 1] == 1
model <- make_model('A->Y<-B') %>%
  set_restrictions(statement = '(Y[A = 1] == 1)', join_by = '|', keep = TRUE)
dim(get_parameter_matrix(model))

# Keep node if *for all values of B* Y[A = 1] == 1
model <- make_model('A->Y<-B') %>%
  set_restrictions(statement = '(Y[A = 1] == 1)', join_by = '&', keep = TRUE)
dim(get_parameter_matrix(model))

# Restrict multiple nodes
model <- make_model('X->Y<-M; X -> M' ) %>%
  set_restrictions(statement = c('Y[X = 1] == 1', '(M[X = 1] == 1)'), join_by = '&', keep = TRUE)
get_parameter_matrix(model)

# Restrict using statements and given:
model <- make_model("X -> Y -> Z; X <-> Z") %>%
  set_restrictions(list(decreasing('X','Y'), decreasing('Y','Z')), given = c(NA,'X.0'))
get_parameter_matrix(model)

# Restrictions on levels for endogenous nodes aren't allowed
## Not run:
model <- make_model('X->Y') %>%

```

```

set_restrictions(statement = '(Y == 1)')

## End(Not run)

# 2. Restrict parameter space Using labels:
model <- make_model('X->Y') %>%
  set_restrictions(labels = list(X = '0', Y = '00'))

# Restrictions can be with wildcards
model <- make_model('X->Y') %>%
  set_restrictions(labels = list(Y = '?0'), wildcard = TRUE)
get_parameter_matrix(model)

# Deterministic model
model <- make_model('S -> C -> Y <- R <- X; X -> C -> R') %>%
  set_restrictions(labels = list(C = '1000', R = '0001', Y = '0001'), keep = TRUE)
get_parameter_matrix(model)

# Restrict using labels and given:
model <- make_model("X -> Y -> Z; X <-> Z") %>%
  set_restrictions(labels = list(X = '0', Z = '00'), given = c(NA, 'X.0'))
get_parameter_matrix(model)

```

**simulate\_data***simulate\_data* is an alias for `make_data`**Description**

`simulate_data` is an alias for `make_data`

**Usage**

```
simulate_data(...)
```

**Arguments**

...	arguments for <code>make_model</code>
-----	---------------------------------------

**Value**

A `data.frame` with simulated data.

**Examples**

```
simulate_data(make_model("X->Y"))
```

---

strategy\_statements    *Generate strategy statements given data*

---

### Description

Helper to generate statements of the form "X = 1 & Y = 0" from realized data on one observation

### Usage

```
strategy_statements(data, strategies)
```

### Arguments

data	A data frame with one row
strategies	A list of strategies where each strategy is a set of nodes to be observed

### Value

A string

### Examples

```
data.frame(X = 1, M = 0, Y = NA) %>%
  strategy_statements(list(c("X", "M", "Y"), "X", "Y"))
```

---

---

substitutes    *Make statement for substitutes*

---

### Description

Generate a statement for X1, X1 substitute for each other in the production of Y

### Usage

```
substitutes(X1, X2, Y)
```

### Arguments

X1	A character. The quoted name of the input node 1.
X2	A character. The quoted name of the input node 2.
Y	A character. The quoted name of the outcome node.

### Value

A character statement of class statement

**See Also**

Other statements: [complements\(\)](#), [decreasing\(\)](#), [increasing\(\)](#), [interacts\(\)](#), [non\\_decreasing\(\)](#), [non\\_increasing\(\)](#), [te\(\)](#)

**Examples**

```
get_query_types(model = make_model('A -> B <- C'),
                 query = substitutes('A', 'C', 'B'), map = "causal_type")

query_model(model = make_model('A -> B <- C'),
            queries = substitutes('A', 'C', 'B'),
            using = 'parameters')
```

te

*Make treatment effect statement (positive)***Description**

Generate a statement for  $(Y(1) - Y(0))$ . This statement when applied to a model returns an element in  $(1,0,-1)$  and not a set of cases. This is useful for some purposes such as querying a model, but not for uses that require a list of types, such as `set_restrictions`.

**Usage**

```
te(X, Y)
```

**Arguments**

- |   |  |
|---|--|
| X | A character. The quoted name of the input node   |
| Y | A character. The quoted name of the outcome node |

**Value**

A character statement of class `statement`

**See Also**

Other statements: [complements\(\)](#), [decreasing\(\)](#), [increasing\(\)](#), [interacts\(\)](#), [non\\_decreasing\(\)](#), [non\\_increasing\(\)](#), [substitutes\(\)](#)

## Examples

```
te('A', 'B')

model <- make_model('X->Y') %>% set_restrictions(increasing('X', 'Y'))
query_model(model, list(ate = te('X', 'Y'))), using = 'parameters')

# set_restrictions breaks with te because it requires a listing
# of causal types, not numeric output.

## Not run:
model <- make_model('X->Y') %>% set_restrictions(te('X', 'Y'))

## End(Not run)
```

update\_model

*Fit causal model using 'stan'*

## Description

Takes a model and data and returns a model object with data attached and a posterior model

## Usage

```
update_model(
  model,
  data = NULL,
  data_type = "long",
  keep_fit = FALSE,
  keep_transformed = TRUE,
  censored_types = NULL,
  ...
)
```

## Arguments

<code>model</code>	A <code>causal_model</code> . A model object generated by <a href="#">make_model</a> .
<code>data</code>	A <code>data.frame</code> . Data of nodes that can take three values: 0, 1, and NA. In long form as generated by <a href="#">make_events</a>
<code>data_type</code>	Either 'long' (as made by <a href="#">simulate_data</a> ) or 'compact' (as made by <a href="#">collapse_data</a> ). Compact data must have entries for each member of each strategy family to produce a valid simplex.
<code>keep_fit</code>	Logical. Whether to append the <code>stanfit</code> object to the model. Defaults to 'FALSE'

```

keep_transformed
  Logical. Whether to keep transformed parameters, prob_of_types, P_lambdas,
  w, w_full
censored_types  vector of data types that are selected out of the data, eg c("X0Y0")
...
  Options passed onto stan call.

```

## Value

An object of class `causal_model`. It essentially returns a list containing the elements comprising a model (e.g. 'statement', 'nodal\_types' and 'DAG') with the 'posterior\_distribution' returned by `stan` attached to it.

## Examples

```

model <- make_model('X->Y')
data_long   <- simulate_data(model, n = 4)
data_short  <- collapse_data(data_long, model)

model_1 <- update_model(model, data_long)

model_2 <- update_model(model, data_long, keep_transformed = FALSE)

## Not run:
# Throws error unless compact data indicated:

model_3 <- update_model(model, data_short)

## End(Not run)

model_4 <- update_model(model, data_short, data_type = 'compact')

# It is possible to implement updating without data, in which case the posterior
# is a stan object that reflects the prior
model_5 <- update_model(model)

# Censored data types
make_model("X->Y") %>%
  update_model(data.frame(X=c(1,1), Y=c(1,1)), censored_types = c("X1Y0")) %>%
  query_model(te("X", "Y"), using = "posteriors")

# Censored data: Learning nothing
make_model("X->Y") %>%
  update_model(data.frame(X=c(1,1), Y=c(1,1)), censored_types = c("X1Y0", "X0Y0", "X0Y1")) %>%
  query_model(te("X", "Y"), using = "posteriors")

```

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