# Package 'tame'

February 23, 2023

<b>Title</b> Timing, Anatomical, Therapeutic and Chemical Based Medication Clustering
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complications

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complications

A Simulated Data Set About Pregnancy Complications

### Description

We use this data set in all the examples in the package.

### Usage

complications

### **Format**

An object of class data. frame with 149 rows and 8 columns.

eczema

A Simulated Data Set About Eczema

### Description

A Simulated Data Set About Eczema

### Usage

eczema

#### **Format**

An object of class data. frame with 50644 rows and 7 columns.

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employ

Employ a Clustering to New Data

#### **Description**

Employ a clustering to new data

#### Usage

```
employ(
  object,
  new_data,
  only = NULL,
  additional_data = NULL,
  assignment_method = "nearest_cluster",
  parallel = FALSE,
   ...
)
```

#### **Arguments**

object A medic clustering object for which employment is desired.

new\_data A data frame in which to look for variables with

only <a href="https://data-masking">data-masking</a> Expressions that return a logical value, and are defined in

terms of the variables in object and/or  $additional\_data$  and specifies which

clusterings should be employed to the new data.

additional\_data

A data frame with additional data that may be (left-)joined onto the parameters in object. This is often used in conjuction with only to select specific cluster-

ings based on additional\_data.

assignment\_method

A character naming the employment method. The default assignment method "nearest\_cluster" matches people in new\_data to their nearest cluster in the chosen clusterings from object. As finding exact matches (the next assignment method) is contained within this strategy the "exact\_only" matches are also reported in additional columns in the output.

The assignment method "exact\_only" only matches a person from new\_data to a cluster if they are a perfect match to anyone in object. Thus, people from new\_data are not guaranteed assignment to a cluster.

parallel

A logical or an integer. If FALSE, the default, no parallelization is done.

If TRUE or an integer larger than 2L parallelization is implemented via parLapply from the **parallel** package. When parallel is TRUE the number of clusters is set to detectCores - 1, and when parallel is an integer then the number of clusters is set to parallel. For more details on the parallelization method see parallel::parLapply.

parallel::parLapply

. . Additional arguments affecting the employment procedure.

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#### Value

employ returns a medic object.

#### **Examples**

```
part1 <- complications[1:100,]
part2 <- complications[101:149,]

clust <- medic(part1, id = id, atc = atc, k = 3)

# Nearest cluster matching
employ(clust, part2)

# Only exact matching
employ(clust, part2, assignment_method = "exact_only")</pre>
```

enrich

Enrich Clustering Parameter

#### **Description**

Enrich the parameter information in a clustering with user-defined data.

#### Usage

```
enrich(object, additional_data = NULL, by = NULL)
```

#### **Arguments**

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A data frame with additional data that may be (left-)joined onto the parameters in object.

by

A character vector of variables to join by. This variables is passed to the by term in a dplyr::left\_join() and inherits its behavior:

If NULL, the default, the join will perform a natural join, using all variables in common across the parameters and additional\_data.

To join by different variables on parameters and additional\_data, use a named vector. For example, by =  $c("k" = "cluster\_size")$  will match parameters\$k to additional\_data\$cluster\_size.

To join by multiple variables, use a vector with length > 1. For example, by =  $c("k", "summation_method")$  will match parametersk to additional\_datak and parameters $summation_method$  to

additional\_data\$summation\_method. Use a named vector to match different variables in parameters and additional\_data.

For example, by =  $c("k" = "cluster\_size", "summation\_method" = "sm")$  will match parameters\$k to additional\_data\$cluster\_size and parameters\$summation\_method to additional\_data\$sm.

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### **Details**

The enrich() function is a joining function used for enriching the clustering characteristics with user-defined data. This function is used in all of the investigative functions with a additional\_data statement such as frequencies() and amounts().

#### Value

An object of class medic.

### **Examples**

```
clust <- medic(
   complications,
   id = id,
   atc = atc,
   timing = first_trimester:third_trimester,
   k = 3:5
)

new_parameters <- data.frame(k = 3:5, size = c("small", "small", "large"))
enrich(clust, new_parameters)</pre>
```

is.medic

Test if an object is a medic-object

### Description

Test if an object is a medic-object

### Usage

```
is.medic(object)
```

### Arguments

object

Any object.

#### Value

TRUE is the object inherits from the medic class and has the required elements.

### **Examples**

```
clust <- medic(complications, id = id, atc = atc, k = 3) is.medic(clust)
```

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medic

Medication clustering (based on ATC and timing)

### Description

The medic method uses agglomerative hierarchical clustering with a bespoke distance measure based on medication ATC codes similarities, medication timing and medication amount or dosage.

### Usage

```
medic(
  data,
  k = 5,
  id,
  atc,
  timing,
  base_clustering,
  linkage = "complete",
  summation_method = "sum_of_minima",
  alpha = 1,
  beta = 1,
  gamma = 1,
  p = 1,
  theta = (5:0)/5,
  parallel = FALSE,
  return_distance_matrix = FALSE,
  set_seed = FALSE,
)
```

### Arguments

data	A data frame containing all the variables for the clustering.
k	a vector specifying the number of clusters to identify.
id	<pre><tidy-select> An unquoted expression naming the variable in data describ- ing person id.</tidy-select></pre>
atc	<tidy-select> An unquoted expression naming the variable in data containing ATC codes.</tidy-select>
timing	<tidy-select> An unquoted expression naming the variable or variables in data describing medication timing. Variable names can be used as if they were positions in the data frame, so expressions like x:y can be used to select a range of variables. Moreover, pattern matching selection helpers such as starts_with or num_range may also be used to select timing variables.</tidy-select>

base\_clustering

<tidy-select> An unquoted expression naming the variable in data that gives an initial clustering to start the medic from or NULL.

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linkage The agglomeration method to be used in the clustering. This should be (an un-

ambiguous abbreviation of) one of "ward.D", "ward.D2", "single", "complete", "average" (= UPGMA), "mcquitty" (= WPGMA), "median" (= WPGMC) or "centroid" (= UPGMC). See stats::hclust for more information. For a discussion

of linkage criterion choice see details below.

summation\_method

The summation method used in the distance measure. This should be either

"double\_sum" or "sum\_of\_minima". See *details* below for more information.

alpha A number giving the tuning of the normalization. See *details* below for more

information.

A number giving the power of the individual medication combinations. See

details below for more information.

gamma A number giving the weight of the timing terms. See *details* below for more

information.

The power of the Minkowski distance used in the timing-specific distance. See

details below for more information.

theta A vector of length 6 specifying the tuning of the ATC measure. See *details* 

below for more information.

parallel A logical or an integer. If FALSE, the default, no parallelization is done.

If TRUE or an integer larger than 2L parallelization is implemented via parLapply from the **parallel** package. When parallel is TRUE the number of clusters is set to detectCores - 1, and when parallel is an integer then the number of clusters is set to parallel. For more details on the parallelization method see

parallel::parLapply.

return\_distance\_matrix

A logical.

set\_seed A logical or an integer.

. . . Additional arguments not currently in use.

### Details

The medic method uses agglomerative hierarchical clustering with a bespoke distance measure based on medication ATC codes and timing similarities to assign medication pattern clusters to people.

Two versions of the distance measure are available:

The double sum:

$$d(p_i, p_j) = N_{\alpha}(M_i \times M_j) \sum_{m \in M_i} \sum_{n \in M_j} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta}.$$

and the sum of minima:

$$d(p_i, p_j) = \frac{1}{2} (N_{\alpha}(M_i) \sum_{m \in M_i} \min_{n \in M_j} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_j} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} \min_{m \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} ((1 + D_{\theta}(m, n))(1 + \gamma T_p(t_{im}, t_{jn})) - 1)^{\beta} + N_{\alpha}(M_j) \sum_{n \in M_i} ((1 + D_{\theta}(m, n))(1 +$$

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#### **Normalization:**

$$N_{\alpha}(x) = |x|^{-\alpha}$$

If the normalization tuning, alpha, is 0, then no normalization is preformed and the distance measure becomes highly dependent on the number of distinct medications given. That is, people using more medication will have larger distances to others. If the normalization tuning, alpha, is 1 - the default - then the summation is normalized with the number of terms in the sum, in other words, the average is calculated.

#### **ATC distance:**

The central idea of this method, namely the ATC distance, is given as

$$D_{\theta}(x,y) = \sum_{i=1,\dots,5} 1\{xandy match on level i, but not level i+1\}\theta_i$$

The ATC distance is tuned using the vector theta.

Note that two ATC codes are said to match at level i when they are identical at level i. E.g. the two codes N06AB01 and N06AA01 match on level 1, 2, and 3 as they are both "N" at level 1, "N06" at level 2, and "N06A" at level 3, but at level 4 they differ ("N06AB" and "N06AA" are not the same).

#### **Timing distance:**

The timing distance is a simple Minkowski distance:

$$T(x,y) = (\sum_{t \in T} |x_t - y_t|^p)^{1/p}.$$

When p is 1, the default, the Manhattan distance is used.

#### Value

An object of class *medic* which describes the clusters produced the hierarchical clustering process. The object is a list with components:

data the inputted data frame data with the cluster assignments appended at the end.

**clustering** a data frame with the person id as given by id, the .analysis\_order and the clusters found.

variables a list of the variables used in the clustering.

**parameters** a data frame with all the inputted clustering parameters and the corresponding method names. These method names correspond to the column names for each cluster in the clustering data frame described right above.

**key** a list of keys used internally in the function to keep track of simplified versions of the data.

**distance\_matrix** the distance matrices for each method if return\_distance\_matrix is TRUE otherwise NULL.

call the matched call.

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#### See Also

```
summary.medic for summaries and plots.

employ for employing an existing clustering to new data.

enrich for enriching the meta data in the medic object with additional data.

bind for binding together two comparable lists of clusterings.
```

### **Examples**

```
# A simple clustering based only on ATC
clust <- medic(complications, id = id, atc = atc, k = 3)
# A simple clustering with both ATC and timing
clust <- medic(
   complications,
   id = id,
   atc = atc,
   timing = first_trimester:third_trimester,
   k = 3
)</pre>
```

parameters\_constructor

Internal option constructor

### **Description**

Given the input of the medic this function checks the input and constructs a data frame with the analysis parameters specified by the user.

### Usage

```
parameters_constructor(
  data,
  id,
  k = 5,
  atc,
  timing,
  base_clustering,
  linkage = "complete",
  summation_method = "sum_of_minima",
  alpha = 1,
  beta = 1,
  gamma = 1,
  p = 1,
  theta = (5:0)/5,
```

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)

### **Arguments**

data A data frame containing all the variables for the clustering.

id <tidy-select> An unquoted expression naming the variable in data describ-

ing person id.

k a vector specifying the number of clusters to identify.

atc <tidy-select> An unquoted expression naming the variable in data contain-

ing ATC codes.

timing <tidy-select> An unquoted expression naming the variable or variables in

data describing medication timing. Variable names can be used as if they were positions in the data frame, so expressions like x:y can be used to select a range of variables. Moreover, pattern matching selection helpers such as starts\_with

or num\_range may also be used to select timing variables.

base\_clustering

<tidy-select> An unquoted expression naming the variable in data that gives

an initial clustering to start the medic from or NULL.

linkage The agglomeration method to be used in the clustering. This should be (an un-

ambiguous abbreviation of) one of "ward.D", "ward.D2", "single", "complete", "average" (= UPGMA), "mcquitty" (= WPGMA), "median" (= WPGMC) or "centroid" (= UPGMC). See stats::hclust for more information. For a discussion

of linkage criterion choice see details below.

summation\_method

The summation method used in the distance measure. This should be either

"double\_sum" or "sum\_of\_minima". See details below for more information.

alpha A number giving the tuning of the normalization. See *details* below for more

information.

beta A number giving the power of the individual medication combinations. See

details below for more information.

gamma A number giving the weight of the timing terms. See *details* below for more

information.

p The power of the Minkowski distance used in the timing-specific distance. See

details below for more information.

theta A vector of length 6 specifying the tuning of the ATC measure. See *details* 

below for more information.

... Additional arguments not currently in use.

#### Value

A data.frame with the parameters for clustering.

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#### **Examples**

```
parameters_constructor(
   data = complications,
   k = 3,
   id = id,
   atc = atc
)
```

refactor

Refactor Cluster Levels

### Description

Refactor the levels of the chosen clusters.

### Usage

```
refactor(object, ..., inheret_parameters = TRUE)
```

### **Arguments**

object

A medic object.

. . .

<data-masking> Name-value pairs. ... is passed to dplyr::mutate, and
therefor inherits its behavior:

The name gives the name of the new clustering in the output. The value can be:

- A vector of length 1, which will be recycled to the correct length.
- A function of another clustering.

When a recording uses the name of an existing clustering, this new clustering will overwrite the existing one.

inheret\_parameters

A logical. If TRUE a new clustering overwriting an existing clustering inherits the parameters of the old.

#### Value

A medic object with relevant clusterings refactored.

### **Examples**

```
clust <- medic(complications, id = id, atc = atc, k = 3:4)

# Refactor one clustering
refactor(
  clust,
  `cluster_1_k=4` = dplyr::recode(`cluster_1_k=4`, IV = "III")</pre>
```

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```
# Refactor all clusterings
refactor(
  clust,
  dplyr::across(
    dplyr::everything(),
    ~dplyr::recode(., IV = "III")
)
)
```

summary.medic

Summary of medic object

### **Description**

Make cluster characterizing summaries.

### Usage

```
## S3 method for class 'medic'
summary(
   object,
   only = NULL,
   clusters = NULL,
   outputs = c("frequencies", "medications", "amounts", "trajectories", "interactions"),
   additional_data = NULL,
   ...
)

## S3 method for class 'summary.medic'
print(x, ...)

## S3 method for class 'summary.medic'
plot(x, by, facet, ...)
```

#### **Arguments**

object An object for which a summary is desired.

only <data-masking> Expressions that return a logical value, and are defined in

terms of the variables in object and/or additional\_data.

The default NULL selects all clusterings in object.

clusters <tidy-select> An unquoted expression naming the cluster or clusters in object

one wants to see summaries of. Names can be used as if they were positions in the data frame, so expressions like I:IV can be used to select a range of clusters.

The default NULL selects all clusters in the chosen clusterings of object.

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outputs

A character vector naming the desired characteristics to output. The default names all possible output types.

additional\_data

A data frame with additional data that may be (left-)joined onto the parameters in object. This is often used in conjuction with only to select specific clusterings based on additional\_data.

.. Additional arguments passed to the internal summary function.

- cluster\_wise an option in the medications() function.
- m an option in the medications() function. A numeric restricting the number of distinct ATC codes plotted within each cluster. That is, the (at most) m most frequent ATC codes within that cluster is given a color.
- q an option in the medications() function. A numeric between 0 and 1 restricting the minimal ATC codes frequency displayed within each cluster.
- count\_grouper an option in the amounts() function. A function for grouping counts. As a standard it groups counts as 1 medication, 2 medications, and 3+ medications.
- atc\_groups A data.frame specifying the ATC groups to summaries by. The
  data.frame must have two columns: (1) regex giving regular expressions
  specifying the wanted ATC groups and (2) atc\_groups the name of this
  ATC grouping. As a standard the anatomical level (first level) of the ATC
  codes is used.

x A summary.medic object for printing or plotting.

#### Value

A list of clustering characteristics of class summary.medic is returned. It can contain any of the following characteristics:

#### **Frequencies:**

The number of individuals assigned to each cluster and the associated frequency of assignment.

#### **Medications:**

The number of individuals with a specific ATC code within a cluster. Moreover, it calculates the percentage of people with this medication assigned to this cluster and the percent of people within the cluster with this medication.

### **Amounts:**

The number of ATC codes an individual has, and then outputs the number of individuals within a cluster that has that many ATC codes. Moreover, various relevant percentages or calculated. See Value below for more details on these percentages.

#### **Trajectories:**

The number of unique timing trajectories in each cluster, and the average timing trajectories in each cluster.

summary.medic

### **Interactions:**

The number of people with unique timing trajectory and ATC group, as given by atc\_groups, in each cluster.

### Methods (by generic)

- print(summary.medic): Print method for medic-objects
- plot(summary.medic): Plot method for medic-objects

### **Examples**

```
clust <- medic(complications, id = id, atc = atc, k = 3:5)</pre>
```

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