

# Package ‘healthyR.ai’

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**Title** The Machine Learning and AI Modeling Companion to 'healthyR'

**Version** 0.0.3

**Description**

Hospital machine learning and ai data analysis workflow tools, modeling, and automations. This library provides many useful tools to review common administrative hospital data. Some of these include predicting length of stay, and readmits. The aim is to provide a simple and consistent verb framework that takes the guesswork out of everything.

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**Encoding** UTF-8

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**URL** <https://github.com/spsanderson/healthyR.ai>

**BugReports** <https://github.com/spsanderson/healthyR.ai/issues>

**Imports** magrittr, rlang (>= 0.1.2), utils, broom, ggrepel, tibble, dplyr, ggplot2, tidyr, forcats, recipes, parsnip, rsample, purrr, h2o

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**VignetteBuilder** knitr

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**NeedsCompilation** no

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get_juiced_data	<i>Get the Juiced Data</i>
-----------------	----------------------------

---

## Description

This is a simple function that will get the juiced data from a recipe.

## Usage

```
get_juiced_data(.recipe_object)
```

## Arguments

`.recipe_object` The recipe object you want to pass.

## Details

Instead of typing out something like: `recipe_object %>% prep() %>% juice() %>% glimpse()`

## Value

A tibble of the prepped and juiced data from the given recipe

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Data Wrangling: [pca\\_your\\_recipe\(\)](#)

**Examples**

```
suppressPackageStartupMessages(library(timetk))
suppressPackageStartupMessages(library(dplyr))
suppressPackageStartupMessages(library(purrr))
suppressPackageStartupMessages(library(healthyR.data))
suppressPackageStartupMessages(library(rsample))
suppressPackageStartupMessages(library(recipes))

data_tbl <- healthyR_data %>%
  select(visit_end_date_time) %>%
  summarise_by_time(
    .date_var = visit_end_date_time,
    .by       = "month",
    value     = n()
  ) %>%
  set_names("date_col", "value") %>%
  filter_by_time(
    .date_var = date_col,
    .start_date = "2013",
    .end_date = "2020"
  )

splits <- initial_split(data = data_tbl, prop = 0.8)

rec_obj <- recipe(value ~., training(splits))

get_juiced_data(rec_obj)
```

---

hai\_control\_chart      *Create a control chart*

---

**Description**

Create a control chart, aka Shewhart chart: [https://en.wikipedia.org/wiki/Control\\_chart](https://en.wikipedia.org/wiki/Control_chart).

**Usage**

```
hai_control_chart(
  .data,
  .value_col,
  .x_col,
  .center_line = mean,
  .std_dev = 3,
  .plt_title = NULL,
```

```

    .plt_catpion = NULL,
    .plt_font_size = 11,
    .print_plot = TRUE
  )

```

### Arguments

<code>.data</code>	data frame or a path to a csv file that will be read in
<code>.value_col</code>	variable of interest mapped to y-axis (quoted, ie as a string)
<code>.x_col</code>	variable to go on the x-axis, often a time variable. If unspecified row indices will be used (quoted)
<code>.center_line</code>	Function used to calculate central tendency. Defaults to mean
<code>.std_dev</code>	Number of standard deviations above and below the central tendency to call a point influenced by "special cause variation." Defaults to 3
<code>.plt_title</code>	Plot title
<code>.plt_catpion</code>	Plot caption
<code>.plt_font_size</code>	Font size; text elements will be scaled to this
<code>.print_plot</code>	Print the plot? Default = TRUE. Set to FALSE if you want to assign the plot to a variable for further modification, as in the last example.

### Details

Control charts, also known as Shewhart charts (after Walter A. Shewhart) or process-behavior charts, are a statistical process control tool used to determine if a manufacturing or business process is in a state of control. It is more appropriate to say that the control charts are the graphical device for Statistical Process Monitoring (SPM). Traditional control charts are mostly designed to monitor process parameters when underlying form of the process distributions are known. However, more advanced techniques are available in the 21st century where incoming data streaming can be monitored even without any knowledge of the underlying process distributions. Distribution-free control charts are becoming increasingly popular.

### Value

Generally called for the side effect of printing the control chart. Invisibly, returns a ggplot object for further customization.

### Author(s)

Steven P. Sanderson II, MPH

### Examples

```

data_tbl <- tibble::tibble(
  day = sample(c("Monday", "Tuesday", "Wednesday", "Thursday", "Friday"),
             100, TRUE),
  person = sample(c("Tom", "Jane", "Alex"), 100, TRUE),
  count = rbinom(100, 20, ifelse(day == "Friday", .5, .2)),
  date = Sys.Date() - sample.int(100))

```

```

hai_control_chart(.data = data_tbl, .value_col = count, .x_col = date)

# In addition to printing or writing the plot to file, hai_control_chart
# returns the plot as a ggplot2 object, which you can then further customize

library(ggplot2)
my_chart <- hai_control_chart(data_tbl, count, date)
my_chart +
  ylab("Number of Adverse Events") +
  scale_x_date(name = "Week of ... ", date_breaks = "week") +
  theme(axis.text.x = element_text(angle = -90, vjust = 0.5, hjust=1))

```

---

hai\_fourier\_augment     *Augment Function Fourier*

---

## Description

Takes a numeric vector(s) or date and will return a tibble of one of the following:

- "sin"
- "cos"
- "sincos"
- c("sin","cos","sincos")

## Usage

```

hai_fourier_augment(
  .data,
  .value,
  .period,
  .order,
  .names = "auto",
  .scale_type = c("sin", "cos", "sincos")
)

```

## Arguments

.data	The data being passed that will be augmented by the function.
.value	This is passed <code>rlang::enquo()</code> to capture the vectors you want to augment.
.period	The number of observations that complete a cycle
.order	The fourier term order
.names	The default is "auto"
.scale_type	A character of one of the following: "sin","cos", or sincos" All can be passed by setting the param equal to c("sin","cos","sincos")

**Details**

Takes a numeric vector or date and will return a vector of one of the following:

- "sin"
- "cos"
- "sincos"
- c("sin", "cos", "sincos")

This function is intended to be used on its own in order to add columns to a tibble.

**Value**

A augmented tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Augment Function: [hai\\_fourier\\_discrete\\_augment\(\)](#), [hai\\_hyperbolic\\_augment\(\)](#), [hai\\_polynomial\\_augment\(\)](#)

**Examples**

```
suppressPackageStartupMessages(library(dplyr))

len_out    = 10
by_unit    = "month"
start_date = as.Date("2021-01-01")

data_tbl <- tibble(
  date_col = seq.Date(from = start_date, length.out = len_out, by = by_unit),
  a        = rnorm(len_out),
  b        = runif(len_out)
)

hai_fourier_augment(data_tbl, b, .period = 12, .order = 1, .scale_type = "sin")
hai_fourier_augment(data_tbl, b, .period = 12, .order = 1, .scale_type = "cos")
```

## Description

Takes a numeric vector(s) or date and will return a tibble of one of the following:

- "sin"
- "cos"
- "sincos"
- c("sin","cos","sincos") When either of these values falls below zero, then zero else one

## Usage

```
hai_fourier_discrete_augment(  
  .data,  
  .value,  
  .period,  
  .order,  
  .names = "auto",  
  .scale_type = c("sin", "cos", "sincos")  
)
```

## Arguments

<code>.data</code>	The data being passed that will be augmented by the function.
<code>.value</code>	This is passed <code>rlang::enquo()</code> to capture the vectors you want to augment.
<code>.period</code>	The number of observations that complete a cycle
<code>.order</code>	The fourier term order
<code>.names</code>	The default is "auto"
<code>.scale_type</code>	A character of one of the following: "sin","cos", or "sincos" All can be passed by setting the param equal to c("sin","cos","sincos")

## Details

Takes a numeric vector or a date and will return a vector of one of the following:

- "sin"
- "cos"
- "sincos"
- c("sin","cos","sincos")

This function is intended to be used on its own in order to add columns to a tibble.

## Value

A augmented tibble

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Augment Function: [hai\\_fourier\\_augment\(\)](#), [hai\\_hyperbolic\\_augment\(\)](#), [hai\\_polynomial\\_augment\(\)](#)

**Examples**

```
suppressPackageStartupMessages(library(dplyr))

len_out   = 24
by_unit   = "month"
start_date = as.Date("2021-01-01")

data_tbl <- tibble(
  date_col = seq.Date(from = start_date, length.out = len_out, by = by_unit),
  a        = rnorm(len_out),
  b        = runif(len_out)
)

hai_fourier_discrete_augment(data_tbl, b, .period = 2*12, .order = 1, .scale_type = "sin")
hai_fourier_discrete_augment(data_tbl, b, .period = 2*12, .order = 1, .scale_type = "cos")
```

---

hai\_fourier\_discrete\_vec

*Vector Function Discrete Fourier*

---

**Description**

Takes a numeric vector or date and will return a vector of one of the following:

- "sin"
- "cos"
- "sincos" This will do value =  $\sin(x) * \cos(x)$  When either of these values falls below zero, then zero else one

**Usage**

```
hai_fourier_discrete_vec(
  .x,
  .period,
  .order,
  .scale_type = c("sin", "cos", "sincos")
)
```

**Arguments**

.x	A numeric vector
.period	The number of observations that complete a cycle
.order	The fourier term order
.scale_type	A character of one of the following: "sin","cos","sincos"



## Details

Takes a numeric vector or date and will return a vector of one of the following:

- "sin"
- "cos"
- "sincos"

The internal calculation is straightforward:

- $\sin = \sin(2 * \pi * h * x)$ , where  $h = .order/.period$
- $\cos = \cos(2 * \pi * h * x)$ , where  $h = .order/.period$
- $\text{sincos} = \sin(2 * \pi * h * x) * \cos(2 * \pi * h * x)$  where  $h = .order/.period$

This function can be used on its own. It is also the basis for the function [hai\\_fourier\\_discrete\\_augment\(\)](#).

## Value

A numeric vector of 1's and 0's

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Vector Function: [hai\\_fourier\\_vec\(\)](#), [hai\\_hyperbolic\\_vec\(\)](#)

## Examples

```
suppressPackageStartupMessages(library(dplyr))

len_out    = 24
by_unit    = "month"
start_date = as.Date("2021-01-01")

data_tbl <- tibble(
  date_col = seq.Date(from = start_date, length.out = len_out, by = by_unit),
  a        = rnorm(len_out),
  b        = runif(len_out)
)

vec_1 <- hai_fourier_discrete_vec(data_tbl$a, .period = 12, .order = 1, .scale_type = "sin")
vec_2 <- hai_fourier_discrete_vec(data_tbl$a, .period = 12, .order = 1, .scale_type = "cos")
vec_3 <- hai_fourier_discrete_vec(data_tbl$a, .period = 12, .order = 1, .scale_type = "sincos")

plot(data_tbl$b)
lines(vec_1, col = "blue")
lines(vec_2, col = "red")
lines(vec_3, col = "green")
```

---

hai\_fourier\_vec      *Vector Function Fourier*

---

### Description

Takes a numeric vector and will return a vector of one of the following:

- "sin"
- "cos"
- "sincos" This will do value =  $\sin(x) * \cos(x)$

### Usage

```
hai_fourier_vec(.x, .period, .order, .scale_type = c("sin", "cos", "sincos"))
```

### Arguments

.x	A numeric vector
.period	The number of observations that complete a cycle
.order	The fourier term order
.scale_type	A character of one of the following: "sin","cos","sincos"

### Details

Takes a numeric vector and will return a vector of one of the following:

- "sin"
- "cos"
- "sincos"

The internal calculation is straightforward:

- $\sin = \sin(2 * \pi * h * x)$ , where  $h = .order / .period$
- $\cos = \cos(2 * \pi * h * x)$ , where  $h = .order / .period$
- $\text{sincos} = \sin(2 * \pi * h * x) * \cos(2 * \pi * h * x)$  where  $h = .order / .period$

This function can be used on it's own. It is also the basis for the function [hai\\_fourier\\_augment\(\)](#).

### Value

A numeric vector

### Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Vector Function: [hai\\_fourier\\_discrete\\_vec\(\)](#), [hai\\_hyperbolic\\_vec\(\)](#)

**Examples**

```
suppressPackageStartupMessages(library(dplyr))

len_out    = 25
by_unit    = "month"
start_date = as.Date("2021-01-01")

data_tbl <- tibble(
  date_col = seq.Date(from = start_date, length.out = len_out, by = by_unit),
  a        = rnorm(len_out),
  b        = runif(len_out)
)

vec_1 <- hai_fourier_vec(data_tbl$b, .period = 12, .order = 1, .scale_type = "sin")
vec_2 <- hai_fourier_vec(data_tbl$b, .period = 12, .order = 1, .scale_type = "cos")
vec_3 <- hai_fourier_vec(data_tbl$date_col, .period = 12, .order = 1, .scale_type = "sincos")

plot(data_tbl$b)
lines(vec_1, col = "blue")
lines(vec_2, col = "red")
lines(vec_3, col = "green")
```

---

hai\_hyperbolic\_augment

*Augment Function Hyperbolic*

---

**Description**

Takes a numeric vector(s) or date and will return a tibble of one of the following:

- "sin"
- "cos"
- "tan"
- "sincos"
- c("sin", "cos", "tan", "sincos")

**Usage**

```
hai_hyperbolic_augment(
  .data,
  .value,
  .names = "auto",
  .scale_type = c("sin", "cos", "tan", "sincos")
)
```

**Arguments**

<code>.data</code>	The data being passed that will be augmented by the function.
<code>.value</code>	This is passed <code>rlang::enquo()</code> to capture the vectors you want to augment.
<code>.names</code>	The default is "auto"
<code>.scale_type</code>	A character of one of the following: "sin","cos","tan", "sincos" All can be passed by setting the param equal to <code>c("sin","cos","tan","sincos")</code>

**Details**

Takes a numeric vector or date and will return a vector of one of the following:

- "sin"
- "cos"
- "tan"
- "sincos"
- `c("sin","cos","tan", "sincos")`

This function is intended to be used on its own in order to add columns to a tibble.

**Value**

A augmented tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Augment Function: [hai\\_fourier\\_augment\(\)](#), [hai\\_fourier\\_discrete\\_augment\(\)](#), [hai\\_polynomial\\_augment\(\)](#)

**Examples**

```
suppressPackageStartupMessages(library(dplyr))

len_out    = 10
by_unit    = "month"
start_date = as.Date("2021-01-01")

data_tbl <- tibble(
  date_col = seq.Date(from = start_date, length.out = len_out, by = by_unit),
  a        = rnorm(len_out),
  b        = runif(len_out)
)

hai_hyperbolic_augment(data_tbl, b, .scale_type = "sin")
hai_hyperbolic_augment(data_tbl, b, .scale_type = "tan")
```

---

hai\_hyperbolic\_vec      *Vector Function Hyperbolic*

---

### Description

Takes a numeric vector and will return a vector of one of the following:

- "sin"
- "cos"
- "tan"
- "sincos" This will do value =  $\sin(x) * \cos(x)$

### Usage

```
hai_hyperbolic_vec(.x, .scale_type = c("sin", "cos", "tan", "sincos"))
```

### Arguments

.x                    A numeric vector  
.scale\_type        A character of one of the following: "sin","cos","tan","sincos"

### Details

Takes a numeric vector and will return a vector of one of the following:

- "sin"
- "cos"
- "tan"
- "sincos"

This function can be used on it's own. It is also the basis for the function [hai\\_hyperbolic\\_augment\(\)](#).

### Value

A numeric vector

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Vector Function: [hai\\_fourier\\_discrete\\_vec\(\)](#), [hai\\_fourier\\_vec\(\)](#)

**Examples**

```

suppressPackageStartupMessages(library(dplyr))

len_out    = 25
by_unit    = "month"
start_date = as.Date("2021-01-01")

data_tbl <- tibble(
  date_col = seq.Date(from = start_date, length.out = len_out, by = by_unit),
  a        = rnorm(len_out),
  b        = runif(len_out)
)

vec_1 <- hai_hyperbolic_vec(data_tbl$b, .scale_type = "sin")
vec_2 <- hai_hyperbolic_vec(data_tbl$b, .scale_type = "cos")
vec_3 <- hai_hyperbolic_vec(data_tbl$b, .scale_type = "sincos")

plot(data_tbl$b)
lines(vec_1, col = "blue")
lines(vec_2, col = "red")
lines(vec_3, col = "green")

```

---

hai\_kmeans\_automl      *Automatic K-Means H2O*

---

**Description**

This is a wrapper around the `h2o:h2o.kmeans()` function that will return a list object with a lot of useful and easy to use tidy style information.

**Usage**

```

hai_kmeans_automl(
  .data,
  .split_ratio = 0.8,
  .seed = 1234,
  .centers = 10,
  .standardize = TRUE,
  .print_model_summary = TRUE,
  .predictors,
  .categorical_encoding = "auto",
  .initialization_mode = "Furthest",
  .max_iterations = 100
)

```

**Arguments**

- `.data` The data that is to be passed for clustering.
- `.split_ratio` The ratio for training and testing splits.
- `.seed` The default is 1234, but can be set to any integer.
- `.centers` The default is 1. Specify the number of clusters (groups of data) in a data set.
- `.standardize` The default is set to TRUE. When TRUE all numeric columns will be set to zero mean and unit variance.
- `.print_model_summary`  
This is a boolean and controls if the model summary is printed to the console. The default is TRUE.
- `.predictors` This must be in the form of `c("column_1", "column_2", ... "column_n")`
- `.categorical_encoding`  
Can be one of the following:
- "auto"
  - "enum"
  - "one\_hot\_explicit"
  - "binary"
  - "eigen"
  - "label\_encoder"
  - "sort\_by\_response"
  - "enum\_limited"
- `.initialization_mode`  
This can be one of the following:
- "Random"
  - "Furthest (default)"
  - "PlusPlus"
- `.max_iterations`  
The default is 100. This specifies the number of training iterations

**Value**

A list object

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Kmeans: [hai\\_kmeans\\_automl\\_predict\(\)](#), [hai\\_kmeans\\_mapped\\_tbl\(\)](#), [hai\\_kmeans\\_obj\(\)](#), [hai\\_kmeans\\_scree\\_data\\_tbl\(\)](#), [hai\\_kmeans\\_scree\\_plt\(\)](#), [hai\\_kmeans\\_tidy\\_tbl\(\)](#), [hai\\_kmeans\\_user\\_item\\_tbl\(\)](#)

## Examples

```
## Not run:
h2o.init()
output <- hai_kmeans_automl(
  .data = iris,
  .predictors = c("Sepal.Width", "Sepal.Length", "Petal.Width", "Petal.Length"),
  .standardize = FALSE
)
h2o.shutdown()

## End(Not run)
```

---

hai\_kmeans\_automl\_predict  
*Automatic K-Means H2O*

---

## Description

This is a wrapper around the `h2o::h2o.predict()` function that will return a list object with a lot of useful and easy to use tidy style information.

## Usage

```
hai_kmeans_automl_predict(.input)
```

## Arguments

`.input` This is the output of the `hai_kmeans_automl()` function.

## Details

This function will internally take in the output assigned from the `hai_kmeans_automl()` function only and return a list of useful information. The items that are returned are as follows:

1. `prediction` - The h2o dataframe of predictions
2. `prediction_tbl` - The h2o predictions in tibble format
3. `valid_tbl` - The validation data in tibble format
4. `pred_full_tbl` - The entire validation set with the predictions attached using `base::cbind()`. The predictions are in a column called `predicted_cluster` and are in the format of a factor using `forcats::as_factor()`

## Value

A list object



**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Kmeans: [hai\\_kmeans\\_automl\(\)](#), [hai\\_kmeans\\_mapped\\_tbl\(\)](#), [hai\\_kmeans\\_obj\(\)](#), [hai\\_kmeans\\_scee\\_data\\_tbl\(\)](#), [hai\\_kmeans\\_scee\\_plt\(\)](#), [hai\\_kmeans\\_tidy\\_tbl\(\)](#), [hai\\_kmeans\\_user\\_item\\_tbl\(\)](#)

**Examples**

```
## Not run:
h2o.init()

output <- hai_kmeans_automl(
  .data = iris,
  .predictors = c("Sepal.Width", "Sepal.Length", "Petal.Width", "Petal.Length"),
  .standardize = FALSE
)

pred <- hai_kmeans_automl_predict(output)

h2o.shutdown()

## End(Not run)
```

---

hai\_kmeans\_mapped\_tbl *K-Means Mapping Function*

---

**Description**

Create a tibble that maps the [hai\\_kmeans\\_obj\(\)](#) using `purrr::map()` to create a nested data.frame/tibble that holds n centers. This tibble will be used to help create a scree plot.

**Usage**

```
hai_kmeans_mapped_tbl(.data, .centers = 15)
```

**Arguments**

<code>.data</code>	You must have a tibble in the working environment from the <a href="#">hai_kmeans_user_item_tbl()</a>
<code>.centers</code>	How many different centers do you want to try

**Details**

Takes in a single parameter of `.centers`. This is used to create the tibble and map the [hai\\_kmeans\\_obj\(\)](#) function down the list creating a nested tibble.

**Value**

A nested tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

[https://en.wikipedia.org/wiki/Scree\\_plot](https://en.wikipedia.org/wiki/Scree_plot)

Other Kmeans: [hai\\_kmeans\\_automl\\_predict\(\)](#), [hai\\_kmeans\\_automl\(\)](#), [hai\\_kmeans\\_obj\(\)](#), [hai\\_kmeans\\_scree\\_data\\_tbl\(\)](#), [hai\\_kmeans\\_scree\\_plt\(\)](#), [hai\\_kmeans\\_tidy\\_tbl\(\)](#), [hai\\_kmeans\\_user\\_item\\_tbl\(\)](#)

**Examples**

```
library(healthyR.data)
library(dplyr)

data_tbl <- healthyR_data%>%
  filter(ip_op_flag == "I") %>%
  filter(payer_grouping != "Medicare B") %>%
  filter(payer_grouping != "?") %>%
  select(service_line, payer_grouping) %>%
  mutate(record = 1) %>%
  as_tibble()

ui_tbl <- hai_kmeans_user_item_tbl(
  .data      = data_tbl
  , .row_input = service_line
  , .col_input = payer_grouping
  , .record_input = record
)

hai_kmeans_mapped_tbl(ui_tbl)
```

---

hai\_kmeans\_obj

*K-Means Object*

---

**Description**

Takes the output of the [hai\\_kmeans\\_user\\_item\\_tbl\(\)](#) function and applies the k-means algorithm to it using `stats::kmeans()`

**Usage**

```
hai_kmeans_obj(.data, .centers = 5)
```

**Arguments**

.data            The data that gets passed from [hai\\_kmeans\\_user\\_item\\_tbl\(\)](#)  
.centers        How many initial centers to start with

**Details**

Uses the [stats::kmeans\(\)](#) function and creates a wrapper around it.

**Value**

A stats k-means object

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Kmeans: [hai\\_kmeans\\_automl\\_predict\(\)](#), [hai\\_kmeans\\_automl\(\)](#), [hai\\_kmeans\\_mapped\\_tbl\(\)](#),  
[hai\\_kmeans\\_scree\\_data\\_tbl\(\)](#), [hai\\_kmeans\\_scree\\_plt\(\)](#), [hai\\_kmeans\\_tidy\\_tbl\(\)](#), [hai\\_kmeans\\_user\\_item\\_tbl\(\)](#)

**Examples**

```
library(healthyR.data)
library(dplyr)

data_tbl <- healthyR_data%>%
  filter(ip_op_flag == "I") %>%
  filter(payer_grouping != "Medicare B") %>%
  filter(payer_grouping != "?") %>%
  select(service_line, payer_grouping) %>%
  mutate(record = 1) %>%
  as_tibble()

hai_kmeans_user_item_tbl(
  .data            = data_tbl
  , .row_input     = service_line
  , .col_input     = payer_grouping
  , .record_input = record
) %>%
hai_kmeans_obj()
```

---

`hai_kmeans_scree_data_tbl`*K-Means Scree Plot Data Table*

---

### Description

Take data from the `hai_kmeans_mapped_tbl()` and unnest it into a tibble for inspection and for use in the `hai_kmeans_scree_plt()` function.

### Usage

```
hai_kmeans_scree_data_tbl(.data)
```

### Arguments

`.data` You must have a tibble in the working environment from the `hai_kmeans_mapped_tbl()`

### Details

Takes in a single parameter of `.data` from `hai_kmeans_mapped_tbl()` and transforms it into a tibble that is used for `hai_kmeans_scree_plt()`. It will show the values (tot.withinss) at each center.

### Value

A nested tibble

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Kmeans: `hai_kmeans_automl_predict()`, `hai_kmeans_automl()`, `hai_kmeans_mapped_tbl()`, `hai_kmeans_obj()`, `hai_kmeans_scree_plt()`, `hai_kmeans_tidy_tbl()`, `hai_kmeans_user_item_tbl()`

### Examples

```
library(healthyR.data)
library(dplyr)

data_tbl <- healthyR_data%>%
  filter(ip_op_flag == "I") %>%
  filter(payer_grouping != "Medicare B") %>%
  filter(payer_grouping != "?") %>%
  select(service_line, payer_grouping) %>%
  mutate(record = 1) %>%
  as_tibble()

ui_tbl <- hai_kmeans_user_item_tbl(
```

```
.data      = data_tbl
, .row_input = service_line
, .col_input = payer_grouping
, .record_input = record
)

kmm_tbl <- hai_kmeans_mapped_tbl(ui_tbl)

hai_kmeans_scree_data_tbl(kmm_tbl)
```

---

*hai\_kmeans\_scree\_plt*    *K-Means Scree Plot*

---

### **Description**

Create a scree-plot from the [hai\\_kmeans\\_mapped\\_tbl\(\)](#) function.

### **Usage**

```
hai_kmeans_scree_plt(.data)
```

### **Arguments**

`.data`            The data from the [hai\\_kmeans\\_mapped\\_tbl\(\)](#) function

### **Details**

Outputs a scree-plot

### **Value**

A `ggplot2` plot

### **Author(s)**

Steven P. Sanderson II, MPH

### **See Also**

[https://en.wikipedia.org/wiki/Scree\\_plot](https://en.wikipedia.org/wiki/Scree_plot)

Other Kmeans: [hai\\_kmeans\\_automl\\_predict\(\)](#), [hai\\_kmeans\\_automl\(\)](#), [hai\\_kmeans\\_mapped\\_tbl\(\)](#), [hai\\_kmeans\\_obj\(\)](#), [hai\\_kmeans\\_scree\\_data\\_tbl\(\)](#), [hai\\_kmeans\\_tidy\\_tbl\(\)](#), [hai\\_kmeans\\_user\\_item\\_tbl\(\)](#)

**Examples**

```

library(healthyR.data)
library(dplyr)
library(tidyquant)

data_tbl <- healthyR_data%>%
  filter(ip_op_flag == "I") %>%
  filter(payer_grouping != "Medicare B") %>%
  filter(payer_grouping != "?") %>%
  select(service_line, payer_grouping) %>%
  mutate(record = 1) %>%
  as_tibble()

ui_tbl <- hai_kmeans_user_item_tbl(
  .data      = data_tbl
  , .row_input = service_line
  , .col_input = payer_grouping
  , .record_input = record
)

kmm_tbl <- hai_kmeans_mapped_tbl(ui_tbl)

hai_kmeans_scree_plt(.data = kmm_tbl)

```

---

hai\_kmeans\_tidy\_tbl    *K-Means Object Tidy Functions*

---

**Description**

K-Means tidy functions

**Usage**

```
hai_kmeans_tidy_tbl(.kmeans_obj, .data, .tidy_type = "tidy")
```

**Arguments**

.kmeans_obj	A <code>stats::kmeans()</code> object
.data	The user item tibble created from <code>hai_kmeans_user_item_tbl()</code>
.tidy_type	"tidy", "glance", or "augment"

**Details**

Takes in a k-means object and its associated user item tibble and then returns one of the items asked for. Either: `broom::tidy()`, `broom::glance()` or `broom::augment()`. The function defaults to `broom::tidy()`.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Kmeans: [hai\\_kmeans\\_automl\\_predict\(\)](#), [hai\\_kmeans\\_automl\(\)](#), [hai\\_kmeans\\_mapped\\_tbl\(\)](#), [hai\\_kmeans\\_obj\(\)](#), [hai\\_kmeans\\_scree\\_data\\_tbl\(\)](#), [hai\\_kmeans\\_scree\\_plt\(\)](#), [hai\\_kmeans\\_user\\_item\\_tbl\(\)](#)

**Examples**

```
library(healthyR.data)
library(dplyr)
library(broom)

data_tbl <- healthyR_data%>%
  filter(ip_op_flag == "I") %>%
  filter(payer_grouping != "Medicare B") %>%
  filter(payer_grouping != "?") %>%
  select(service_line, payer_grouping) %>%
  mutate(record = 1) %>%
  as_tibble()

uit_tbl <- hai_kmeans_user_item_tbl(
  .data      = data_tbl
  , .row_input = service_line
  , .col_input = payer_grouping
  , .record_input = record
)

km_obj <- hai_kmeans_obj(uit_tbl)

hai_kmeans_tidy_tbl(
  .kmeans_obj = km_obj
  , .data      = uit_tbl
  , .tidy_type = "augment"
)

hai_kmeans_tidy_tbl(
  .kmeans_obj = km_obj
  , .data      = uit_tbl
  , .tidy_type = "glance"
)

hai_kmeans_tidy_tbl(
  .kmeans_obj = km_obj
  , .data      = uit_tbl
  , .tidy_type = "tidy"
) %>%
```

```
glimpse()
```

---

```
hai_kmeans_user_item_tbl
```

*K-Means User Item Tibble*

---

### Description

Takes in a `data.frame/tibble` and transforms it into an aggregated/normalized user-item tibble of proportions. The user will need to input the parameters for the rows/user and the columns/items.

### Usage

```
hai_kmeans_user_item_tbl(.data, .row_input, .col_input, .record_input)
```

### Arguments

<code>.data</code>	The data that you want to transform
<code>.row_input</code>	The column that is going to be the row (user)
<code>.col_input</code>	The column that is going to be the column (item)
<code>.record_input</code>	The column that is going to be summed up for the aggregation and normalization process.

### Details

This function should be used before using a k-mean model. This is commonly referred to as a user-item matrix because "users" tend to be on the rows and "items" (e.g. orders) on the columns. You must supply a column that can be summed for the aggregation and normalization process to occur.

### Value

A aggregated/normalized user item tibble

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Kmeans: [hai\\_kmeans\\_automl\\_predict\(\)](#), [hai\\_kmeans\\_automl\(\)](#), [hai\\_kmeans\\_mapped\\_tbl\(\)](#), [hai\\_kmeans\\_obj\(\)](#), [hai\\_kmeans\\_scree\\_data\\_tbl\(\)](#), [hai\\_kmeans\\_scree\\_plt\(\)](#), [hai\\_kmeans\\_tidy\\_tbl\(\)](#)



**Examples**

```

library(healthyR.data)
library(dplyr)

data_tbl <- healthyR_data%>%
  filter(ip_op_flag == "I") %>%
  filter(payer_grouping != "Medicare B") %>%
  filter(payer_grouping != "?") %>%
  select(service_line, payer_grouping) %>%
  mutate(record = 1) %>%
  as_tibble()

hai_kmeans_user_item_tbl(
  .data = data_tbl
  , .row_input = service_line
  , .col_input = payer_grouping
  , .record_input = record
)

```

---

```
hai_polynomial_augment
```

*Augment Polynomial Features*

---

**Description**

This function takes in a data table and a predictor column. A user can either create their own formula using the `.formula` parameter or, if they leave the default of `NULL` then the user must enter a `.degree` **AND** `.pred_col` column.

**Usage**

```

hai_polynomial_augment(
  .data,
  .formula = NULL,
  .pred_col = NULL,
  .degree = 1,
  .new_col_prefix = "nt_"
)

```

**Arguments**

<code>.data</code>	The data being passed that will be augmented by the function.
<code>.formula</code>	This should be a valid formula like <code>'y ~ .^2'</code> or <code>NULL</code> .
<code>.pred_col</code>	This is passed <code>rlang::enquo()</code> to capture the vector that you designate as the 'y' column.

- `.degree` This should be an integer and is used to set the degree in the poly function. The degree must be less than the unique data points or it will error out.
- `.new_col_prefix` The default is "nt\_" which stands for "new\_term". You can set this to whatever you like, as long as it is a quoted string.

### Details

A valid data.frame/tibble must be passed to this function. It is required that a user either enter a `.formula` or a `.degree` **AND** `.pred_col` otherwise this function will stop and error out.

Under the hood this function will create a `stats::poly()` function if the `.formula` is left as NULL. For example:

- `.formula = A ~ .^2`
- OR `.degree = 2, .pred_col = A`

There is also a parameter `.new_col_prefix` which will add a character string to the column names so that they are easily identified further down the line. The default is 'nt\_'

### Value

An augmented tibble

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Augment Function: [hai\\_fourier\\_augment\(\)](#), [hai\\_fourier\\_discrete\\_augment\(\)](#), [hai\\_hyperbolic\\_augment\(\)](#)

### Examples

```
suppressPackageStartupMessages(library(dplyr))
data_tbl <- data.frame(
  A = c(0,2,4),
  B = c(1,3,5),
  C = c(2,4,6)
)

hai_polynomial_augment(.data = data_tbl, .pred_col = A, .degree = 2, .new_col_prefix = "n")
hai_polynomial_augment(.data = data_tbl, .formula = A ~ .^2, .degree = 1)
```

---

pca_your_recipe	<i>Perform PCA</i>
-----------------	--------------------

---

**Description**

This is a simple function that will perform PCA analysis on a passed recipe.

**Usage**

```
pca_your_recipe(.recipe_object, .data, .threshold = 0.75)
```

**Arguments**

- `.recipe_object` The recipe object you want to pass.
- `.data` The full data set that is used in the original recipe object passed into `.recipe_object` in order to obtain the baked data of the transform.
- `.threshold` A number between 0 and 1. A fraction of the total variance that should be covered by the components.

**Details**

This is a simple wrapper around some recipes functions to perform a PCA on a given recipe. This function will output a list and return it invisible. All of the components of the analysis will be returned in a list as their own object that can be selected individually. A scree plot is also included. The items that get returned are:

1. `pca_transform` - This is the `pca` recipe.
2. `variable_loadings`
3. `variable_variance`
4. `pca_estimates`
5. `pca_juiced_estimates`
6. `pca_baked_data`
7. `pca_variance_df`
8. `pca_variance_scree_plt`
9. `pca_rotation_df`

**Value**

A list object with several components.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

[https://recipes.tidymodels.org/reference/step\\_pca.html](https://recipes.tidymodels.org/reference/step_pca.html)

Other Data Wrangling: [get\\_juiced\\_data\(\)](#)

Other Data Recipes: [hai\\_data\\_impute\(\)](#), [hai\\_data\\_poly\(\)](#), [hai\\_data\\_scale\(\)](#), [hai\\_data\\_transform\(\)](#), [hai\\_data\\_trig\(\)](#)

**Examples**

```
suppressPackageStartupMessages(library(timetk))
suppressPackageStartupMessages(library(dplyr))
suppressPackageStartupMessages(library(purrr))
suppressPackageStartupMessages(library(healthyR.data))
suppressPackageStartupMessages(library(rsample))
suppressPackageStartupMessages(library(recipes))
suppressPackageStartupMessages(library(ggplot2))

data_tbl <- healthyR_data %>%
  select(visit_end_date_time) %>%
  summarise_by_time(
    .date_var = visit_end_date_time,
    .by       = "month",
    value     = n()
  ) %>%
  set_names("date_col", "value") %>%
  filter_by_time(
    .date_var = date_col,
    .start_date = "2013",
    .end_date = "2020"
  )

splits <- initial_split(data = data_tbl, prop = 0.8)

rec_obj <- recipe(value ~ ., training(splits)) %>%
  step_timeseries_signature(date_col) %>%
  step_rm(matches("(iso$)|(xts$)|(hour)|(min)|(sec)|(am.pm)"))

output_list <- pca_your_recipe(rec_obj, .data = data_tbl)
output_list$pca_variance_scee_plt
```

---

step\_hai\_fourier

*Recipes Step Fourier Generator*

---

**Description**

step\_hai\_fourier creates a *specification* of a recipe step that will convert numeric data into either a 'sin', 'cos', or 'sincos' feature that can aid in machine learning.

**Usage**

```
step_hai_fourier(
  recipe,
  ...,
  role = "predictor",
  trained = FALSE,
  columns = NULL,
  scale_type = c("sin", "cos", "sincos"),
  period = 1,
  order = 1,
  skip = FALSE,
  id = rand_id("hai_fourier")
)
```

**Arguments**

recipe	A recipe object. The step will be added to the sequence of operations for this recipe.
...	One or more selector functions to choose which variables that will be used to create the new variables. The selected variables should have class <code>numeric</code>
role	For model terms created by this step, what analysis role should they be assigned?. By default, the function assumes that the new variable columns created by the original variables will be used as predictors in a model.
trained	A logical to indicate if the quantities for preprocessing have been estimated.
columns	A character string of variables that will be used as inputs. This field is a placeholder and will be populated once <code>recipes::prep()</code> is used.
scale_type	A character string of a scaling type, one of "sin", "cos", or "sincos"
period	The number of observations that complete a cycle
order	The fourier term order
skip	A logical. Should the step be skipped when the recipe is baked by <code>bake.recipe()</code> ? While all operations are baked when <code>prep.recipe()</code> is run, some operations may not be able to be conducted on new data (e.g. processing the outcome variable(s)). Care should be taken when using <code>skip = TRUE</code> as it may affect the computations for subsequent operations.
id	A character string that is unique to this step to identify it.

**Details**

**Numeric Variables** Unlike other steps, `step_hai_fourier` does *not* remove the original numeric variables. `recipes::step_rm()` can be used for this purpose.

**Value**

For `step_hai_fourier`, an updated version of `recipe` with the new step added to the sequence of existing steps (if any).

Main Recipe Functions:

- `recipes::recipe()`
- `recipes::prep()`
- `recipes::bake()`

### See Also

Other Recipes: [step\\_hai\\_fourier\\_discrete\(\)](#), [step\\_hai\\_hyperbolic\(\)](#)

### Examples

```
suppressPackageStartupMessages(library(dplyr))
suppressPackageStartupMessages(library(recipes))

len_out    = 10
by_unit    = "month"
start_date = as.Date("2021-01-01")

data_tbl <- tibble(
  date_col = seq.Date(from = start_date, length.out = len_out, by = by_unit),
  a        = rnorm(len_out),
  b        = runif(len_out)
)

# Create a recipe object
rec_obj <- recipe(a ~ ., data = data_tbl) %>%
  step_hai_fourier(b, scale_type = "sin") %>%
  step_hai_fourier(b, scale_type = "cos") %>%
  step_hai_fourier(b, scale_type = "sincos")

# View the recipe object
rec_obj

# Prepare the recipe object
prep(rec_obj)

# Bake the recipe object - Adds the Time Series Signature
bake(prep(rec_obj), data_tbl)

rec_obj %>% get_juiced_data()
```

---

step\_hai\_fourier\_discrete

*Recipes Step Fourier Discrete Generator*

---

### Description

`step_hai_fourier_discrete` creates a *specification* of a recipe step that will convert numeric data into either a 'sin', 'cos', or 'sincos' feature that can aid in machine learning.

**Usage**

```
step_hai_fourier_discrete(
  recipe,
  ...,
  role = "predictor",
  trained = FALSE,
  columns = NULL,
  scale_type = c("sin", "cos", "sincos"),
  period = 1,
  order = 1,
  skip = FALSE,
  id = rand_id("hai_fourier_discrete")
)
```

**Arguments**

recipe	A recipe object. The step will be added to the sequence of operations for this recipe.
...	One or more selector functions to choose which variables that will be used to create the new variables. The selected variables should have class numeric or date, POSIXct
role	For model terms created by this step, what analysis role should they be assigned?. By default, the function assumes that the new variable columns created by the original variables will be used as predictors in a model.
trained	A logical to indicate if the quantities for preprocessing have been estimated.
columns	A character string of variables that will be used as inputs. This field is a placeholder and will be populated once <code>recipes::prep()</code> is used.
scale_type	A character string of a scaling type, one of "sin", "cos", or "sincos"
period	The number of observations that complete a cycle
order	The fourier term order
skip	A logical. Should the step be skipped when the recipe is baked by <code>bake.recipe()</code> ? While all operations are baked when <code>prep.recipe()</code> is run, some operations may not be able to be conducted on new data (e.g. processing the outcome variable(s)). Care should be taken when using <code>skip = TRUE</code> as it may affect the computations for subsequent operations.
id	A character string that is unique to this step to identify it.

**Details**

**Numeric Variables** Unlike other steps, `step_hai_fourier_discrete` does *not* remove the original numeric variables. `recipes::step_rm()` can be used for this purpose.

**Value**

For `step_hai_fourier_discrete`, an updated version of recipe with the new step added to the sequence of existing steps (if any).

Main Recipe Functions:

- `recipes::recipe()`
- `recipes::prep()`
- `recipes::bake()`

### See Also

Other Recipes: [step\\_hai\\_fourier\(\)](#), [step\\_hai\\_hyperbolic\(\)](#)

### Examples

```
suppressPackageStartupMessages(library(dplyr))
suppressPackageStartupMessages(library(recipes))

len_out    = 10
by_unit    = "month"
start_date = as.Date("2021-01-01")

data_tbl <- tibble(
  date_col = seq.Date(from = start_date, length.out = len_out, by = by_unit),
  a        = rnorm(len_out),
  b        = runif(len_out)
)

# Create a recipe object
rec_obj <- recipe(a ~ ., data = data_tbl) %>%
  step_hai_fourier_discrete(b, scale_type = "sin") %>%
  step_hai_fourier_discrete(b, scale_type = "cos") %>%
  step_hai_fourier_discrete(b, scale_type = "sincos")

# View the recipe object
rec_obj

# Prepare the recipe object
prep(rec_obj)

# Bake the recipe object - Adds the Time Series Signature
bake(prep(rec_obj), data_tbl)

rec_obj %>% get_juiced_data()
```

---

step\_hai\_hyperbolic    *Recipes Step Hyperbolic Generator*

---

### Description

`step_hai_hyperbolic` creates a *specification* of a recipe step that will convert numeric data into either a 'sin', 'cos', or 'tan' feature that can aid in machine learning.



**Usage**

```
step_hai_hyperbolic(
  recipe,
  ...,
  role = "predictor",
  trained = FALSE,
  columns = NULL,
  scale_type = c("sin", "cos", "tan", "sincos"),
  skip = FALSE,
  id = rand_id("hai_hyperbolic")
)
```

**Arguments**

recipe	A recipe object. The step will be added to the sequence of operations for this recipe.
...	One or more selector functions to choose which variables that will be used to create the new variables. The selected variables should have class <code>numeric</code>
role	For model terms created by this step, what analysis role should they be assigned?. By default, the function assumes that the new variable columns created by the original variables will be used as predictors in a model.
trained	A logical to indicate if the quantities for preprocessing have been estimated.
columns	A character string of variables that will be used as inputs. This field is a placeholder and will be populated once <code>recipes::prep()</code> is used.
scale_type	A character string of a scaling type, one of "sin", "cos", "tan" or "sincos"
skip	A logical. Should the step be skipped when the recipe is baked by <code>bake.recipe()</code> ? While all operations are baked when <code>prep.recipe()</code> is run, some operations may not be able to be conducted on new data (e.g. processing the outcome variable(s)). Care should be taken when using <code>skip = TRUE</code> as it may affect the computations for subsequent operations.
id	A character string that is unique to this step to identify it.

**Details**

**Numeric Variables** Unlike other steps, `step_hai_hyperbolic` does *not* remove the original numeric variables. `recipes::step_rm()` can be used for this purpose.

**Value**

For `step_hai_hyperbolic`, an updated version of `recipe` with the new step added to the sequence of existing steps (if any).

Main Recipe Functions:

- `recipes::recipe()`
- `recipes::prep()`
- `recipes::bake()`

**See Also**

Other Recipes: [step\\_hai\\_fourier\\_discrete\(\)](#), [step\\_hai\\_fourier\(\)](#)

**Examples**

```
suppressPackageStartupMessages(library(dplyr))
suppressPackageStartupMessages(library(recipes))

len_out    = 10
by_unit    = "month"
start_date = as.Date("2021-01-01")

data_tbl <- tibble(
  date_col = seq.Date(from = start_date, length.out = len_out, by = by_unit),
  a        = rnorm(len_out),
  b        = runif(len_out)
)

# Create a recipe object
rec_obj <- recipe(a ~ ., data = data_tbl) %>%
  step_hai_hyperbolic(b, scale_type = "sin") %>%
  step_hai_hyperbolic(b, scale_type = "cos")

# View the recipe object
rec_obj

# Prepare the recipe object
prep(rec_obj)

# Bake the recipe object - Adds the Time Series Signature
bake(prep(rec_obj), data_tbl)

rec_obj %>% get_juiced_data()
```

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