# Package 'evsim'

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```

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Adapt charging features

## **Description**

Calculate connection and charging times according to energy, power and time resolution

# Usage

```
adapt_charging_features(
  sessions,
  time_resolution = 15,
  power_resolution = 0.01
)
```

## **Arguments**

```
sessions tibble, sessions data set in standard format marked by {evprof} package (see this article)

time_resolution integer, time resolution (in minutes) of the sessions' datetime variables

power_resolution numeric, power resolution (in kW) of the sessions' power
```

## **Details**

All sessions' Power must be higher than 0, to avoid NaN values from dividing by zero. The ConnectionStartDateTime is first aligned to the desired time resolution, and the ConnectionEndDateTime is calculated according to the ConnectionHours. The ChargingHours is recalculated with the values of Energy and Power, limited by ConnectionHours. Finally, the charging times are also calculated.

## Value

tibble

## **Examples**

```
suppressMessages(library(dplyr))
sessions <- head(evsim::california_ev_sessions, 10)
sessions %>%
    select(ConnectionStartDateTime, ConnectionEndDateTime, Power)
adapt_charging_features(
    sessions,
    time_resolution = 60,
    power_resolution = 0.01
) %>%
    select(ConnectionStartDateTime, ConnectionEndDateTime, Power)
adapt_charging_features(
    sessions,
    time_resolution = 15,
    power_resolution = 1
) %>%
    select(ConnectionStartDateTime, ConnectionEndDateTime, Power)
```

 ${\tt add\_charging\_infrastructure}$ 

Assign a charging station to EV charging sessions

## **Description**

Variable ChargingStation and Socketwill be assigned to the sessions tibble with a name pattern being: names\_prefix + "CHS" + number

## Usage

```
add_charging_infrastructure(
  sessions,
  resolution = 15,
  min_stations = 0,
  n_sockets = 2,
  names_prefix = NULL,
  duration_th = 0
)
```

#### Arguments

sessions tibble, sessions data set in standard format marked by {evprof} package (see

this article)

resolution integer, time resolution in minutes

min\_stations integer, minimum number of charging stations to consider

n\_sockets integer, number of sockets per charging station

names\_prefix character, prefix of the charging station names (optional)

duration\_th integer between 0 and 100, minimum share of time (in percentage) of the "occu-

pancy duration curve" (see function plot\_occupancy\_duration\_curve). This is used to avoid sizing a charging infrastructure to host for example 100 vehicles when only 5% of time there are more than 80 vehicles connected. Then, setting duration\_th = 5 will ensure that we don't over-size the charging infrastructure for the 100 vehicles. It is recommended to find this value through multiple

iterations.

#### Value

tibble

```
# Assign a `ChargingStation` to every session according to the occupancy
sessions_infrastructure <- add_charging_infrastructure(
   sessions = head(evsim::california_ev_sessions, 50),
   resolution = 60
)
print(unique(sessions_infrastructure$ChargingStation))

# Now without considering the occupancy values that only represent
# a 10% of the time
sessions_infrastructure <- add_charging_infrastructure(
   sessions = head(evsim::california_ev_sessions, 50),
   resolution = 60, duration_th = 10
)
print(unique(sessions_infrastructure$ChargingStation))</pre>
```

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exnand	sessions

Expand sessions along time slots

## **Description**

Every session in sessions is divided in multiple time slots with the corresponding Power consumption, among other variables.

## Usage

```
expand_sessions(sessions, resolution)
```

## **Arguments**

sessions tibble, sessions data set in standard format marked by evprof package (see this

article)

resolution integer, time resolution (in minutes) of the time slots

## **Details**

The Power value is calculated for every time slot according to the original required energy. The columns PowerNominal, EnergyRequired and FlexibilityHours correspond to the values of the original session, and not to the expanded session in every time slot. The column ID shows the number of the time slot corresponding to the original session.

## Value

tibble

```
library(dplyr)
sessions <- head(evsim::california_ev_sessions, 10)
expand_sessions(
   sessions,
   resolution = 60
)</pre>
```

## **Description**

Get charging rates distribution in percentages from a charging sessions data set

## Usage

```
get_charging_rates_distribution(sessions, unit = "year", power_interval = NULL)
```

## Arguments

sessions tibble, sessions data set in standard format marked by {evprof} package (see

this article)t

unit character. Valid base units are second, minute, hour, day, week, month, bimonth,

quarter, season, halfyear and year. It corresponds to unit parameter in

lubridate::floor\_date function.

power\_interval numeric, interval of kW between power rates. It is used to round the Power

values into this interval resolution. It can also be NULL to use all the original

Power values.

#### Value

tibble

## **Examples**

```
get_charging_rates_distribution(evsim::california_ev_sessions, unit = "year")
```

get\_custom\_ev\_model

Create the custom EV model

## **Description**

Get the EV model object of class evmodel

get\_custom\_ev\_model

## Usage

```
get_custom_ev_model(
  names,
  months_lst = list(1:12, 1:12),
  wdays_lst = list(1:5, 6:7),
  parameters_lst,
  connection_log,
  energy_log,
  data_tz
)
```

# Arguments

names	character vector with the given names of each time-cycle model
months_lst	list of integer vectors with the corresponding months of the year for each time-cycle model
wdays_lst	list of integer vectors with the corresponding days of the week for each time-cycle model (week start = $1$ )
parameters_lst	list of tibbles corresponding to the GMM parameters of every time-cycle model
connection_log	logical, true if connection models have logarithmic transformations
energy_log	logical, true if energy models have logarithmic transformations
data_tz	character, time zone of the original data (necessary to properly simulate new sessions)

## Value

object of class evmodel

```
# For workdays time cycle
workdays_parameters <- dplyr::tibble(</pre>
  profile = c("Worktime", "Visit"),
  ratio = c(80, 20),
  start_mean = c(9, 11),
  start_sd = c(1, 4),
  duration_mean = c(8, 4),
  duration_sd = c(0.5, 2),
  energy_mean = c(15, 6),
  energy_sd = c(4, 3)
)
# For weekends time cycle
weekends_parameters <- dplyr::tibble(</pre>
  profile = "Visit",
  ratio = 100,
  start_mean = 12,
  start_sd = 4,
```

get\_demand

```
duration_mean = 3,
  duration\_sd = 2,
  energy_mean = 4,
  energy_sd = 4
)
parameters_lst <- list(workdays_parameters, weekends_parameters)</pre>
# Get the whole model
ev_model <- get_custom_ev_model(</pre>
  names = c("Workdays", "Weekends"),
  months_lst = list(1:12, 1:12),
  wdays_lst = list(1:5, 6:7),
  parameters_lst = parameters_lst,
  connection_log = FALSE,
  energy_log = FALSE,
  data_tz = "Europe/Amsterdam"
)
```

get\_demand

Time-series EV demand

# Description

Obtain time-series of EV demand from sessions data set

## Usage

```
get_demand(
   sessions,
   dttm_seq = NULL,
   by = "Profile",
   resolution = 15,
   mc.cores = 1
)
```

## **Arguments**

sessions	tibble, sessions data set in standard format marked by {evprof} package (see this article)
dttm_seq	sequence of datetime values that will be the datetime variable of the returned time-series data frame.
by	character, being 'Profile' or 'Session'. When by='Profile' each column corresponds to an EV user profile.
resolution	integer, time resolution (in minutes) of the sessions datetime variables. If dttm_seq is defined this parameter is ignored.

mc.cores

integer, number of cores to use. Must be at least one, and parallelization requires at least two cores.

#### **Details**

Note that the time resolution of variables ConnectionStartDateTime and ChargingStartDateTime must coincide with resolution parameter. For example, if a charging session in sessions starts charging at 15:32 and resolution = 15, the load of this session won't be computed. To solve this, the function automatically aligns charging sessions' start time according to resolution, so following the previous example the session would start at 15:30.

#### Value

time-series tibble with first column of type datetime

## **Examples**

```
suppressMessages(library(lubridate))
suppressMessages(library(dplyr))
# Get demand with the complete datetime sequence from the sessions
sessions <- head(evsim::california_ev_sessions, 100)</pre>
demand <- get_demand(</pre>
 sessions,
 by = "Session",
 resolution = 60
demand %>% plot_ts(ylab = "EV demand (kW)")
# Get demand with a custom datetime sequence and resolution of 15 minutes
sessions <- head(evsim::california_ev_sessions_profiles, 100)</pre>
dttm_seq <- seq.POSIXt(</pre>
 as_datetime(dmy(08102018)) %>% force_tz(tz(sessions$ConnectionStartDateTime)),
 as_datetime(dmy(11102018)) %>% force_tz(tz(sessions$ConnectionStartDateTime)),
 by = "15 \text{ mins}"
)
demand <- get_demand(</pre>
 sessions,
 dttm_seq = dttm_seq,
 by = "Profile",
 resolution = 15
demand %>% plot_ts(ylab = "EV demand (kW)")
```

get\_evmodel\_parameters

Get evmodel parameters in a list

## **Description**

Every time cycle is an element of the returned list, containing a list with the user profile as elements, each one containing the ratio and the corresponding tables with the statistic parameters of connection and energy GMM.

## Usage

```
get_evmodel_parameters(evmodel)
```

## Arguments

evmodel

object of class evmodel

## Value

list

# **Examples**

```
get_evmodel_parameters(evsim::california_ev_model)
```

get\_evmodel\_summary

Get evmodel parameters in a list of summary tables

## **Description**

Every time cycle is an element of the returned list, containing a table with a user profile in every row and the mean and standard deviation values of the GMM variables (connection duration, connection start time and energy). If the energy models were built by charging rate, the average mean and sd are provided without taking into account different charging rates (this information is lost in this summary).

## Usage

```
get_evmodel_summary(evmodel)
```

## **Arguments**

evmodel

object of class evmodel

#### Value

list

```
get_evmodel_summary(evsim::california_ev_model)
```

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get_occupancy	Time-series EV occupancy	

# Description

Obtain time-series of simultaneously connected EVs from sessions data set

## Usage

```
get_occupancy(
  sessions,
  dttm_seq = NULL,
  by = "Profile",
  resolution = 15,
  mc.cores = 1
)
```

## **Arguments**

sessions	tibble, sessions data set in standard format marked by {evprof} package (see this article)
dttm_seq	sequence of datetime values that will be the datetime variable of the returned time-series data frame.
by	character, being 'Profile' or 'Session'. When by='Profile' each column corresponds to an EV user profile.
resolution	integer, time resolution (in minutes) of the sessions date time variables. If ${\tt dttm\_seq}$ is defined this parameter is ignored.
mc.cores	integer, number of cores to use. Must be at least one, and parallelization requires at least two cores.

## **Details**

Note that the time resolution of variable ConnectionStartDateTime must coincide with resolution parameter. For example, if a charging session in sessions starts charging at 15:32 and resolution = 15, the load of this session won't be computed. To solve this, the function automatically aligns charging sessions' start time according to resolution, so following the previous example the session would start at 15:30.

## Value

time-series tibble with first column of type datetime

## **Examples**

```
library(lubridate)
library(dplyr)
# Get occupancy with the complete datetime sequence from the sessions
sessions <- head(evsim::california_ev_sessions, 100)</pre>
connections <- get_occupancy(</pre>
  sessions,
  by = "ChargingStation",
  resolution = 60
connections %>%
  plot_ts(ylab = "Vehicles connected", legend_show = "onmouseover")
# Get occupancy with a custom datetime sequence and resolution of 15 minutes
sessions <- head(evsim::california_ev_sessions_profiles, 100)</pre>
dttm_seg <- seg.POSIXt(</pre>
  as_datetime(dmy(08102018)) %>% force_tz(tz(sessions$ConnectionStartDateTime)),
  as_datetime(dmy(11102018)) %>% force_tz(tz(sessions$ConnectionStartDateTime)),
  by = "15 \text{ mins}"
)
connections <- get_occupancy(</pre>
  sessions,
  dttm_seq = dttm_seq,
  by = "Profile"
)
connections %>%
  plot_ts(ylab = "Vehicles connected", legend_show = "onmouseover")
```

# **Description**

Get the user profiles distribution from the original data set used to build the model

## Usage

```
get_user_profiles_distribution(evmodel)
```

## **Arguments**

evmodel object of class evmodel

## Value

tibble

## **Examples**

```
get_user_profiles_distribution(evsim::california_ev_model)
```

```
\verb|plot_occupancy_duration_curve|\\
```

Plot the occupancy duration curve

# Description

This term is based on the "load duration curve" and is useful to see the behavior of occupancy over the time in your charging installation. The steeper the curve, the shorter the duration that higher number of connections are sustained. Conversely, the flatter the curve, the longer the duration that higher number of connections are sustained. This information is crucial for various purposes, such as infrastructure planning, capacity sizing, and resource allocation.

## Usage

```
plot_occupancy_duration_curve(
   sessions,
   dttm_seq = NULL,
   by = "Profile",
   resolution = 15,
   mc.cores = 1
)
```

## **Arguments**

sessions	tibble, sessions data set in standard format marked by {evprof} package (see this article)
dttm_seq	sequence of datetime values that will be the datetime variable of the returned time-series data frame.
by	character, being 'Profile' or 'Session'. When by='Profile' each column corresponds to an EV user profile.
resolution	integer, time resolution (in minutes) of the sessions date time variables. If ${\tt dttm\_seq}$ is defined this parameter is ignored.
mc.cores	integer, number of cores to use. Must be at least one, and parallelization requires at least two cores.

#### Value

ggplot

plot\_ts

## **Examples**

```
library(dplyr)
sessions <- head(evsim::california_ev_sessions_profiles, 100)
plot_occupancy_duration_curve(
   sessions,
   by = "Profile",
   resolution = 15
)</pre>
```

plot\_ts

Interactive plot for time-series tibbles

#### **Description**

First column of the df tibble must be a datetime or date variable. The rest of columns must be numeric of the same units. This functions makes use of dygraphs package to generate an HTML Dygraphs plot.

#### Usage

```
plot_ts(
    df,
    title = NULL,
    xlab = NULL,
    ylab = NULL,
    legend_show = "auto",
    legend_width = 250,
    group = NULL,
    width = NULL,
    height = NULL,
    ...
)
```

## **Arguments**

df data.frame or tibble, first column of name datetime being of class datetime and

rest of columns being numeric

title character, title of the plot (accepts HTML code)
xlab character, X axis label (accepts HTML code)
ylab character, Y axis label (accepts HTML code)

legend\_show character, when to display the legend. Specify "always" to always show the

legend. Specify "onmouseover" to only display it when a user mouses over the chart. Specify "follow" to have the legend show as overlay to the chart which follows the mouse. The default behavior is "auto", which results in "always" when more than one series is plotted and "onmouseover" when only a single parties in plotted.

series is plotted.

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legend\_width integer, width (in pixels) of the div which shows the legend.

group character, dygraphs group to associate this plot with. The x-axis zoom level of

dygraphs plots within a group is automatically synchronized.

width Width in pixels (optional, defaults to automatic sizing)

height Height in pixels (optional, defaults to automatic sizing)

extra arguments to pass to dygraphs::dyOptions function.

## Value

dygraph

## **Examples**

```
suppressMessages(library(lubridate))
suppressMessages(library(dplyr))

# Get demand with the complete datetime sequence from the sessions
sessions <- head(evsim::california_ev_sessions, 100)
demand <- get_demand(
    sessions,
    by = "Session",
    resolution = 60
)
demand %>% plot_ts()
```

read\_ev\_model

Read EV model

# Description

Read an EV model JSON file and convert it to object of class evmodel

## Usage

```
read_ev_model(file)
```

# Arguments

file

path to the JSON file

#### Value

object of class evmodel

simulate\_sessions

## **Examples**

```
ev_model <- california_ev_model # Model of example
save_ev_model(ev_model, file = file.path(tempdir(), "evmodel.json"))
read_ev_model(file = file.path(tempdir(), "evmodel.json"))</pre>
```

save\_ev\_model

Save the EV model

## **Description**

Save the EV model object of class evmodel to a JSON file

## Usage

```
save_ev_model(evmodel, file)
```

## **Arguments**

evmodel object of class evmodel (see this link for more information) file character string with the path or name of the file

## Value

nothing but saves the evmodel object in a JSON file

# **Examples**

```
ev_model <- california_ev_model # Model of example
save_ev_model(ev_model, file = file.path(tempdir(), "evmodel.json"))</pre>
```

simulate\_sessions

Simulation of EV sessions

## Description

Simulate EV charging sessions given the evmodel object and other contextual parameters.

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## Usage

```
simulate_sessions(
  evmodel,
  sessions_day,
  user_profiles,
  charging_powers,
  dates,
  resolution
)
```

## **Arguments**

evmodel object of class evmodel built with {evprof} (see this link for more information)

sessions\_day tibble with variables time\_cycle (names corresponding to evmodel\$models\$time\_cycle)

and n\_sessions (number of daily sessions per day for each time-cycle model)

user\_profiles tibble with variables time\_cycle, profile, ratio and optionally power. It can

also be NULL to use the evmodel original user profiles distribution. The powers must be in kW and the ratios between 0 and 1. The user profiles with a value of power will be simulated with this specific charging power. If power is NA then it

is simulated according to the ratios of next parameter charging\_powers.

charging\_powers

tibble with variables power and ratio. The powers must be in kW and the ratios between 0 and 1. This is used to simulate the charging power of user profiles

without a specific charging power in user\_profiles parameter.

dates date sequence that will set the time frame of the simulated sessions

resolution integer, time resolution (in minutes) of the sessions datetime variables

#### **Details**

Some adaptations have been done to the output of the Gaussian models: the minimum simulated energy is considered to be 1 kWh, while the minimum connection duration is 30 minutes.

#### Value

tibble

```
library(dplyr)
library(lubridate)

# Get the example `evmodel`
ev_model <- evsim::california_ev_model

# Simulate EV charging sessions, considering that the Worktime sessions
# during Workdays have 11 kW, while all Visit sessions charge at 3.7kW or
# 11kW, with a distribution of 30% and 70% respectively.</pre>
```

simulate\_sessions

```
simulate_sessions(
  ev_model,
  sessions_day = tibble(
   time_cycle = c("Workday", "Weekend"),
   n_{sessions} = c(15, 10)
 ),
  user_profiles = tibble(
   time_cycle = c("Workday", "Workday", "Weekend"),
   profile = c("Visit", "Worktime", "Visit"),
   ratio = c(0.5, 0.5, 1),
   power = c(NA, 11, NA)
  ),
  charging_powers = tibble(
   power = c(3.7, 11),
   ratio = c(0.3, 0.7)
  dates = seq.Date(today(), today()+days(4), length.out = 4),
  resolution = 15
)
```

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