# Package 'TSAR'

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```
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Description This package automates analysis workflow for Thermal Shift
      Analysis (TSA) data. Processing, analyzing, and visualizing data through
      both shiny applications and command lines. Package aims to simplify
        data analysis and offer front to end workflow, from raw data to
        multiple trial analysis.
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analyze\_norm

Analyze to Normalize

# Description

The analyze\_norm function allows users to process analysis through an UI interface. Function wraps together all functions with in TSA\_analysis family and read\_write\_analysis family.

# Usage

```
analyze_norm(raw_data)
```

# **Arguments**

raw\_data T

The raw data for analysis.

# Value

shiny application

# See Also

```
gam_analysis, read_tsar, write_tsar, join_well_info
```

```
if (interactive()) {
    data("qPCR_data1")
    shiny::runApp(analyze_norm(qPCR_data1))
}
```

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condition\_IDs

TSAR Condition IDs

# **Description**

This function is used to extract information of the condition IDs from a loaded TSA Analysis Data file. Condition IDs are automatically generated by the read\_analysis function in the automated workflow. This returns either a character vector of unique IDs present or a numeric value of the number of unique IDs.

# Usage

```
condition_IDs(analysis_data, n = FALSE)
```

## **Arguments**

n

analysis\_data a data frame that is unmerged and generated by TSAR::read\_analysis() or a merged TSA data frame generated by TSAR::merge\_TSA(). Data frames require a column named 'condition\_ID'.

logical value; n = FALSE by default. When TRUE, a numeric value of unique IDs is returned. When FALSE, a character vector of unique IDs are returned.

#### Value

Either a character vector of condition\_IDs or a numeric value.

## See Also

```
merge_TSA and read_analysis for preparing input.
Other TSA Summary Functions: TSA_ligands(), TSA_proteins(), well_IDs()
```

```
data("example_tsar_data")
condition_IDs(example_tsar_data)
```

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example\_tsar\_data

Example tsar\_data file

# Description

Dataset Description: This is an example dataset of the tsar\_data strucutre. The data frame contains well ID, conditions, and experimental details.

# Usage

```
data(example_tsar_data)
```

#### **Format**

A data frame with the following columns:

Well Well position

**Temperature** Temperature in degrees

Fluorescence Fluorescence reading

Normalized Normalized value

norm\_deriv Calculated first derivative

Tm Tm value

Protein Protein information

Ligand Ligand information

ExperimentFileName Experiment file name

well\_ID Well ID

condition\_ID Condition ID

# Value

example tsar\_data in data frame

# Source

experimentally obtained

gam\_analysis

| gam_analysis | Analysis of all 96 wells through gam modeling |  |
|--------------|---|--|
|              |   |  |

# Description

Function pipeline that combines separated functions and iterate through each well to estimate the Tm.

# Usage

```
gam_analysis(
  raw_data,
  keep = TRUE,
  fit = FALSE,
  smoothed = FALSE,
  boltzmann = FALSE,
  fluo_col = NA,
  selections = c("Well.Position", "Temperature", "Fluorescence", "Normalized")
)
```

# Arguments

| raw_data   | data frame; raw data frame  |
|------------|---|
| keep       | Boolean; set to keep = TRUE by default to return normalized data and fitted data  |
| fit        | Boolean; set to fit = FALSE by default, fit = TRUE returns access to information of each model fit. Not accessible in shiny.  |
| smoothed   | Boolean; set to smoothed = FALSE by default, if data is already smoothed, set smoothed to true  |
| boltzmann  | Boolean; set to boltzmann = FALSE by default. Set to boltzmann = TRUE if a botlzmann fit is preferred.  |
| fluo_col   | integer; the Fluorescence variable column id (e.g. fluo = 5 when 5th column of data frame is the Fluorescence value) if fluorescence variable is named exactly as "Fluorescence", fluo does not need to be specified. |
| selections | list of characters; the variables in raw data user intends to keep. It is set, by default, to c("Well.Position", "Temperature", "Fluorescence", "Normalized").  |

# Value

List of data frames, list of three data frame outputs, Tm estimation by well, data set, fit of model by well.

# See Also

```
Other tsa_analysis: Tm_est()
```

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# **Examples**

```
data("qPCR_data1")
gam_analysis(qPCR_data1,
    smoothed = TRUE, boltzmann = FALSE, fluo_col = 5,
    selections = c(
        "Well.Position", "Temperature", "Fluorescence",
        "Normalized"
    )
)
model <- gam_analysis(qPCR_data1, smoothed = FALSE, boltzmann = TRUE)</pre>
```

get\_legend

Extract ggplot2 legend

# Description

Function enables separation of legends from plots within the TSAR package.

# Usage

```
get_legend(input_plot)
```

# **Arguments**

```
input_plot a ggplot2 object
```

#### Value

two ggplots, one containing the legend and another containing all else.

# See Also

```
Other TSA Plots: TSA_boxplot(), TSA_compare_plot(), TSA_wells_plot(), graph_tsar(), view_deriv()
```

join\_well\_info

graph\_tsar

Graph tsar\_data

# Description

Function allows users to graph out tsar\_data, building boxplot, compare plots, and curves by condition. Input of data as parameter is optional. graph\_tsar wraps together all graphing functions and relative helper functions.

# Usage

```
graph_tsar(tsar_data = data.frame())
```

# **Arguments**

tsar\_data

tsar data outputted by merge\_norm or merge\_tsa. Parameter is optional. If no data is passed, access the merge panel to merge norm\_data into tsar\_data.

## Value

prompts separate app window for user interaction, does not return specific value; generates boxplot and compare plots according to user input

## See Also

```
TSA_boxplot, TSA_compare_plot, condition_IDs, well_IDs, merge_norm, TSA_Tms, Tm_difference Other TSA Plots: TSA_boxplot(), TSA_compare_plot(), TSA_wells_plot(), get_legend(), view_deriv()
```

# **Examples**

```
if (interactive()) {
    data("example_tsar_data")
    shiny::runApp(graph_tsar(example_tsar_data))
}
```

join\_well\_info

Well information input function

# Description

Reads in the ligand and protein information and joins them accordingly to the big data frame for graphing purposes.

join\_well\_info

## Usage

```
join_well_info(
   file_path,
   file = NULL,
   analysis_file,
   skips = 0,
   nrows = 96,
   type
)
```

## **Arguments**

file\_path string; file path to read in the file file object; use file to override the need of file\_path if information is already read in analysis\_file data frame; data frame containing smoothed fluorescence data and tm values skips integer; number indicating the number of headers present in input file, default set to 0 when file input is "by\_well" If the input follows the excel template, this parameter does not apply. integer; number indicating the number of rows the data is. Default set to 96 asnrows suming analysis on 96 well plate. Parameter is only applicable when file input is "by\_well". If inputting by excel template, this parameter does not apply, please ignore. string; variable specifies the type of input read in. type = "by\_well" requires intype put of csv or txt files of three variables: Well, Protein, Ligand. type = "by template"

requires input of excel file following the template format provided

#### Value

outputs data frame joining data information with well information

#### See Also

```
Other read_write_analysis: read_tsar(), write_tsar()
```

```
data("qPCR_data1")
result <- gam_analysis(qPCR_data1, smoothed = TRUE, fluo = 5)
data("well_information")
join_well_info(
    file_path = NULL, file = well_information,
    read_tsar(result, output_content = 2), type = "by_template"
)</pre>
```

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| merge_norm Mea | ge and format norm_ | data into tsar data |
|----------------|---------------------|---------------------|
|----------------|---------------------|---------------------|

# **Description**

This function merges data of experiment replicates across different dates. It merges and produces information variables used to group wells of same set up.

# Usage

```
merge_norm(data, name, date)
```

# **Arguments**

data list, a character vector specifying the file paths of the data files or data frame

objects of analysis data set. For example, given data frames named "data1" and

"data2", specify parameter as data = list(data1, data2).

name list, character vector specifying the experiment names.

date list, character vector specifying the dates. Does not require any date format

restrictions.

#### **Details**

This function merges and normalizes test data from multiple files. The lengths of the data, name, and date vectors must match, otherwise an error is thrown.

#### Value

data frame in the format of tsar\_data

# See Also

```
Other TSAR Formatting: TSA_Tms(), TSA_average(), Tm_difference(), merge_TSA(), normalize_fluorescence(), rescale()
```

```
data("qPCR_data1")
result <- gam_analysis(qPCR_data1, smoothed = TRUE, fluo = 5)
data("well_information")
norm_data <- join_well_info(
    file_path = NULL, file = well_information,
    read_tsar(result, output_content = 2), type = "by_template"
)
norm_data <- na.omit(norm_data)
data("qPCR_data2")
result2 <- gam_analysis(qPCR_data1, smoothed = TRUE, fluo = 5)
norm_data2 <- join_well_info(</pre>
```

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```
file_path = NULL, file = well_information,
    read_tsar(result2, output_content = 2), type = "by_template"
)
norm_data2 <- na.omit(norm_data2)
tsar_data <- merge_norm(
    data = list(norm_data, norm_data2),
    name = c("Thermal Shift_162.eds.csv", "Thermal Shift_168.eds.csv"),
    date = c("20230203", "20230209")
)</pre>
```

merge\_TSA

Merge TSA Raw Data and Analysis Files

# **Description**

This function is used to load both the Raw Data and the Analysis Results which are returned by the TSA software. Both output files have unique information regarding the experiment, and these need reunited for downstream analysis. Automatically generated Well IDs are use to merge similar data within the same experiment from the different files. Both Raw Data and the Analysis Results files must be specified. The returned, merged results from this function are required for downstream analysis as the format is set up for the automated workflow.

# Usage

```
merge_TSA(analysis_file_path, raw_data_path, protein = NA, ligand = NA)
```

# **Arguments**

raw\_data\_path, analysis\_file\_path

a character string or vector of character strings; the path or the name of the file which the 'RawData' or "AnalysisData' are to be read from. Raw data and Analysis Data are to be loaded as a pair of results from the same experiment. When loading multiple files, the index/position of the pairs are merged where the first file specified by raw\_data\_path is to be merged with the first file specified by analysis\_file\_path. The same is done for the second, third, .. etc.

Either a .txt or .csv file; file type can vary between file pairs.

raw\_data\_path path must contain the term <code>RawData</code> and analysis\_file\_path must contain the term <code>AnalysisResults</code> as the TSA software automatically assigns this when exporting data. Data is loaded from the <code>read\_raw\_data</code> and <code>read\_analysis</code> functions within this merge\_TSA function.

protein

can be used to select for an individual or multiple protein(s) as a character string matching protein names assigned in the TSA software. NA by default.

ligand

can be used to select for an individual or multiple ligand(s) as a character string matching ligand names assigned in the TSA software. NA by default.

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#### Value

A data frame of merged TSA data.

#### **IDs**

The TSAR package relies on matching conditions and file names for each well and for each set of conditions between multiple files output by the TSA software. Conditions are assigned to individual wells within the TSA software; these assigned values are detected by <code>read\_analysis</code> and <code>read\_raw\_data</code> then are converted into IDs. Ensure your labeling of values within the TSA software is consistent so that similar values can be merged - typos or varying terms will be treated as distinct values within TSAR unless the values are manually specified by the user. Automatically generated well IDs within a TSA file can be found using the <code>well\_IDs</code> function; condition IDs can be found using the <code>condition\_IDs</code> function.

**Condition IDs** are generated only in the read\_analysis, see that function's documentation for more details. Condition IDs are assigned to raw data in the merge\_TSA function.

Well IDs are similar to Condition IDs, as they are generated from columns in TSA output. Well IDs are used to match the analysis and raw data files for the same experiment, as both files contain unique, useful information for each well. The well ID includes the .eds file name saved from the PCR machine to match equivalent wells between files of the same experiment. Each well on all plates should have a unique well ID. If you wish to change or specify the file name used for the well ID, a new name can be manually assigned with the "manual\_file" argument.

#### See Also

```
read_raw_data and read_analysis for loading data.
Other TSAR Formatting: TSA_Tms(), TSA_average(), Tm_difference(), merge_norm(), normalize_fluorescence(),
rescale()
```

## **Examples**

```
# note: example does not contain example data to run
# merge_TSA(analysis_file_path, raw_data_path)
```

model\_boltzmann

Boltzmann Modeling on TSA data

# **Description**

Function finds fitted fluorescence values by imposing Boltzmann function.

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## Usage

```
model_boltzmann(norm_data)
```

## **Arguments**

norm\_data data frame input, preferably normalized using normalize.

#### Value

dtaa frame containing gam model fitted values

#### See Also

```
Other data_preprocess: model_fit(), model_gam(), normalize(), remove_raw(), run_boltzmann(), screen(), view_model(), weed_raw()
```

# **Examples**

```
data("qPCR_data1")
A01 <- subset(qPCR_data1, Well.Position == "A01")
A01 <- normalize(A01)
model_boltzmann(A01)</pre>
```

model\_fit

Refit and calculate derivative function

# Description

Model\_fit calculates derivatives by refitting model onto data. Only runs on data of a single well.

# Usage

```
model_fit(norm_data, model, smoothed)
```

# **Arguments**

norm\_data data frame; the raw data set input model fitted model containing fitted values

smoothed inform whether data already contains a smoothed model; Input the column name

of the smoothed data to override values of gam model fitting. For example, existing "Fluorescence" column contains data already smoothed, set smoothed = "Flourescence" to calculate derivative function upon the called smoothed

data.

# Value

data frame; with calculated derivative columns

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## See Also

```
Other data_preprocess: model_boltzmann(), model_gam(), normalize(), remove_raw(), run_boltzmann(), screen(), view_model(), weed_raw()
```

# **Examples**

```
data("qPCR_data1")
test <- subset(qPCR_data1, Well.Position == "A01")
test <- normalize(test, fluo = 5, selected = c(
    "Well.Position", "Temperature",
    "Fluorescence", "Normalized"
))
gammodel <- model_gam(test, x = test$Temperature, y = test$Normalized)
model_fit(test, model = gammodel)
# if data come smoothed, run ...
model_fit(test, smoothed = "Fluorescence")</pre>
```

model\_gam

Generalized Addidtive Modeling on TSA data

# **Description**

Function finds fitted fluorescence values by imposing generalized additive model on fluorescence data by temperature. Model assumes method = "GACV.Cp" and sets to formula =  $y \sim s(x, bs = "ad")$ . Function inherits function from gam package, gam().

# Usage

```
model_gam(norm_data, x, y)
```

# **Arguments**

norm\_data data frame input of only one well's reading, preferably normalized using normalize.

x temperature column

y normalized fluorescence column

#### Value

data frame containing gam model fitted values

#### See Also

```
Other data_preprocess: model_boltzmann(), model_fit(), normalize(), remove_raw(), run_boltzmann(), screen(), view_model(), weed_raw()
```

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## **Examples**

```
data("qPCR_data1")
test <- subset(qPCR_data1, Well.Position == "A01")
test <- normalize(test, fluo = 5, selected = c(
    "Well.Position", "Temperature",
    "Fluorescence", "Normalized"
))
model_gam(test, x = test$Temperature, y = test$Normalized)</pre>
```

normalize

Normalize Fluorescence

# **Description**

normalize() reads in raw\_data. This function normalizes data by standardizing them according to maximum and minimum fluorescence per well, with maximum set to 1 and minimum set to 0. It also reformats data types by checking for potential error. i.e. a string specifying 100,000 will be read in as number, 100000, without issue. Function is applicable only to data of a single well, do not call on an entire data frame of all 96 well data. It is intended for single well screening purposes.

## Usage

```
normalize(
  raw_data,
  fluo = NA,
  selected = c("Well.Position", "Temperature", "Fluorescence", "Normalized")
)
```

#### **Arguments**

raw\_data data frame; raw dataset input, should be of only one well. If multiple wells

need to be normalized, use gam\_analysis() for 96 well application. If only

preliminary screening is needed, use screen().

fluo integer; the Fluorescence variable column id (e.g. fluo = 5 when 5th column

of the data frame is the Fluorescence value) if fluorescence variable is named exactly as "Fluorescence", fluo does not need to be specified. i.e. fluo is set to

NA by default, suggesting the variable is named "Fluorescence".

selected list of character strings; variables from the original data set users intend to keep.

Variable default set to c("Well.Position", "Temperature", "Fluorescence", "Normalized") if not otherwise specified. If data frame variables are named differ-

ently, user needs to specify what column variables to keep.

#### Value

cleaned up data framed with selected columns

#### See Also

```
Other data_preprocess: model_boltzmann(), model_fit(), model_gam(), remove_raw(), run_boltzmann(), screen(), view_model(), weed_raw()
```

# **Examples**

```
data("qPCR_data1")
test <- subset(qPCR_data1, Well.Position == "A01")
normalize(test)</pre>
```

normalize\_fluorescence

Normalize Fluorescence Curve

# **Description**

This function will take the TSA data and normalize the arbitrary fluorescence measurements based on the specified method. Each well, determined by a unique well ID, is normalized independently. All measurements can be normalized to the minimum or maximum value. Alternatively, setting by = "rescale" (the default) will cause all values to be normalized between the minimum and maximum values, with the maximum = 1 and the minimum = 0 and all other values normalized inbetween. Finally, the user can supply a single value or vector of values to normalize the data to. The returned data frame will be the input tsa data frame with a new column named "RFU" containing the normalized TSA data.

# Usage

```
normalize_fluorescence(tsa_data = tsa_data, by = "rescale", control_vect = NA)
```

## **Arguments**

tsa\_data

a data frame that is unmerged and generated by TSAR::read\_raw\_data() or a merged TSA data frame generated by TSAR::merge\_TSA(). Data frames require a column named "Fluorescence" containing numeric values for normalizing.

by

character string; either c("rescale", "min", "max", "control"). by = "rescale" by default, scaling Fluorescence values in-between the minimum and maximum observation. Each well can be normalized to either the minimum or maximum value with by = "min" or by = "max", respectively. To normalize all values to a numeric value or vector, set by = "control".

control\_vect

numeric vector to normalize the column "Fluorescence" to. An individual number will normalize all measurements to be normalized to it. The vector will need to align with tsa\_data\$Fluorescence. Ensure by = "control", else the supplied vector will be ignored.

qPCR\_data1

## Value

a data frame identical to the tsa\_data input with a new column named "RFU" containing the normalized values

#### See Also

```
read_raw_data and merge_TSA for loading data.
Other TSAR Formatting: TSA_Tms(), TSA_average(), Tm_difference(), merge_TSA(), merge_norm(),
rescale()
```

## **Examples**

```
# examples not ran without example dataset
# raw_data <- read_raw_data(raw_data_path)
# normalize_fluorescence(raw_data, by == "control)</pre>
```

qPCR\_data1

qPCR data1 Dataset

# **Description**

Dataset Description: This dataset contains qPCR data for the CA121 protein and common vitamins. It provides fluorescence measurements obtained using QuantStudio3. Dataset is experimentally obtained by author of this package.

# Usage

```
data(qPCR_data1)
```

## **Format**

A data frame with the following columns:

Well Well Count, not required for user

Well.Position Well Label, i.e. A01; required input

**Reading** reading count in time series, not required for user

Temperature temperature reading, required input

Fluorescence fluorescence reading, required input

# Value

```
qPCR_data1 data frame
```

# Source

experimentally obtained

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qPCR\_data2

qPCR\_data2 Dataset

## **Description**

Dataset Description: This dataset contains qPCR data for the CA121 protein and common vitamins. It provides fluorescence measurements obtained using QuantStudio3. A different experiemnt trial containing data of similar property as data, qPCR\_data1. Dataset is experimentally obtained by author of this package.

## Usage

data(qPCR\_data2)

#### **Format**

A data frame with the following columns:

Well Well Count, not required for user

Well.Position Well Label, i.e. A01; required input

**Reading** reading count in time series, not required for user

**Temperature** temperature reading, required input

Fluorescence fluorescence reading, required input

#### Value

qPCR\_data2 data frame

read\_analysis

Read TSA Analysis Data

#### **Description**

Open TSA Analysis files. This function is used to load data output from the thermal shift software analysis tab. Can be either .txt or .csv file with a path / file name including the string "Analysis-Results" due to its automatic naming from the software. The values assigned to wells within the TSA software are automatically extracted from the loaded file; values must be assigned within the TSA software for the automated workflow (See IDs Section Below). **Note:** Wells that do not have an Analysis Group assigned are removed. The TSA software automatically assigns all wells to Analysis Group 1 by default, and can be changed but not removed by the software.

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## Usage

```
read_analysis(
  path,
  type = "derivative",
  conditions = c("Protein", "Ligand"),
  manual_conditions = NA,
  manual_wells = NA,
  skip_flags = FALSE,
  manual_file = NA
)
```

#### **Arguments**

path a character string; the path or the name of the file which the 'AnalysisResults'

data are to be read from. Either a .txt or .csv file. The path must contain the term *AnalysisResults* as the TSA software automatically assigns this when exporting

data.

type either c("boltzmann", "derivative"); type = "derivative") by default. Deter-

mines what model of Tm estimation to load from the TSA software. Loads Tms as 'Tm B' when type = "boltzmann"); loads Tms as 'Tm D' when type

= "derivative").

conditions A character vector of condition types assigned within the TSA software to load.

conditions = c("Protein", "Ligand") by default. These conditions are used

to generate the IDs discussed.

manual\_conditions, manual\_wells

NA by default, enabling automated analysis. A character vector of Condition

IDs and Well IDs to manually assign each row of the read data.

skip\_flags logical value; type = FALSE by default. When type = TRUE, wells that have flags

reported by TSA software are removed.

manual\_file NA by default. User can specify .eds for merging if needed for Well IDs if

needed with a character string.

#### Value

A data frame of TSA analysis data.

## IDs

The TSAR package relies on matching conditions and file names for each well and for each set of conditions between multiple files output by the TSA software. Conditions are assigned to individual wells within the TSA software; these assigned values are detected by read\_analysis and read\_raw\_data then are converted into IDs. Ensure your labeling of values within the TSA software is consistent so that similar values can be merged - typos or varying terms will be treated as distinct values within TSAR unless the values are manually specified by the user. Automatically generated well IDs within a TSA file can be found using the well\_IDs function; condition IDs can be found using the condition\_IDs function.

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Condition IDs are generated from columns in TSA output specified by the 'conditions' argument. Protein and Ligand values, the default conditions within the TSA software, are the values used to create these IDs. You can manually specify the condition categories from the TSA software, including user-made conditions. Condition IDs are used to match equivalent observations between technical and biological replicates. Wells with identical condition IDs, specified by the 'conditions' argument, will be aggregated in down-stream analysis; user-specified conditions must remain consistent in use and order to create compatible IDs between TSA files from the same experiment and between replicates.

Well IDs are similar to Condition IDs, as they are generated from columns in TSA output that are specified by the 'conditions' argument. Well IDs are used to match the analysis and raw data files for the same experiment, as both files contain unique, useful information for each well. In addition to the condition ID, the well ID includes the .eds file name saved from the PCR machine to match equivalent wells between files of the same experiment. Each well on all plates should have a unique well ID. If you wish to change or specify the file name used for the well ID, a new name can be manually assigned with the "manual file" argument.

The user may manually assign condition IDs using the 'manual\_conditions' argument rather than using the automatically generated IDs. The same is true for well IDs, which can be manually assigned with 'manual\_wells'. This is not suggested, as there may be issues with matching if well/conditions are not properly matching. This gives the potential for errors in downstream applications as well.

# See Also

read\_raw\_data for loading accompanying data. merge\_TSA for joining Analysis Results and Raw Data files from the TSA software.

```
Other Read TSA Data: read_raw_data()
```

```
path <- "~/Desktop/analysis_data"
# note: example does not contain example data to run
# read_analysis(path)</pre>
```

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#### **Description**

Open TSA Raw Data files. This function is used to load data output from the thermal shift software Raw Data tab. Can be either .txt or .csv file with a path / file name including the string "Raw-Data" due to its automatic naming from the software. The values assigned to wells within the TSA software are automatically extracted from the loaded file; values must be assigned within the TSA software for the automated workflow (See IDs Section Below).

#### Usage

```
read_raw_data(path, manual_file = NA, type = "fluorescence")
```

## **Arguments**

path a character string; the path or the name of the file which the 'RawData' data

are to be read from. Either a .txt or .csv file. The path must contain the term *RawData* as the TSA software automatically assigns this when exporting data.

manual\_file NA by default. User can specify .eds for merging if needed for Well IDs with a

character string.

type either c("boltzmann", "derivative", "fluorescence"); type = "fluorescence")

by default. Determines what data track to load. When type = "fluorescence"), the arbitrary fluorescence of the TSA dye is loaded; this is the primary data. Alternately, derivatives van be loaded: Loads data as boltzman estimated tracks when type = "boltzmann"); loads the 2nd derivative of emissions when type

= "derivative").

#### Value

A data frame of TSA raw data.

#### IDs

The TSAR package relies on matching conditions and file names for each well and for each set of conditions between multiple files output by the TSA software. Conditions are assigned to individual wells within the TSA software; these assigned values are detected by read\_analysis and read\_raw\_data then are converted into IDs. Ensure your labeling of values within the TSA software is consistent so that similar values can be merged - typos or varying terms will be treated as distinct values within TSAR unless the values are manually specified by the user. Automatically generated well IDs within a TSA file can be found using the well\_IDs function; condition IDs can be found using the condition\_IDs function.

**Condition IDs** are generated only in the read\_analysis, see that function's documentation for more details. Condition IDs are assigned to raw data in the merge\_TSA function.

Well IDs are similar to Condition IDs, as they are generated from columns in TSA output. Well IDs are used to match the analysis and raw data files for the same experiment, as both files contain unique, useful information for each well. The well ID includes the .eds file name saved from the

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PCR machine to match equivalent wells between files of the same experiment. Each well on all plates should have a unique well ID. If you wish to change or specify the file name used for the well ID, a new name can be manually assigned with the "manual\_file" argument.

#### See Also

read\_analysis for loading accompanying data. merge\_TSA for joining Analysis Results and Raw Data files from the TSA software.

Other Read TSA Data: read\_analysis()

# **Examples**

```
path <- "~/Desktop/raw_data"
# note: example does not contain example data to run
# read_raw_data(path)</pre>
```

read\_tsar

Read analysis result

# **Description**

reads previous pipeline output lists from gam\_analysis() and organizes them into separate data frames.

## Usage

```
read_tsar(gam_result, output_content)
```

#### **Arguments**

gam\_result list; input uses resulting output of gam\_analysis() function

output\_content integer; output\_content = 0 returns only the tm value by wells output\_content

= 1 returns data table with fitted values output\_content = 2 returns the combi-

nation of 0 and 1

# Value

output files with select dataset

#### See Also

```
Other read_write_analysis: join_well_info(), write_tsar()
```

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# **Examples**

```
data("qPCR_data1")
result <- gam_analysis(qPCR_data1,
    smoothed = TRUE, fluo_col = 5,
    selections = c(
        "Well.Position", "Temperature", "Fluorescence", "Normalized"
    )
)
read_tsar(result, output_content = 0)
output_data <- read_tsar(result, output_content = 2)</pre>
```

remove\_raw

Remove selected raw curves

# Description

Removes selected curves with specified wells and range.

# Usage

```
remove_raw(raw_data, removerange = NULL, removelist = NULL)
```

# **Arguments**

raw\_data dataframe; to be processed data

removerange list type input identifying range of wells to select. For example, if removing all

12 wells from row D to H is needed, one can specify the row letters and column

numbers like this: removerange = c("D", "H", "1", "12")

removelist use this parameter to remove selected Wells with full Well names. For example,

removelist = c('A01', 'D11')

#### Value

dataframe; data frame with specified well removed

#### See Also

```
Other data_preprocess: model_boltzmann(), model_fit(), model_gam(), normalize(), run_boltzmann(), screen(), view_model(), weed_raw()
```

```
data("qPCR_data1")
remove_raw(qPCR_data1, removelist = c("A01", "D11"))
```

24 run\_boltzmann

rescale

Rescale values between minimum and maximum.

# **Description**

For a vector of numeric values, the minimum and maximum values are determined and each value of the vector is rescaled between 0 and 1. Values near 0 are close to the minimum, values near 1 are close to the max. This function is utilized by other TSAR functions.

# Usage

```
rescale(x)
```

# **Arguments**

Х

a numeric vector to be rescaled

## Value

A numeric vector of rescaled values.

## See Also

```
Other TSAR Formatting: TSA_Tms(), TSA_average(), Tm_difference(), merge_TSA(), merge_norm(), normalize_fluorescence()
```

# **Examples**

```
x <- c(0, 1, 3)
rescale(x)</pre>
```

run\_boltzmann

Run Boltzmann Modeling

# **Description**

Function runs function model\_boltzmann() and raises warning when modeling generates error or warnings.

# Usage

```
run_boltzmann(norm_data)
```

# **Arguments**

norm\_data

data frame input, preferably normalized using normalize.

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## Value

data frame containing gam model fitted values

#### See Also

```
Other data_preprocess: model_boltzmann(), model_fit(), model_gam(), normalize(), remove_raw(), screen(), view_model(), weed_raw()
```

# **Examples**

```
data("qPCR_data1")
A01 <- subset(qPCR_data1, Well.Position == "A01")
A01 <- normalize(A01)
run_boltzmann(A01)</pre>
```

screen

Screen raw curves

# Description

screens multiple wells of data and prepares to assist identification of corrupted wells and odd out behaviors

# Usage

```
screen(raw_data, checkrange = NULL, checklist = NULL)
```

## **Arguments**

raw\_data input raw\_data

checkrange list type input identifying range of wells to select. For example, if viewing first

8 wells from row A to C is needed, one can specify the row letters and column

numbers like this: checkrange = c("A", "C", "1", "8")

checklist use this parameter to view selected Wells with full Well names. For example,

checklist = c('A01', 'D11')

#### Value

returns a ggplot graph colors by well IDs

#### See Also

```
Other data_preprocess: model_boltzmann(), model_fit(), model_gam(), normalize(), remove_raw(), run_boltzmann(), view_model(), weed_raw()
```

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# **Examples**

```
data("qPCR_data1")
screen(qPCR_data1, checkrange = c("A", "C", "1", "12"))
```

Tm\_difference

Calculate Tm difference for all conditions

# **Description**

From a specified control condition, the change in Tm is calculated for each condition in the tsa\_data. Specifically, Tm = condition - control. Individual Tm values are averaged by condition, see TSA\_average for details. To see all conditions use condition\_IDs(tsa\_data).

#### Usage

```
Tm_difference(tsa_data, control_condition)
```

#### Arguments

tsa\_data

a data frame that is merged and generated by TSAR::merge\_TSA(). If y = 'RFU', tsa\_data must also be generated by TSAR::normalize\_fluorescence. The Temperature column will be rounded and the average & sd of each rounded temperature is calculated.

control\_condition

character string matching a Condition ID. Must be equal to a value within tsa\_data\$condition\_ID. See unique condition IDs with condition\_IDs.

## Value

a data frame of reformatted data with the TSA\_average data and the Tm.

#### See Also

```
merge_TSA for preparing data. TSA_average for more information on the output data. condition_IDs to get unique Condition IDs within the input. TSA_boxplot for application.
```

```
Other TSAR Formatting: TSA_Tms(), TSA_average(), merge_TSA(), merge_norm(), normalize_fluorescence(), rescale()
```

```
data("example_tsar_data")
control <- condition_IDs(example_tsar_data)[1]
Tm_difference(example_tsar_data, control_condition = control)</pre>
```

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Tm\_est

Find inflection point function

# **Description**

Looks for Tm temperature values by finding the inflection point in the fluorescence data. The inflection point is approximated by locating the maximum first derivative stored in "norm\_deriv" column.

# Usage

```
Tm_est(norm_data, min, max)
```

## **Arguments**

norm\_data data frame; data frame input containing derivative values can only be data frames

for one well; finding inflections points across multiple wells require iteration

through individual wells

min restricts finding to be above the given minimum temperature

max restricts finding to be below the given maximum temperature parameter min and

max can be used to remove messy or undesired data for better accuracy in tm estimation; removing data is before fitting the model is more recommended than

removing here

## Value

integer; tm estimation

## See Also

```
Other tsa_analysis: gam_analysis()
```

```
data("qPCR_data1")
test <- subset(qPCR_data1, Well.Position == "A01")
test <- normalize(test, fluo = 5, selected = c(
    "Well.Position", "Temperature",
    "Fluorescence", "Normalized"
))
gammodel <- model_gam(test, x = test$Temperature, y = test$Normalized)
fit <- model_fit(test, model = gammodel)
Tm_est(fit)</pre>
```

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TSA\_average

Average TSA Curves

#### **Description**

This function will take either Fluorescence or Normalized Fluorescence curves from the submitted data frame and find the average (mean) and standard deviation (sd) for each temperature measured in the TSA curve. Mean and sd are smoothened by default to generate cleaner curves. The function gam from the mgcv package is used for regression to smoothen lines. Smoothing can be turned off and the true average for each point can be given, however, plots will look messier. The qPCR machine may return temperatures with many decimal places, and TSAR only merges identical values, therefore rounding is necessary. Data is rounded to one decimal place to improve regression smoothing.

**Note:** All submitted data is averaged, regardless of condition or well ID. If you wish to average by condition, you will need to sort the data frame and run this function on subsets.

#### Usage

```
TSA_average(
  tsa_data,
  y = "Fluorescence",
  digits = 1,
  avg_smooth = TRUE,
  sd_smooth = TRUE
)
```

# Arguments

|                       | tsa_data | a data frame that is merged and generated by TSAR::merge_TSA(). If y = 'RFU', tsa_data must also be generated by TSAR::normalize_fluorescence. The  |
|-----------------------|----------|---|
|                       |          | Temperature column will be rounded and the average & sd of each rounded temperature is calculated.  |
|                       | У        | character string; c('Fluorescence', 'RFU'). When $y = 'Fluorescence'$ , the original Fluorescence data from TSAR::read_raw_data() is averaged. When $y = 'RFU'$ , the average is calculated by the rescaled fluorescence. |
|                       | digits   | an integer; digits = 1 by default. The number of decimal places to round temperature to for averaging.  |
| avg_smooth, sd_smooth |          |   |
|                       |          | logical; TRUE by default. Decides if the average (avg_smooth) or standard deviation (sd_smooth) will be smoothened by regression via mgcv::gam()  |

#### Value

a data frame of each temperature measured with the average, sd, and n(# of averaged values) calculated. Depending on avg\_smooth and sd\_smooth, the smoothened lines for the maximum and mimimum sd and the average will also be returned.

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## See Also

```
merge_TSA and merge_TSA for preparing data.
Other TSAR Formatting: TSA_Tms(), Tm_difference(), merge_TSA(), merge_norm(), normalize_fluorescence(),
rescale()
```

## **Examples**

```
data("example_tsar_data")
TSA_average(example_tsar_data,
    y = "Fluorescence", digits = 1,
    avg_smooth = TRUE, sd_smooth = TRUE)
```

TSA\_boxplot

TSA Box Plot

#### **Description**

Generates a box and whiskers plot for each condition specified. This is used to compare Tm values between the data set. See Tm\_difference for details.

## Usage

```
TSA_boxplot(
   tsa_data,
   control_condition = NA,
   color_by = "Protein",
   label_by = "Ligand",
   separate_legend = TRUE
)
```

# **Arguments**

tsa\_data

a data frame that is merged and generated by TSAR::merge\_TSA(). If y = 'RFU', tsa\_data must also be generated by TSAR::normalize\_fluorescence. The Temperature column will be rounded and the average & sd of each rounded temperature is calculated.

 $control\_condition$ 

Either a condition\_ID or NA; NA by default. When a valid Condition ID is provided, a vertical line appears at the average Tm for the specified condition. When NA, this is skipped.

color\_by

character string, either c("Ligand", "Protein"). The condition category to color the boxes within the box plot for comparison. This is represented in the legend. Set to NA to skip.

label\_by

character string, either c("Ligand", "Protein"). The condition category to group the boxes within the box plot. This is represented in the axis. Set to NA to skip.

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```
separate_legend
```

logical; separate\_legend = TRUE by default. When TRUE, the ggplot2 legend is separated from the TSA curve. This is to help with readability. One ggplot is returned when FALSE.

## Value

by default, two ggplots are returned: one TSA curve and one key. When separate\_legend = FALSE one ggplot is returned.

#### See Also

```
merge_TSA for preparing data. See Tm_difference and get_legend for details on function parameters.
```

```
Other TSA Plots: TSA_compare_plot(), TSA_wells_plot(), get_legend(), graph_tsar(), view_deriv()
```

# **Examples**

```
data("example_tsar_data")
TSA_boxplot(example_tsar_data,
    color_by = "Protein",
    label_by = "Ligand", separate_legend = FALSE
)
```

TSA\_compare\_plot

Compare TSA curves to control

# Description

Generate a number of plots based on the input data to compare the average and standard deviation (sd) of each unique condition to a specified control condition. To see all conditions use condition\_IDs(tsa\_data).

## Usage

```
TSA_compare_plot(
   tsa_data,
   control_condition,
   y = "Fluorescence",
   show_Tm = FALSE,
   title_by = "both",
   digits = 1
)
```

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## **Arguments**

| tsa_data       | a data frame that is merged and generated by TSAR::merge_TSA(). If y = 'RFU', tsa_data must also be generated by TSAR::normalize_fluorescence. The Temperature column will be rounded and the average & sd of each rounded |
|----------------|--|
|                | temperature is calculated.   |
| control_condit | ion  |
|                | character string matching a Condition ID. Must be equal to a value within tsa_data\$condition_ID. See unique condition IDs with condition_IDs.   |
| у              | character string; c('Fluorescence', 'RFU'). When y = 'Fluorescence', the original Fluorescence data from TSAR::read_raw_data() is averaged. When y = 'RFU', the average is calculated by the rescaled fluorescence.        |
| show_Tm        | logical; show_Tm = FALSE by default. When TRUE, the Tm is displayed on the plot. When FALSE, the Tm is not added to the plot.  |
| title_by       | character string; c("ligand", "protein", "both"). Automatically names the plots by the specified condition category.   |
| digits         | integer; the number of decimal places to round for change in Tm calculations displayed in the subtitle of each plot  |

## Value

Generates a number of ggplot objects equal to the number of unique Condition IDs present in the input data.

# See Also

merge\_TSA and normalize\_fluorescence for preparing data. See TSA\_average and get\_legend for details on function parameters. See TSA\_wells\_plot for individual curves of the averaged conditions shown.

```
Other TSA Plots: TSA_boxplot(), TSA_wells_plot(), get_legend(), graph_tsar(), view_deriv()
```

# **Examples**

```
data("example_tsar_data")
TSA_compare_plot(example_tsar_data,
    y = "RFU",
    control_condition = "CA FL_DMSO"
)
```

TSA\_ligands TSA Ligands

# **Description**

This function is used to extract information from a data frame of TSA data. The Ligand values should be assigned in the TSA software.

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# Usage

```
TSA\_ligands(tsa\_data, n = FALSE)
```

# **Arguments**

tsa\_data a data frame that is merged and generated by TSAR::merge\_TSA(), or an un-

merged data frame read by TSAR::read\_analysis() or TSAR::read\_raw\_data().

The data frame must have a column named 'Ligand'.

n logical value; n = FALSE by default. When TRUE, a numeric value describing

the number of unique ligand names is returned. When FALSE, a character vector

of unique IDs are returned.

#### Value

Either a character vector of unique well\_IDs or a numeric value.

# See Also

```
merge_TSA, read_raw_data, and read_analysis for preparing input.

Other TSA Summary Functions: TSA_proteins(), condition_IDs(), well_IDs()
```

## **Examples**

```
data("example_tsar_data")
TSA_ligands(example_tsar_data)
```

TSA\_proteins

TSA Proteins

# **Description**

This function is used to extract information from a data frame of TSA data. The Protein values should be assigned in the TSA software.

# Usage

```
TSA_proteins(tsa_data, n = FALSE)
```

# Arguments

| tsa_data | a data frame that is merged and generated by TSAR::merge_TSA(), or an un- |
|----------|---|
|----------|---|

merged data frame read by TSAR::read\_analysis() or TSAR::read\_raw\_data().

The data frame must have a column named 'Protein'.

n logical value; n = FALSE by default. When TRUE, a numeric value describing

the number of unique protein names is returned. When FALSE, a character

vector of unique IDs are returned.

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#### Value

Either a character vector of unique well\_IDs or a numeric value.

#### See Also

```
merge_TSA, read_raw_data, and read_analysis for preparing input.
Other TSA Summary Functions: TSA_ligands(), condition_IDs(), well_IDs()
```

# **Examples**

```
data("example_tsar_data")
TSA_proteins(example_tsar_data)
```

 $\mathsf{TSA\_Tms}$ 

Reformat TSA data into TSA Tms

# **Description**

This function is used to output calculated Tm data from TSA analysis. The input data frame will be transformed into a new format that is helpful for user reading and automated analysis. The Tm values can be listed as a data frame of individual wells or the Tms from identical conditions can be averaged. When condition\_average is TRUE (the default), samples with identical condition IDs will be aggregated and the average / standard deviation will be calculated where appropriate. To analyze multiple TSA experiments, use merge\_TSA() to make a single data frame for analysis.

#### Usage

```
TSA_Tms(analysis_data, condition_average = TRUE)
```

# **Arguments**

analysis\_data

a data frame that is unmerged and generated by TSAR::read\_analysis() or a merged TSA data frame generated by TSAR::merge\_TSA(). Data frames require a column named 'condition\_ID' for averaging.

condition\_average

logical value; n = TRUE by default. When TRUE, the average Tm is calculated by matched condition IDs within the data frame. When FALSE, each well is reported as a unique value with the corresponding Tm.

#### Value

A data frame of Tm values.

TSA\_wells\_plot

## See Also

```
merge_TSA, read_raw_data, and read_analysis for preparing input.

Other TSAR Formatting: TSA_average(), Tm_difference(), merge_TSA(), merge_norm(), normalize_fluorescence()
rescale()
```

# **Examples**

```
data("example_tsar_data")
TSA_Tms(example_tsar_data)
```

TSA\_wells\_plot

TSA Well Curves Plot

# **Description**

Generates the individual curves for each well in the merged tsa data input. Options to create an average and standard deviation sd of the plot in addition to the individual curves. The average and sd will be smoothened by linear regression; see TSA\_average for details.

# Usage

```
TSA_wells_plot(
  tsa_data,
  y = "RFU",
  show_Tm = TRUE,
  Tm_label_nudge = 7.5,
  show_average = TRUE,
  plot_title = NA,
  plot_subtitle = NA,
  smooth = TRUE,
  separate_legend = TRUE)
```

# **Arguments**

| tsa_data | a data frame that is merged and generated by TSAR::merge_TSA(). If y = 'RFU', tsa_data must also be generated by TSAR::normalize_fluorescence. The Temperature column will be rounded and the average & sd of each rounded temperature is calculated. |
|----------|---|
| у        | character string; c('Fluorescence', 'RFU'). When $y = \text{'Fluorescence'}$ , the original Fluorescence data from TSAR::read_raw_data() is averaged. When $y = \text{'RFU'}$ , the average is calculated by the rescaled fluorescence.               |
| show_Tm  | logical; show_Tm = TRUE by default. When TRUE, the Tm is displayed on the plot. When FALSE, the Tm is not added to the plot.  |

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 $Tm\_label\_nudge \quad numeric; Tm\_label\_nudge = 7.5 the direction in the x direction to move the Tm$ 

label. This is used prevent the label from covering data. Ignored if  $show_Tm =$ 

FALSE.

show\_average logical; show\_average = TRUE by default. When TRUE, the average is and sd is

plotted as generated by merge\_TSA.

plot\_title, plot\_subtitle

characer string, NA by default. User-specified plots to overright automatic nam-

ing.

smooth logical; smooth = TRUE by default. When TRUE, linear regression by gam is

used to make clean lines on the plot. See TSA\_average for more details. When

FALSE, individual points are plotted (slows down rendering).

separate\_legend

logical; separate\_legend = TRUE by default. When TRUE, the ggplot2 legend is separated from the TSA curve. This is to help with readability. One ggplot is

returned when FALSE.

#### Value

by default, two ggplots are returned: one TSA curve and one key. When separate\_legend = FALSE one ggplot is returned.

#### See Also

merge\_TSA and normalize\_fluorescence for preparing data. See TSA\_average and get\_legend for details on function parameters.

```
Other TSA Plots: TSA_boxplot(), TSA_compare_plot(), get_legend(), graph_tsar(), view_deriv()
```

#### **Examples**

```
data("example_tsar_data")
check <- subset(example_tsar_data, condition_ID == "CA FL_PyxINE HCl")
TSA_wells_plot(check, separate_legend = FALSE)</pre>
```

view\_deriv

View Derivative Curves

# **Description**

Function reviews data by well and output a graph of the all derivatives wanted. Function called within graph\_tsar function but also runnable outside.

# Usage

```
view_deriv(tsar_data, frame_by = "Well")
```

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# **Arguments**

tsar\_data dataset input, analyzed must have norm\_deriv as a variable; dataset qualifying

norm\_data or tsar\_data both fulfills this parameter, although tsar\_data is more

recommended given more data options.

frame\_by builds plotly by specified frame variable. To graph by a concentration gradient,

well position, or other specified variable, simple specify frame\_by = "condition\_ID".

To view all derivative curves without frames, set to frame\_by = FALSE, else it is

defaulted to frame by well labels.

## Value

plotly object of derivative curves

#### See Also

```
Other TSA Plots: TSA_boxplot(), TSA_compare_plot(), TSA_wells_plot(), get_legend(), graph_tsar()
```

# **Examples**

```
data("example_tsar_data")
view_deriv(example_tsar_data, frame_by = "condition_ID")
```

view\_model

View Model

# Description

Function reviews data by well and output a graph of the fit and a graph of derivative. Function called within analyze\_norm function.

# Usage

```
view_model(raw_data)
```

# **Arguments**

raw\_data

dataset input, not processing needed

#### Value

list of two ggplot graphs

#### See Also

```
Other data_preprocess: model_boltzmann(), model_fit(), model_gam(), normalize(), remove_raw(), run_boltzmann(), screen(), weed_raw()
```

weed\_raw 37

# **Examples**

```
data("qPCR_data1")
test <- subset(qPCR_data1, Well.Position == "A01")
test <- normalize(test)
gammodel <- model_gam(test, x = test$Temperature, y = test$Normalized)
test <- model_fit(test, model = gammodel)
view_model(test)</pre>
```

weed\_raw

Weed raw data for corrupt curves

## **Description**

The weed\_raw function allows users to interact with a screening graph and select curves to weed out before entering analysis. Function wraps together screen and remove\_raw.

#### Usage

```
weed_raw(raw_data, checkrange = NULL, checklist = NULL)
```

# Arguments

raw\_data The raw data for screening.

checkrange list type input identifying range of wells to select. For example, if viewing first

8 wells from row A to C is needed, one can specify the row letters and column

numbers like this: checkrange = c("A", "C", "1", "8")

checklist use this parameter to view selected Wells with full Well names. For example,

checklist = c('A01', 'D11')

# Value

prompts separate app window for user interaction, does not return specific value

#### See Also

```
screen and remove_raw
Other data_preprocess: model_boltzmann(), model_fit(), model_gam(), normalize(), remove_raw(),
run_boltzmann(), screen(), view_model()
```

```
data("qPCR_data1")
if (interactive()) {
   runApp(weed_raw(qPCR_data1, checkrange = c("A", "B", "1", "12")))
}
```

38 well\_IDs

# Description

This function is used to extract information of the well IDs from a merged TSA data frame. Well IDs are automatically generated by the read\_analysis and read\_raw\_data functions in the automated workflow. This function returns either a character vector of unique IDs present or a numeric value of the number of unique IDs.

# Usage

```
well_IDs(tsa_data, n = FALSE)
```

# **Arguments**

| tsa_data | a data frame that is merged and generated by TSAR::merge_TSA(), or an unmerged data frame read by TSAR::read_analysis() or TSAR::read_raw_data(). Data frames require a column named 'well_ID'. |
|----------|---|
| n        | logical value; n = FALSE by default. When TRUE, a numeric value of unique IDs is returned. When FALSE, a character vector of unique IDs are returned.   |

#### Value

Either a character vector of unique well IDs or a numeric value.

# See Also

```
merge_TSA, read_raw_data, and read_analysis for preparing input.
Other TSA Summary Functions: TSA_ligands(), TSA_proteins(), condition_IDs()
```

```
data("example_tsar_data")
well_IDs(example_tsar_data)
```

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well\_information

example well information Data

## **Description**

Dataset Description: This file is a readin using well\_information\_template. File contains the conditions of well, specifying protein and ligand content in well. All experimental setup and relevant information are determined and manually put in by the author of this package.

# Usage

```
data(well_information)
```

## **Format**

A data frame with the following columns:

...1 n/a

**Protein...2** Protein in Well 1

Ligand...3 Ligand in Well 1

**Protein...4** Protein in Well 2

**Ligand...5** Ligand in Well 2

Protein...6 Protein in Well 3

**Ligand...7** Ligand in Well 3

**Protein...8** Protein in Well 4

Ligand...9 Ligand in Well 4

Protein...10 Protein in Well 5

Ligand...11 Ligand in Well 5

**Protein...12** Protein in Well 6

Ligand...13 Ligand in Well 6

Protein...14 Protein in Well 7

Ligand...15 Ligand in Well 7

**Protein...16** Protein in Well 8

Ligand...17 Ligand in Well 8

Protein...18 Protein in Well 9

Ligand...19 Ligand in Well 9

**Protein...20** Protein in Well 10

**Ligand...21** Ligand in Well 10

Protein...22 Protein in Well 11

Ligand...23 Ligand in Well 11

**Protein...24** Protein in Well 12

Ligand...25 Ligand in Well 12

## Value

well information data frame

```
well_information_template
```

Well Information Template

# **Description**

Dataset Description: Template specifies the way condition information will be read in as, specifying protein and ligand content in well.

# Usage

```
data(well_information_template)
```

#### **Format**

A data frame with the following columns:

...1 n/a

**Protein...2** Protein in Well 1

Ligand...3 Ligand in Well 1

**Protein...4** Protein in Well 2

**Ligand...5** Ligand in Well 2

**Protein...6** Protein in Well 3

**Ligand...7** Ligand in Well 3

Protein...8 Protein in Well 4

Ligand...9 Ligand in Well 4

**Protein...10** Protein in Well 5

**Ligand...11** Ligand in Well 5

Protein...12 Protein in Well 6

**Ligand...13** Ligand in Well 6

**Protein...14** Protein in Well 7

Ligand...15 Ligand in Well 7

Protein...16 Protein in Well 8

Ligand...17 Ligand in Well 8

Protein...18 Protein in Well 9

Ligand...19 Ligand in Well 9

Protein...20 Protein in Well 10

Ligand...21 Ligand in Well 10

write\_tsar 41

```
Protein...22 Protein in Well 11Ligand...23 Ligand in Well 11Protein...24 Protein in Well 12Ligand...25 Ligand in Well 12
```

# Value

well information template in data frame

write\_tsar

write output files

# Description

writes output into csv or txt files

# Usage

```
write_tsar(data, name, file = "txt")
```

# Arguments

```
data input data frame

name string, name file to be saved as. Final name will be appended "tsar_output"

file = "txt" writes txt output files; file = "csv" writes csv output files; default set to file = "txt"
```

## Value

file output on the working directory where data was read in

# See Also

```
Other read_write_analysis: join_well_info(), read_tsar()
```

```
data("qPCR_data1")
result <- gam_analysis(qPCR_data1,
     smoothed = TRUE, fluo_col = 5,
     selections = c(
        "Well.Position", "Temperature", "Fluorescence", "Normalized"
    )
)
output_data <- read_tsar(result, output_content = 2)
# example does not run, will build excessive file in package
# write_tsar(output_data, name = "2022_03_18_test", file = "txt")</pre>
```

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