

Package ‘uneqmixr’

December 1, 2022

Type Package

Title Modelling the Quantity of Material Sampled in the Risk Assessment

Version 0.0.1

Authors Mayooraan Thevaraja [aut, cre], Kondaswamy Govindaraju [aut], Mark Bebbington [aut]

URL <https://github.com/Mayooraan1987/uneqmixr>

BugReports <https://github.com/Mayooraan1987/uneqmixr/issues>

Description This package allows practitioners to get probability estimations and graphical displays in the risk assessment. This study mainly focuses on the risk assessment when aggregating unequal incremental samples in the production process.

License GPL (>= 2)

Encoding UTF-8

LazyData true

Imports extraDistr, ggplot2, ggthemes, reshape2, stats

Suggests spelling, testthat

RoxygenNote 7.2.2

Depends R (>= 4.0)

Maintainer Mayooraan Thevaraja <mayooraan@eng.jfn.ac.lk>

Language en-US

R topics documented:

AOQL_scenarios	2
compare_ex_var_scenario_1	3
compare_ex_var_scenario_2	4
compare_ex_var_scenario_3	5
compare_prevalence_scenario_4	6
compare_prevalence_scenario_5	7
Ex_var_scenario_1	8
Ex_var_scenario_2	9
Ex_var_scenario_3	10
scenario_1_OC	11
scenario_1_pa	13
scenario_1_pd	14
scenario_1_pd_curve	15

scenario_2_OC	16
scenario_2_pa	17
scenario_2_pd	18
scenario_2_pd_curve	19
scenario_3_OC	20
scenario_3_pa	21
scenario_3_pd	22
scenario_3_pd_curve	23
scenario_4_prevalence	24
scenario_5_prevalence	25
uneqmixr	26

Index	27
--------------	-----------

AOQL_scenarios	<i>Construction of AOQ curve and calculate AOQL value based on average microbial counts</i>
----------------	---

Description

[AOQL_scenarios](#) provides the Average Outgoing Quality (AOQ) curve and calculates Average Outgoing Quality Level (AOQL) value based on expected microbial counts in each scenario.

Usage

```
AOQL_scenarios(c, llim, sd, m1, m2, scenario, n, type, K, n_sim)
```

Arguments

c	acceptance number
llim	the upper limit for graphing the arithmetic mean of cell count
sd	standard deviation on the log10 scale (default value 0.8).
m1	the vector of the first set of incremental samples (with equal/unequal weights).
m2	the vector of the second set of incremental samples (with equal/unequal weights).
scenario	what scenario we have considered such as "1" or "2" or "3"
n	number of aggregate samples which are used for inspection.
type	what type of the results you would like to consider such as "theory" or "simulation".
K	dispersion parameter of the Poisson gamma distribution (default value 0.25)
n_sim	number of simulations (large simulations provide more precise estimation).

Details

Since p_a is the probability of acceptance, λ is the arithmetic mean of cell count and the outgoing contaminated arithmetic mean of cell count of incremental samples is given by AOQ as the product λp_a . The quantity $AOQL$ is defined as the maximum proportion of outgoing contaminated incremental samples and is given by

$$AOQL = \max_{\lambda \geq 0} \lambda p_a$$

Value

AOQ curve and AOQL value based on expected microbial counts in each scenario.

See Also

[scenario_1_pa](#), [scenario_2_pa](#), [scenario_3_pa](#)

Examples

```
c <- 0
llim <- 0.02
sd <- 0.8
m1 <- c(10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10)
m2 <- c(15,5,5,5,10,5,10,5,15,10,5,10,5,25,10,5,10,5,5,10,5,15,10,
5,5,20,5,10,5,10,20,5,10,30,5,20,5,10,5,10,20,15,10,15,10,
10,5,10,15,5)
n <- 10
AOQL_scenarios(c,llim, sd, m1, m2, scenario = "1", n, type = "theory")
```

compare_ex_var_scenario_1	<i>Graphical displays of expectation or variance of different sampling schemes under scenario 1.</i>
---------------------------	--

Description

[compare_ex_var_scenario_1](#) provides graphical displays based on expectation or variance under scenario 1 of modelling the quantity of material sampled in the risk assessment study.

Usage

```
compare_ex_var_scenario_1(mulow, muhigh, sd, m1, m2, measure)
```

Arguments

mulow	the lower value of the mean concentration (μ) for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration (μ) for use in the graphical display's x-axis.
sd	standard deviation on the log10 scale (default value 0.8).
m1	the vector of the first set of incremental samples (with equal/unequal weights).
m2	the vector of the second set of incremental samples (with equal/unequal weights).
measure	what type of measure you would like to consider for the graph, such as "expectation" or "variance".

Details

[compare_ex_var_scenario_1](#) provides graphical displays based on expectation or variance under scenario 1 of modelling the quantity of material sampled in the risk assessment study. Under this scenario (a lot with homogeneous contaminations), we employed Poisson distribution to the model number of micro-organisms in the incremental samples. (this section will be updated later on)

Value

Graphical displays based on expectation or variance under lot with homogeneous contaminations.

Examples

```
mulow <- -1
muhigh <- 1
sd <- 0.8
m1 <- c(10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10)
m2 <- c(15,5,5,5,10,5,10,5,15,10,5,10,5,25,10,5,10,5,5,10,5,15,10,
5,5,20,5,10,5,10,20,5,10,30,5,20,5,10,5,10,20,15,10,15,10,
10,5,10,15,5)
compare_ex_var_scenario_1(mulow, muhigh, sd = 0.8, m1, m2, measure = "variance")
compare_ex_var_scenario_1(mulow, muhigh, sd = 0.8, m1, m2, measure = "expectation")
```

compare_ex_var_scenario_2

Graphical displays of expectation or variance of different sampling schemes under scenario 2.

Description

[compare_ex_var_scenario_2](#) provides graphical displays based on expectation or variance under scenario 2 of modelling the quantity of material sampled in the risk assessment study.

Usage

```
compare_ex_var_scenario_2(mulow, muhigh, sd, m1, m2, measure)
```

Arguments

mulow	the lower value of the mean concentration (μ) for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration (μ) for use in the graphical display's x-axis.
sd	standard deviation on the log10 scale (default value 0.8).
m1	the vector of the first set of incremental samples (with equal/unequal weights).
m2	the vector of the second set of incremental samples (with equal/unequal weights).
measure	what type of measure you would like to consider for the graph, such as "expectation" or "variance".

Details

[compare_ex_var_scenario_2](#) provides graphical displays based on expectation or variance under scenario 2 of modelling the quantity of material sampled in the risk assessment study. Under this scenario (a lot with homogeneous contaminations), we employed Poisson lognormal distribution to the model number of micro-organisms in the incremental samples. (this section will be updated later on)

Value

Graphical displays based on expectation or variance when lot with heterogeneous and high-level contamination.

Examples

```
mulow <- -2
muhigh <- 2
m1 <- c(10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10)
m2 <- c(15,5,5,5,10,5,10,5,15,10,5,10,5,25,10,5,10,5,5,10,5,15,10,
5,5,20,5,10,5,10,20,5,10,30,5,20,5,10,5,10,20,15,10,15,10,
10,5,10,15,5)
compare_ex_var_scenario_2(mulow, muhigh, sd = 0.8, m1, m2, measure = "variance")
compare_ex_var_scenario_2(mulow, muhigh, sd = 0.8, m1, m2, measure = "expectation")
```

```
compare_ex_var_scenario_3
```

Graphical displays of expectation or variance of different sampling schemes under scenario 3.

Description

[compare_ex_var_scenario_3](#) provides graphical displays based on expectation or variance under scenario 3 of modelling the quantity of material sampled in the risk assessment study.

Usage

```
compare_ex_var_scenario_3(mulow, muhigh, sd, m1, m2, K, measure)
```

Arguments

mulow	the lower value of the mean concentration (μ) for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration (μ) for use in the graphical display's x-axis.
sd	standard deviation on the log10 scale (default value 0.8).
m1	the vector of the first set of incremental samples (with equal/unequal weights).
m2	the vector of the second set of incremental samples (with equal/unequal weights).
K	shape parameter (default value 0.25).
measure	what type of measure you would like to consider for the graph, such as "expectation" or "variance".

`compare_ex_var_scenario_3` provides a probability of acceptance under scenario 3 of modelling the quantity of material sampled in the risk assessment study. Under this scenario (a lot with homogeneous contaminations), we employed Poisson gamma distribution to the model number of micro-organisms in the incremental samples. Based on the food safety literature, the expected cell count is given by $\lambda = 10^{\mu + \log(10)\sigma^2/2}$. (this section will be updated later on)

Graphical displays based on expectation or variance when lot with heterogeneous and low-level contamination.

```
mulow <- 0
muhigh <- 2
m1 <- c(10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10)
m2 <- c(15,5,5,5,10,5,10,5,15,10,5,10,5,25,10,5,10,5,5,10,5,15,10,
5,5,20,5,10,5,10,20,5,10,30,5,20,5,10,5,10,20,15,10,15,10,
10,5,10,15,5)
K <- 0.05
compare_ex_var_scenario_3(mulow, muhigh, sd = 0.8, m1, m2, K, measure = "variance")
compare_ex_var_scenario_3(mulow, muhigh, sd = 0.8, m1, m2, K, measure = "expectation")
```

Graphical displays of prevalence before inspection of different sampling schemes under scenario 4.

`compare_prevalence_scenario_4` provides graphical displays based on prevalence before inspection under scenario 4 of modelling the quantity of material sampled in the risk assessment study.

```
compare_prevalence_scenario_4(mulow, muhigh, sd, m1, m2, l, type, n_sim)
```

mulow	the lower value of the mean concentration (μ) for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration (μ) for use in the graphical display's x-axis.
sd	standard deviation on the log10 scale (default value 0.8).
m1	the vector of the first set of incremental samples (with equal/unequal weights).
m2	the vector of the second set of incremental samples (with equal/unequal weights).
l	the number of lots in the production process.

type	what type of the results you would like to consider such as "theory" or "simulation".
n_sim	number of simulations (large simulations provide more precise estimation).

Details

[compare_prevalence_scenario_4](#) provides graphical displays based on prevalence before inspection under scenario 4 of modelling the quantity of material sampled in the risk assessment study. (this section will be updated later on)

Value

Graphical displays based on prevalence before inspection when lot with homogeneous contamination and contamination levels fluctuate from lot to lot by using theoretical or simulation-based results.

Examples

```
mulow <- -6
muhigh <- 1
m1 <- c(10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10)
m2 <- c(15,5,5,5,10,5,10,5,15,10,5,10,5,25,10,5,10,5,5,10,5,15,10,
5,5,20,5,10,5,10,20,5,10,30,5,20,5,10,5,10,20,15,10,15,10,
10,5,10,15,5)
l <- 5000
compare_prevalence_scenario_4(mulow, muhigh, sd = 0.8, m1, m2, l, type = "theory")
```

compare_prevalence_scenario_5

Graphical displays of prevalence before inspection of different sampling schemes under scenario 5.

Description

[compare_prevalence_scenario_5](#) provides graphical displays based on prevalence before inspection under scenario 5 of modelling the quantity of material sampled in the risk assessment study.

Usage

```
compare_prevalence_scenario_5(mulow, muhigh, sd, m1, m2, l, type, n_sim)
```

Arguments

mulow	the lower value of the mean concentration (μ) for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration (μ) for use in the graphical display's x-axis.
sd	standard deviation on the log10 scale (default value 0.8).
m1	the vector of the first set of incremental samples (with equal/unequal weights).

Description

Usage

Arguments

Details

Value

Examples

```
mu <- -3  
sd <- 0.8  
m <- c(5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5)  
scenario_1_pd(mu, sd = "0.8", m, type = "theory")  
scenario_1_pd(mu, sd = 0.8, m, type = "simulation", n_sim = 1000000)
```

scenario_2_OC	<i>Construction of Operating Characteristic (OC) curve under lot with heterogeneous and high-level contamination.</i>
---------------	---

Description

[scenario_2_OC](#) provides the Operating Characteristic (OC) curves under scenario 2 of modelling the quantity of material sampled in the risk assessment study.

Usage

```
scenario_2_OC(c, mulow, muhigh, sd, m1, m2, n, type, n_sim)
```

Arguments

c	acceptance number
mulow	the lower value of the mean concentration (μ) for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration (μ) for use in the graphical display's x-axis.
sd	standard deviation on the log10 scale (default value 0.8).
m1	the vector of the first set of incremental samples (with equal/unequal weights).
m2	the vector of the second set of incremental samples (with equal/unequal weights).
n	number of aggregate samples which are used for inspection.
type	what type of the results you would like to consider such as "theory" or "simulation".
n_sim	number of simulations (large simulations provide more precise estimation).

Details

[scenario_2_OC](#) provides the Operating Characteristic (OC) curves under scenario 2 of modelling the quantity of material sampled in the risk assessment study. The purpose of this function used for compares two different sets of sampling schemes when lot with heterogeneous and high-level contamination. Nevertheless, each sampling scheme's total quantity (weight of aggregate sample (say M)) must be equal. The probability of acceptance is plotted against mean log10 concentration and expected cell counts. We employed Poisson lognormal distribution to the model number of micro-organisms in the incremental samples. Based on the food safety literature, the expected cell count is given by $\lambda = 10^{\mu + \log(10)\sigma^2/2}$. (this section will be updated later on)

Value

Operating Characteristic (OC) curves when lot with heterogeneous and high-level contamination.

See Also

[scenario_2_pd](#)

Examples

```
c <- 0
mulow <- -10
muhigh <- 0
sd <- 0.8
m1 <- c(10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,10,
10,10,10,10,10,10,10,10,10,10)
m2 <- c(15,5,5,5,10,5,10,5,15,10,5,10,5,25,10,5,10,5,5,10,5,15,10,
5,5,20,5,10,5,10,20,5,10,30,5,20,5,10,5,10,20,15,10,15,10,
10,5,10,15,5)
n <- 10
scenario_2_OC(c, mulow, muhigh, sd = 0.8, m1, m2, n, type = "theory")
```

scenario_2_pa	<i>Probability of acceptance estimation when lot with heterogeneous and high-level contamination.</i>
---------------	---

Description

[scenario_2_pa](#) provides a probability of acceptance under scenario 2 of modelling the quantity of material sampled in the risk assessment study.

Usage

```
scenario_2_pa(c, mu, sd, m, n, type, n_sim)
```

Arguments

c	acceptance number
mu	the the mean concentration (μ).
sd	standard deviation on the log10 scale (default value 0.8).
m	the vector of incremental samples (with equal/unequal weights).
n	number of aggregate samples which are used for inspection.
type	what type of the results you would like to consider such as "theory" or "simulation".But,not yet established for this scenario at this time.
n_sim	number of simulations (large simulations provide more precise estimation).

Details

[scenario_2_pa](#) provides a probability of acceptance under scenario 2 of modelling the quantity of material sampled in the risk assessment study. Under this scenario (a lot with heterogeneous, high-level contamination), we employed Poisson lognormal distribution to the model number of micro-organisms in the incremental samples. Based on the food safety literature, the expected cell count is given by $\lambda = 10^{\mu + \log(10)\sigma^2/2}$. (this section will be updated later on)

Value

Probability of acceptance when lot with heterogeneous and high-level contamination.

```
c <- 0  
mu <- -6  
sd <- 0.8  
m <- c(5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5)  
n <- 10  
scenario_2_pa(c, mu, sd, m, n, type = "theory")  
scenario_2_pa(c, mu, sd, m, n, type = "simulation", n_sim = 200000)
```

scenario_2_pd	<i>Probability of detection estimation when lot with heterogeneous and high-level contamination.</i>
---------------	--

`scenario_2_pd` provides a probability of detection under scenario 2 of modelling the quantity of material sampled in the risk assessment study.

```
scenario_2_pd(mu, sd, m, type, n_sim)
```

mu	the the mean concentration (μ).
sd	standard deviation on the log10 scale (default value 0.8).
m	the vector of incremental samples (with equal/unequal weights).
type	what type of the results you would like to consider such as "theory" or "simulation".
n_sim	number of simulations (large simulations provide more precise estimation).

`scenario_2_pd` provides a probability of detection under scenario 2 of modelling the quantity of material sampled in the risk assessment study. Under this scenario (a lot with heterogeneous, high-level contamination), we employed Poisson lognormal distribution to the model number of micro-organisms in the incremental samples. Based on the food safety literature, the expected cell count is given by $\lambda = 10^{\mu + \log(10)\sigma^2/2}$. (this section will be updated later on)

Probability of detection when lot with heterogeneous and high-level contamination by using theoretical or simulation-based results.

Examples

```
mu <- -3  
sd <- 0.8  
m <- c(5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5)  
scenario_2_pd(mu, sd = 0.8, m, type = "theory")  
scenario_2_pd(mu, sd = 0.8, m, type = "simulation", n_sim = 2000000)
```

scenario_2_pd_curve	<i>Construction of Operating Characteristic (OC) curve under lot with heterogeneous and high-level contamination.</i>
---------------------	---

Description

scenario_2_pd_curve provides the probability of detection curves under scenario 2 of modelling the quantity of material sampled in the risk assessment study.

Usage

```
scenario_2_pd_curve(mulow, muhigh, sd, m1, m2, type, n_sim)
```

Arguments

mulow	the lower value of the mean concentration (μ) for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration (μ) for use in the graphical display's x-axis.
sd	standard deviation on the log10 scale (default value 0.8).
m1	the vector of the first set of incremental samples (with equal/unequal weights).
m2	the vector of the second set of incremental samples (with equal/unequal weights).
type	what type of the results you would like to consider such as "theory" or "simulation".
n_sim	number of simulations (large simulations provide more precise estimation).

Details

scenario_2_pd_curve provides the probability of detection curves under scenario 2 of modelling the quantity of material sampled in the risk assessment study. (this section will be updated later on)

Value

probability of detection curves when lot with heterogeneous and high-level contamination.

See Also

scenario_2_pd

`scenario_3_pa` provides a probability of acceptance under scenario 3 of modelling the quantity of material sampled in the risk assessment study. Under this scenario (a lot with heterogeneous, low-level contamination), we employed Poisson gamma distribution to the model number of micro-organisms in the incremental samples. Based on the food safety literature, the expected cell count is given by $\lambda = 10^{\mu + \log(10)\sigma^2/2}$. (this section will be updated later on)

Probability of acceptance when lot with heterogeneous and low-level contamination.

```
c <- 0  
mu <- -3  
sd <- 0.8  
m <- c(5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5)  
K <- 0.25  
n <- 10  
scenario_3_pa(c, mu, sd = 0.8, m, K, n, type = "theory")  
scenario_3_pa(c, mu, sd = 0.8, m, K, n, type = "simulation", n_sim = 1000000)
```

scenario_3_pd	<i>Probability of detection estimation when lot with heterogeneous and low-level contamination.</i>
---------------	---

`scenario_3_pd` provides a probability of detection under scenario 3 of modelling the quantity of material sampled in the risk assessment study.

```
scenario_3_pd(mu, sd, m, K, type, n_sim)
```

mu	the the mean concentration (μ).
sd	standard deviation on the log10 scale (default value 0.8).
m	the vector of incremental samples(with equal/unequal weights).
K	shape parameter (default value 0.25).
type	what type of the results you would like to consider such as "theory" or "simulation".
n_sim	number of simulations (large simulations provide more precise estimation).


```
mu <- -3  
sd <- 0.8  
m <- c(5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,  
5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5,5)  
l <- 5000  
scenario_5_prevalence(mu, sd, m, l, type = "theory")
```

uneqmixr

Probability estimations and graphical displays in modelling the quantity of material sampled in the risk assessment.

This package aims to develop for getting probability estimations and graphical displays in the study associated with modelling the quantity of material sampled in the risk assessment.

This package aims to develop probability estimations and graphical displays in modelling the quantity of material sampled in the risk assessment. This study mainly focuses on the risk assessment when aggregating unequal incremental samples in the production process. It mainly focuses on the risk assessment based on compound Poisson mixture distributions to model in five different scenarios.

1. Scenario 1—lots with homogeneous contamination;
2. Scenario 2—lots with heterogeneous, high-level contamination;
3. Scenario 3—lots with heterogeneous, low-level contamination;
4. Scenario 4—lots with homogeneous contamination and concentration levels fluctuating from sub lots; and
5. Scenario 5—lots with heterogeneous contamination and concentration levels fluctuating from sub-lots.

This package allows practitioners to get probability estimations and graphical displays based on this study. Also, this package can be used to validate the derived results in this study by simulation.

Index

AOQL_scenarios, [2](#), [2](#)

compare_ex_var_scenario_1, [3](#), [3](#), [4](#)
compare_ex_var_scenario_2, [4](#), [4](#), [5](#)
compare_ex_var_scenario_3, [5](#), [5](#), [6](#)
compare_prevalence_scenario_4, [6](#), [6](#), [7](#)
compare_prevalence_scenario_5, [7](#), [7](#), [8](#)

Ex_var_scenario_1, [8](#), [8](#), [9](#)
Ex_var_scenario_2, [9](#), [9](#), [10](#)
Ex_var_scenario_3, [10](#), [10](#), [11](#)

scenario_1_OC, [11](#), [11](#), [12](#)
scenario_1_pa, [3](#), [13](#), [13](#)
scenario_1_pd, [12](#), [14](#), [14](#), [15](#), [24](#)
scenario_1_pd_curve, [15](#), [15](#)
scenario_2_OC, [16](#), [16](#)
scenario_2_pa, [3](#), [17](#), [17](#)
scenario_2_pd, [16](#), [18](#), [18](#), [19](#)
scenario_2_pd_curve, [19](#), [19](#)
scenario_3_OC, [20](#), [20](#)
scenario_3_pa, [3](#), [21](#), [21](#), [22](#)
scenario_3_pd, [21](#), [22](#), [22](#), [23](#), [24](#)
scenario_3_pd_curve, [23](#), [23](#), [24](#)
scenario_4_prevalence, [24](#), [24](#)
scenario_5_prevalence, [25](#), [25](#)

uneqmixr, [26](#)