# Package 'uneqmixr'

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AOQL\_scenarios

AOQL.	scenarios Construction of AOQ curve and calculate AOQL value based on average microbial counts
Index	26
	scenario_5_prevalence
	scenario_4_prevalence
	scenario_3_pd_curve
	$scenario\_3\_pd \dots \dots$
	scenario_3_pa
	scenario_3_OC
	scenario_2_pd_curve
	scenario_2_pd
	scenario_2_pa
	scenario_2_OC

### 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, m, scenario, n, type, K, n_sim)
```

### **Arguments**

С	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).
m	the vector 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,  $\lambda$  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  $\lambda p_a$ . The quantity AOQL is defined as the maximum proportion of outgoing contaminated incremental samples and is given by

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

### 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**

```
compare_ex_var_scenario_1
```

Graphical displays of expectation or variance of different sampling schemes under scenario 1.

### **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 $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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)

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

#### **Examples**

```
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 $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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)

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

### **Examples**

```
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 $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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".

### **Details**

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.

### **Examples**

compare\_prevalence\_scenario\_4

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

#### **Description**

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.

#### Usage

```
compare_prevalence_scenario_4(mulow, muhigh, sd, m1, m2, 1, type, n_sim)
```

#### **Arguments**

mulow	the lower value of the mean concentration $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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).
1	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)

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**

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 $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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).
1	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\_5 provides graphical displays based on prevalence before inspection under scenario 5 of modelling the quantity of material sampled in the risk assessment study. (this section will be updated later on)

8 Ex\_var\_scenario\_1

#### Value

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

### **Examples**

Ex\_var\_scenario\_1

Expected value or variance estimation of microorganism in the aggregate sample under scenario 1.

### **Description**

Ex\_var\_scenario\_1 provides the expected value or variance of the number of microorganisms in the aggregate sample under scenario 1 of modelling the quantity of material sampled in the risk assessment study.

### Usage

```
Ex_var_scenario_1(mu, sd = 0.8, m, type, measure, n_sim)
```

#### **Arguments**

mu	the the mean concentration $(\mu)$ .
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".
measure	what type of measure you would like to consider for the graph, such as "expectation" or "variance".
n_sim	number of simulations (large simulations provide more precise estimation).

### **Details**

Ex\_var\_scenario\_1 provides the expected value or variance of the number of microorganisms in the aggregate sample under scenario 1 of modelling the quantity of material sampled in the risk assessment study. (this section will be updated later on)

Ex\_var\_scenario\_2

#### Value

expected value or variance of the number of microorganisms in the aggregate sample.

### **Examples**

Ex\_var\_scenario\_2

Expected value or variance estimation of microorganism in the aggregate sample under scenario 2.

#### **Description**

Ex\_var\_scenario\_2 provides the expected value or variance of the number of microorganisms in the aggregate sample under scenario 2 of modelling the quantity of material sampled in the risk assessment study.

### Usage

```
Ex_var_scenario_2(mu, sd = 0.8, m, type, measure, n_sim)
```

#### **Arguments**

mu	the the mean concentration $(\mu)$ .
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".
measure	what type of measure you would like to consider for the graph, such as "expectation" or "variance".
n_sim	number of simulations (large simulations provide more precise estimation).

#### **Details**

Ex\_var\_scenario\_2 provides the expected value or variance of the number of microorganisms in the aggregate sample under scenario 2 of modelling the quantity of material sampled in the risk assessment study. (this section will be updated later on)

### Value

expected value or variance of the number of microorganisms in the aggregate sample.

10 Ex\_var\_scenario\_3

#### **Examples**

Ex\_var\_scenario\_3

Expected value or variance estimation of microorganism in the aggregate sample under scenario 3.

### **Description**

Ex\_var\_scenario\_3 provides the expected value or variance of the number of microorganisms in the aggregate sample under scenario 3 of modelling the quantity of material sampled in the risk assessment study.

#### Usage

```
Ex_var_scenario_3 (mu, sd, m, K, type, measure, n_sim)
```

#### **Arguments**

mu	the the mean concentration $(\mu)$ .
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".
measure	what type of measure you would like to consider for the graph, such as "expectation" or "variance".
n_sim	number of simulations (large simulations provide more precise estimation).

#### **Details**

Ex\_var\_scenario\_3 provides the expected value or variance of the number of microorganisms in the aggregate sample under scenario 3 of modelling the quantity of material sampled in the risk assessment study. (this section will be updated later on)

### Value

expected value or variance of the number of microorganisms in the aggregate sample.

scenario\_1\_OC

#### **Examples**

scenario\_1\_0C

Construction of Operating Characteristic (OC) curve under lot with homogeneous contaminations based on simulations results.

### **Description**

scenario\_1\_0C provides the Operating Characteristic (OC) curves under scenario 1 of modelling the quantity of material sampled in the risk assessment study.

### Usage

```
scenario_1_OC(c, mulow, muhigh, sd, m1, m2, n, type, n_sim)
```

### **Arguments**

С	acceptance number
mulow	the lower value of the mean concentration $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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\_1\_0C provides the Operating Characteristic (OC) curves under scenario 1 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 homogeneous contaminations. 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 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)

12 scenario\_1\_pa

#### Value

Operating Characteristic (OC) curves when lot with homogeneous contaminations.

#### See Also

```
scenario_1_pd
```

### **Examples**

scenario\_1\_pa

Probability of acceptance estimation when lot with homogeneous contaminations.

#### **Description**

scenario\_1\_pa provides a probability of acceptance under scenario 1 of modelling the quantity of material sampled in the risk assessment study.

### Usage

```
scenario_1_pa(c, mu, sd, m, n, type, n_sim)
```

#### **Arguments**

С	acceptance number
mu	the the mean concentration $(\mu)$ .
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".
n_sim	number of simulations (large simulations provide more precise estimation).

#### **Details**

scenario\_1\_pa provides a probability of acceptance 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. 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)

scenario\_1\_pd 13

#### Value

Probability of acceptance under lot with homogeneous contaminations.

#### **Examples**

scenario\_1\_pd

Probability of detection estimation when lot with homogeneous contaminations.

### Description

scenario\_1\_pd provides a probability of detection under scenario 1 of modelling the quantity of material sampled in the risk assessment study.

#### Usage

```
scenario_1_pd(mu, sd, m, type, n_sim)
```

### **Arguments**

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

### **Details**

scenario\_1\_pd provides a probability of detection under scenario 1 of modelling the quantity of material sampled in the risk assessment study. Under this scenario (lot with homogeneous contaminations), we employed Poisson 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 detection when lot with homogeneous contaminants by using theoretical or simulation-based results.

### **Examples**

scenario\_1\_pd\_curve

Construction of Operating Characteristic (OC) curve under lot with homogeneous contaminations based on simulations results.

### **Description**

scenario\_1\_pd\_curve provides the probability of detection curves under scenario 1 of modelling the quantity of material sampled in the risk assessment study.

### Usage

```
scenario_1_pd_curve(mulow, muhigh, sd, m1, m2, type, n_sim)
```

#### **Arguments**

mulow	the lower value of the mean concentration $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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\_1\_pd\_curve provides the probability of detection curves under scenario 1 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 homogeneous contaminations.

### See Also

```
scenario_1_pd
```

scenario\_2\_OC

#### **Examples**

scenario\_2\_OC

Construction of Operating Characteristic (OC) curve under lot with heterogeneous and high-level contamination.

#### **Description**

scenario\_2\_0C 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

С	acceptance number
mulow	the lower value of the mean concentration $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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\_0C 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)

scenario\_2\_pa

#### Value

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

#### See Also

```
scenario_2_pd
```

### **Examples**

scenario\_2\_pa

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

### 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)
```

С	acceptance number
mu	the the mean concentration $(\mu)$ .
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).

scenario\_2\_pd 17

#### **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.

### **Examples**

scenario\_2\_pd

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

#### **Description**

scenario\_2\_pd provides a probability of detection under scenario 2 of modelling the quantity of material sampled in the risk assessment study.

### Usage

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

### **Arguments**

mu the the mean concentration  $(\mu)$ .

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).

#### **Details**

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**

scenario\_2\_pd\_curve

Construction of Operating Characteristic (OC) curve under lot with heterogeneous and high-level contamination.

### **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 $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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.

scenario\_3\_OC

#### See Also

```
scenario_2_pd
```

### **Examples**

scenario\_3\_OC

Construction of Operating Characteristic (OC) curve under lot with heterogeneous and low-level contamination.

### **Description**

scenario\_3\_0C provides the Operating Characteristic (OC) curves under scenario 3 of modelling the quantity of material sampled in the risk assessment study.

### Usage

```
scenario_3_OC(c, mulow, muhigh, sd, m1, m2, K, n, type, n_sim)
```

С	acceptance number
mulow	the lower value of the mean concentration $(\mu)$ for use in the graphical display's x-axis.
muhigh	the upper value of the mean concentration $(\mu)$ 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).
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).

20 scenario\_3\_pa

#### **Details**

scenario\_3\_0C provides the Operating Characteristic (OC) curves under scenario 3 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 low-level contamination. Nevertheless, each sampling scheme's total quantity (weight of aggregate sample (say M)) must be equal. We employed Poisson gamma distribution to the model number of microorganisms 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 low-level contamination.

#### See Also

```
scenario_3_pd
```

#### **Examples**

scenario\_3\_pa

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

#### **Description**

scenario\_3\_pa provides a probability of acceptance under scenario 3 of modelling the quantity of material sampled in the risk assessment study.

#### Usage

```
scenario_3_pa(c, mu, sd, m, K, n, type, n_sim)
```

С	acceptance number
mu	the the mean concentration $(\mu)$ .
sd	standard deviation on the log10 scale (default value 0.8).
m	the vector of incremental samples (with equal/unequal weights).

scenario\_3\_pd 21

K	shape parameter (default value 0.25).
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\_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 microorganisms 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 low-level contamination.

### **Examples**

scenario\_3\_pd

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

### **Description**

scenario\_3\_pd provides a probability of detection under scenario 3 of modelling the quantity of material sampled in the risk assessment study.

#### Usage

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

mu	the the mean concentration $(\mu)$ .
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).

22 scenario\_3\_pd\_curve

#### **Details**

scenario\_3\_pd provides a probability of detection 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 microorganisms 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 detection when lot with heterogeneous and low-level contamination by using theoretical or simulations based results.

#### **Examples**

scenario\_3\_pd\_curve

Construction of Operating Characteristic (OC) curve under lot with heterogeneous and low-level contamination.

### **Description**

scenario\_3\_pd\_curve provides the probability of detection curves under scenario 3 of modelling the quantity of material sampled in the risk assessment study.

#### Usage

```
scenario_3_pd_curve(mulow, muhigh, sd, m1, m2, K, type, n_sim)
```

mu	low	the lower value of the mean concentration $(\mu)$ for use in the graphical display's x-axis.
mul	high	the upper value of the mean concentration $(\mu)$ 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).
ty	pe	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\_4\_prevalence 23

#### **Details**

scenario\_3\_pd\_curve provides the probability of detection curves under scenario 3 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 low-level contamination.

#### See Also

```
scenario_3_pd
```

### **Examples**

scenario\_4\_prevalence Prevalence estimation before inspection when lot with homogeneous contamination and contamination levels fluctuate from lot to lot.

### **Description**

scenario\_4\_prevalence provides a prevalence before inspection under scenario 4 of modelling the quantity of material sampled in the risk assessment study.

### Usage

```
scenario\_4\_prevalence(mu, sd, m, l, type, n\_sim)
```

### **Arguments**

mu	the the mean concentration $(\mu)$ .
sd	standard deviation on the log10 scale (default value 0.8).
m	the vector of incremental samples (with equal/unequal weights).
1	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**

scenario\_1\_pd provides a 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)

Prevalence estimation before inspection when lot with homogeneous contamination and contamination levels fluctuate from lot to lot by using theoretical or simulation-based results.

#### **Examples**

scenario\_5\_prevalence Prevalence estimation before inspection when lot with heterogeneous contamination and contamination levels fluctuate from lot to lot.

### **Description**

scenario\_5\_prevalence provides a prevalence before inspection under scenario 5 of modelling the quantity of material sampled in the risk assessment study.

#### Usage

```
scenario_5_prevalence(mu, sd, m, l, type, n_sim)
```

#### **Arguments**

mu	the the mean concentration $(\mu)$ .
sd	standard deviation on the log10 scale (default value 0.8).
m	the vector of incremental samples (with equal/unequal weights).
1	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**

scenario\_5\_prevalence provides a prevalence before inspection under scenario 5 of modelling the quantity of material sampled in the risk assessment study. (this section will be updated later on)

### Value

Prevalence estimation before inspection when lot with heterogeneous contamination and contamination levels fluctuate from lot to lot by using theoretical or simulation-based results.

scenario\_5\_prevalence 25

### **Examples**

## **Index**

```
AOQL_scenarios, 2, 2
compare_ex_var_scenario_1, 3, 3
compare_ex_var_scenario_2, 4, 4
compare_ex_var_scenario_3, 5, 5
compare_prevalence_scenario_4, 6, 6
compare_prevalence_scenario_5, 7, 7
Ex_var_scenario_1, 8, 8
Ex_var_scenario_2, 9, 9
Ex_var_scenario_3, 10, 10
scenario_1_0C, 11, 11
scenario_1_pa, 3, 12, 12
scenario_1_pd, 12, 13, 13, 14, 23
scenario_1_pd_curve, 14, 14
scenario_2_OC, 15, 15
scenario_2_pa, 3, 16, 16, 17
scenario_2_pd, 16, 17, 17, 19
scenario_2_pd_curve, 18, 18
scenario_3_OC, 19, 19, 20
scenario_3_pa, 3, 20, 20, 21
scenario_3_pd, 20, 21, 21, 22, 23
scenario_3_pd_curve, 22, 22, 23
scenario_4_prevalence, 23, 23
scenario_5_prevalence, 24, 24
```