

# Package ‘jfa’

October 30, 2020

**Title** Bayesian and Classical Audit Sampling

**Version** 0.4.0

**Description** Implements the audit sampling workflow as discussed in Derks et al. (2019) <doi:10.31234/osf.io/9f6ub>. The package makes it easy for an auditor to plan an audit sample, sample from the population, and evaluating that sample using various confidence bounds according to the International Standards on Auditing. Furthermore, the package implements Bayesian equivalents of these methods.

**Language** en-US

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 7.1.1

**VignetteBuilder** knitr

**Suggests** testthat, knitr, rmarkdown

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auditPrior	<i>Create a Prior Distribution with Existing Audit Information</i>
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## Description

This function creates a prior distribution for Bayesian audit sampling according to several methods discussed in Derks et al. (2020). The returned object is of class `jfaPrior` and can be used with associated `print()` and `plot()` methods. `jfaPrior` objects can be used as input for the prior argument in other functions.

## Usage

```
auditPrior(confidence = 0.95, likelihood = "binomial", method = "none",
           expectedError = 0, N = NULL, materiality = NULL, ir = 1, cr = 1,
           pHmin = NULL, pHplus = NULL, factor = 1, sampleN = 0, sampleK = 0)
```

## Arguments

confidence	the confidence level desired from the confidence bound (on a scale from 0 to 1). Defaults to 0.95, or 95% confidence.
likelihood	can be one of binomial, poisson, or hypergeometric. See the Details section for more information.
method	the method by which the prior distribution is constructed. Defaults to the arm method, which uses the audit risk model (Derks et al., 2019). Can be one of none, median, hypotheses, arm, sample or factor. See the Details section for more information.
expectedError	a fraction representing the percentage of expected mistakes in the sample relative to the total size, or a number ( $\geq 1$ ) that represents the number of expected mistakes.
N	the population size (only required when likelihood = 'hypergeometric').
materiality	a value between 0 and 1 representing the materiality of the audit as a fraction of the total size or value. Can be NULL for some methods.
ir	the inherent risk probability from the audit risk model. Defaults to 1 for 100% risk.
cr	the inherent risk probability from the audit risk model. Defaults to 1 for 100% risk.
pHmin	When using method = 'hypotheses', the prior probability of the hypothesis $\theta < \text{materiality}$ .
pHplus	When using method = 'hypotheses', the prior probability of the hypothesis $\theta > \text{materiality}$ .
factor	When using method = 'factor', the value of the weighting factor for the results of the previous sample.
sampleN	When using method sample or factor, the number of transactions that were inspected in the previous sample.
sampleK	When using method sample or factor, the total taint in the previous sample.

## Details

This section elaborates on the available methods for constructing a prior distribution.

This section elaborates on the available likelihoods and corresponding prior distributions for the likelihood argument.

- poisson: The Poisson likelihood is used as a likelihood for monetary unit sampling (MUS). Its likelihood function is defined as:

$$p(x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

The conjugate  $\text{gamma}(\alpha, \beta)$  prior has probability density function:

$$f(x; \alpha, \beta) = \frac{\beta^\alpha x^{\alpha-1} e^{-\beta x}}{\Gamma(\alpha)}$$

- **binomial:** The binomial likelihood is used as a likelihood for record sampling *with* replacement. Its likelihood function is defined as:

$$p(x) = \binom{n}{k} p^k (1-p)^{n-k}$$

The conjugate  $\text{beta}(\alpha, \beta)$  prior has probability density function:

$$f(x; \alpha, \beta) = \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1}$$

- **hypergeometric:** The hypergeometric likelihood is used as a likelihood for record sampling *without* replacement. Its likelihood function is defined as:

$$p(x = k) = \frac{\binom{K}{k} \binom{N-K}{n-k}}{\binom{N}{n}}$$

The conjugate  $\text{beta-binomial}(\alpha, \beta)$  prior (Dyer and Pierce, 1993) has probability density function:

$$f(k|n, \alpha, \beta) = \binom{n}{k} \frac{B(k + \alpha, n - k + \beta)}{B(\alpha, \beta)}$$

- **none:** This method constructs a prior distribution according to the principle of minimum information.
- **median:** This method constructs a prior distribution so that the prior probabilities of tolerable and intolerable misstatement are equal.
- **hypotheses:** This method constructs a prior distribution with specified prior probabilities for the hypotheses of tolerable and intolerable misstatement. Requires specification of the pHmin and pHplus arguments.
- **arm:** This method constructs a prior distribution according to the assessed risks in the audit risk model. Requires specification of the ir and cr arguments.
- **sample:** This method constructs a prior distribution on the basis of an earlier sample. Requires specification of the sampleN and sampleK arguments.
- **factor:** This method constructs a prior distribution on the basis of last year's results and a weighting factor. Requires specification of the factor, sampleN and sampleK arguments.

## Value

An object of class `jfaPrior` containing:

<code>confidence</code>	the method by which the prior distribution is constructed.
<code>likelihood</code>	the likelihood by which the prior distribution is updated.
<code>method</code>	the method by which the prior distribution is constructed.
<code>expectedError</code>	the expected error input.
<code>N</code>	if specified as input, the population size.
<code>materiality</code>	if specified, the materiality that was used to construct the prior distribution.
<code>description</code>	a description of the prior distribution, including parameters and the implicit sample.
<code>statistics</code>	a list of statistics of the prior distribution, including the mean, mode, median, and upper bound.
<code>specifics</code>	a list of optional specifications of the prior distribution, these depend on the method used.
<code>hypotheses</code>	if a materiality is specified, a list of statistics about the hypotheses including prior probabilities and odds.

**Author(s)**

Koen Derks, <k.derks@nyenrode.nl>

**References**

Derks, K., de Swart, J., Wagenmakers, E.-J., Wille, J., & Wetzels, R. (2019). JASP for audit: Bayesian tools for the auditing practice.

Derks, K., de Swart, J., van Batenburg, P. Wagenmakers, E.-J., & Wetzels, R. (2020). Priors in a Bayesian Audit: How Integrating Information into the Prior Distribution can Improve Audit Transparency and Efficiency.

**See Also**

[planning sampling evaluation](#)

**Examples**

```
library(jfa)

# Specify the materiality, confidence, and expected errors:
materiality <- 0.05 # 5%
confidence <- 0.95 # 95%
expectedError <- 0.025 # 2.5%

# Specify the inherent risk (ir) and control risk (cr):
ir <- 1 # 100%
cr <- 0.6 # 60%

# Create a beta prior distribution according to the Audit Risk Model (arm)
# and a binomial likelihood:
prior <- auditPrior(confidence = confidence, likelihood = "binomial",
                    method = "arm", expectedError = expectedError, materiality = materiality,
                    ir = ir, cr = cr)

print(prior)

# -----
#           jfa Prior Distribution Summary (Bayesian)
# -----
# Input:
#
# Confidence:           0.95
# Expected sample errors: 0.02
# Likelihood:           binomial
# Specifics:            Inherent risk = 1; Internal control risk = 0.6; Detection risk = 0.08
# -----
# Output:
#
# Prior distribution:    beta(2.275, 50.725)
# Implicit sample size: 51
# Implicit errors:      1.27
# -----
# Statistics:
#
# Upper bound:          0.1
# Precision:            7.1%
```

```
# Mode:                0.02
# Mean:                0.04
# Median:              0.04
# -----
```

BuildIt

*BuildIt Construction financial statements***Description**

Fictional data from a construction company in the United States, containing 3500 observations identification numbers, book values, and audit values. The audit values are added for illustrative purposes, as these would need to be assessed by the auditor in the execution stage of the audit.

**Usage**

```
data(BuildIt)
```

**Format**

A data frame with 3500 rows and 3 variables.

**ID** unique record identification number.

**bookValue** book value in US dollars (\$14.47–\$2,224.40).

**auditValue** true value in US dollars (\$14.47–\$2,224.40).

**References**

Derks, K., de Swart, J., Wagenmakers, E.-J., Wille, J., & Wetzels, R. (2019). JASP for audit: Bayesian tools for the auditing practice.

**Examples**

```
data(BuildIt)
```

evaluation

*Evaluation of Audit Samples using Confidence / Credible Bounds***Description**

This function takes a data frame (using `sample`, `bookValue`, and `auditValues`) or summary statistics (using `nSumstats` and `kSumstats`) and evaluates the audit sample according to the specified method. The returned object is of class `jfaEvaluation` and can be used with associated `print()` and `plot()` methods.

## Usage

```
evaluation(confidence = 0.95, method = "binomial", N = NULL,
           sample = NULL, bookValues = NULL, auditValues = NULL, counts = NULL,
           nSumstats = NULL, kSumstats = NULL,
           materiality = NULL, minPrecision = NULL,
           prior = FALSE, nPrior = 0, kPrior = 0,
           rohrbachDelta = 2.7, momentPoptype = "accounts", populationBookValue = NULL,
           csA = 1, csB = 3, csMu = 0.5)
```

## Arguments

confidence	the required confidence level for the bound. Default is 0.95 for 95% confidence.
method	the method that is used to evaluate the sample. This can be either one of poisson, binomial, hypergeometric, stringer, stringer-meikle, stringer-lta, stringer-pvz, rohrbach, moment, direct, difference, quotient, or regression.
N	an integer specifying the total number of units (transactions or monetary units) in the population.
sample	a data frame containing at least a column of Ist values and a column of Soll (true) values.
bookValues	a character specifying the column name for the Ist values in the sample.
auditValues	a character specifying the column name for the Soll values in the sample.
counts	a integer vector of the number of times each transaction in the sample is to be evaluated (due to it being selected multiple times for the sample).
nSumstats	an integer specifying the number of transactions in the sample. If specified, overrides the sample, bookValues and auditValues arguments and assumes that the data come from summary statistics specified by both nSumstats and kSumstats.
kSumstats	a value specifying the sum of taints (proportional errors) found in the sample. If specified, overrides the sample, bookValues and auditValues arguments and assumes that the data come from summary statistics specified by both kSumstats and nSumstats.
materiality	a value specifying the performance materiality as a fraction of the total value (or size) of the population (a value between 0 and 1). If specified, the function also returns the conclusion of the analysis with respect to the performance materiality. The value is discarded when direct, difference, quotient, or regression method is chosen.
minPrecision	a value specifying the required minimum precision. If specified, the function also returns the conclusion of the analysis with respect to the required minimum precision. This value must be specified as a fraction of the total value of the population (a value between 0 and 1).
prior	a logical indicating whether to use a prior distribution when evaluating. Defaults to FALSE for frequentist evaluation. If TRUE, the prior distribution is updated by the corresponding likelihood. Chooses a conjugate gamma distribution for the Poisson likelihood, a conjugate beta distribution for the binomial likelihood, and a conjugate beta-binomial distribution for the hypergeometric likelihood.
nPrior	a value for the prior parameter $\beta$ (number of transactions in the assumed prior sample).
kPrior	a value for the prior parameter $\alpha$ (total tainting in the assumed prior sample).

rohrbachDelta	a value specifying $\Delta$ in Rohrbach's augmented variance bound (Rohrbach, 1993).
momentPoctype	a character specifying the type of population for the modified moment method (Dworin and Grimlund, 1986). Can be either one of accounts or inventory. Options result in different methods for calculating the central moments.
populationBookValue	a value specifying the total value of the transactions in the population. Required when method is one of direct, difference, quotient, or regression, but optional otherwise.
csA	if method = "coxsnell", the $\alpha$ parameter of the prior distribution on the mean taint. Default is set to 1, as recommended by Cox and Snell (1979).
csB	if method = "coxsnell", the $\beta$ parameter of the prior distribution on the mean taint. Default is set to 3, as recommended by Cox and Snell (1979).
csMu	if method = "coxsnell", the mean of the prior distribution on the mean taint. Default is set to 0.5, as recommended by Cox and Snell (1979).

## Details

This section lists the available options for the methods argument.

- poisson: The confidence bound taken from the Poisson distribution. If combined with prior = TRUE, performs Bayesian evaluation using a *gamma* prior and posterior.
- binomial: The confidence bound taken from the binomial distribution. If combined with prior = TRUE, performs Bayesian evaluation using a *beta* prior and posterior.
- hypergeometric: The confidence bound taken from the hypergeometric distribution. If combined with prior = TRUE, performs Bayesian evaluation using a *beta-binomial* prior and posterior.
- stringer: The Stringer bound (Stringer, 1963).
- stringer-meikle: Stringer bound with Meikle's correction for understatements (Meikle, 1972).
- stringer-lta: Stringer bound with LTA correction for understatements (Leslie, Teitlebaum, and Anderson, 1979).
- stringer-pvz: Stringer bound with Pap and van Zuijlen's correction for understatements (Pap and van Zuijlen, 1996).
- rohrbach: Rohrbach's augmented variance bound (Rohrbach, 1993).
- moment: Modified moment bound (Dworin and Grimlund, 1986).
- coxsnell: Cox and Snell bound (Cox and Snell, 1979).
- direct: Confidence interval using the direct method (Touw and Hoogduin, 2011).
- difference: Confidence interval using the difference method (Touw and Hoogduin, 2011).
- quotient: Confidence interval using the quotient method (Touw and Hoogduin, 2011).
- regression: Confidence interval using the regression method (Touw and Hoogduin, 2011).

## Value

An object of class `jfaEvaluation` containing:

confidence	a value specifying the confidence level of the result.
method	the evaluation method that was used.

N	if N is specified, the population size that is used.
n	an integer specifying the sample size used in the evaluation.
k	an integer specifying the number of transactions that contained an error.
t	a value specifying the sum of observed taints.
materiality	if materiality is specified, the performance materiality used.
minPrecision	if minPrecision is specified, the minimum required precision used.
mle	a value specifying the most likely error in the population as a proportion.
precision	a value specifying the difference between the mle and the upper confidence bound as a proportion.
popBookvalue	if specified as input, the total Ist value of the population.
pointEstimate	if method is one of direct, difference, quotient, or regression, the value of the point estimate.
lowerBound	if method is one of direct, difference, quotient, or regression, the value of the lower bound of the interval.
upperBound	if method is one of direct, difference, quotient, or regression, the value of the upper bound of the interval.
confBound	the upper confidence bound on the error percentage.
conclusion	if materiality is specified, the conclusion about whether to approve or not approve the population.
populationK	the assumed total errors in the population. Used in inferences with hypergeometric method.
prior	an object of class 'jfaPrior' to represents the prior distribution.
posterior	an object of class 'jfaPosterior' to represents the posterior distribution.

#### Author(s)

Koen Derks, <k.derks@nyenrode.nl>

#### References

- Cox, D. and Snell, E. (1979). On sampling and the estimation of rare errors. *Biometrika*, 66(1), 125-132.
- Dworin, L., and Grimlund, R. A. (1986). Dollar-unit sampling: A comparison of the quasi-Bayesian and moment bounds. *Accounting Review*, 36-57.
- Leslie, D. A., Teitlebaum, A. D., & Anderson, R. J. (1979). *Dollar-unit sampling: a practical guide for auditors*. Copp Clark Pitman; Belmont, Calif.: distributed by Fearon-Pitman.
- Meikle, G. R. (1972). *Statistical Sampling in an Audit Context: An Audit Technique*. Canadian Institute of Chartered Accountants.
- Pap, G., and van Zuijlen, M. C. (1996). On the asymptotic behavior of the Stringer bound 1. *Statistica Neerlandica*, 50(3), 367-389.
- Rohrbach, K. J. (1993). Variance augmentation to achieve nominal coverage probability in sampling from audit populations. *Auditing*, 12(2), 79.
- Stringer, K. W. (1963). Practical aspects of statistical sampling in auditing. In *Proceedings of the Business and Economic Statistics Section* (pp. 405-411). American Statistical Association.
- Touw, P., and Hoogduin, L. (2011). *Statistiek voor Audit en Controlling*. Boom uitgevers Amsterdam.



**See Also**

[auditPrior planning sampling](#)

**Examples**

```
library(jfa)
set.seed(1)

# Generate some audit data (N = 1000):
data <- data.frame(ID = sample(1000:100000, size = 1000, replace = FALSE),
                   bookValue = runif(n = 1000, min = 700, max = 1000))

# Using monetary unit sampling, draw a random sample from the population.
s1 <- sampling(population = data, sampleSize = 100, units = "mus",
              bookValues = "bookValue", algorithm = "random")
s1_sample <- s1$sample
s1_sample$trueValue <- s1_sample$bookValue
s1_sample$trueValue[2] <- s1_sample$trueValue[2] - 500 # One overstatement is found

# Using summary statistics, calculate the upper confidence bound according
# to the binomial distribution:

e1 <- evaluation(nSumstats = 100, kSumstats = 1, method = "binomial",
                materiality = 0.05)
print(e1)

# -----
#           jfa Evaluation Summary (Frequentist)
# -----
# Input:
#
# Confidence:           95%
# Materiality:          5%
# Minium precision:     Not specified
# Sample size:          100
# Sample errors:         1
# Sum of taints:         1
# Method:               binomial
# -----
# Output:
#
# Most likely error:     1%
# Upper bound:           4.66%
# Precision:             3.66%
# Conclusion:            Approve population
# -----

# Evaluate the raw sample using the stringer bound and the sample counts:

e2 <- evaluation(sample = s1_sample, bookValues = "bookValue", auditValues = "trueValue",
                method = "stringer", materiality = 0.05, counts = s1_sample$counts)
print(e2)

# -----
#           jfa Evaluation Summary (Frequentist)
# -----
```

```

# Input:
#
# Confidence:          95%
# Materiality:         5%
# Minium precision:    Not specified
# Sample size:         100
# Sample errors:       1
# Sum of taints:       1
# Method:              stringer
# -----
# Output:
#
# Most likely error:    0.69%
# Upper bound:         4.12%
# Precision:           3.44%
# Conclusion:          Approve population
# -----

```

---

planning

*Frequentist and Bayesian Planning for Audit Sampling*


---

## Description

This function calculates the required sample size for an audit, based on the poisson, binomial, or hypergeometric likelihood. A prior can be specified to perform Bayesian planning. The returned object is of class `jfaPlanning` and can be used with associated `print()` and `plot()` methods.

## Usage

```

planning(confidence = 0.95, expectedError = 0, likelihood = "poisson", N = NULL,
          materiality = NULL, minPrecision = NULL,
          prior = FALSE, kPrior = 0, nPrior = 0,
          increase = 1, maxSize = 5000)

```

## Arguments

<code>confidence</code>	the confidence level desired from the confidence bound (on a scale from 0 to 1). Defaults to 0.95, or 95% confidence.
<code>expectedError</code>	a fraction representing the percentage of expected mistakes in the sample relative to the total size, or a number ( $\geq 1$ ) that represents the number of expected mistakes.
<code>likelihood</code>	can be one of <code>binomial</code> , <code>poisson</code> , or <code>hypergeometric</code> .
<code>N</code>	the population size (required for hypergeometric calculations).
<code>materiality</code>	a value between 0 and 1 representing the materiality of the audit as a fraction of the total size or value. Can be <code>NULL</code> , but <code>minPrecision</code> should be specified in that case.
<code>minPrecision</code>	The minimum precision to be obtained. Can be <code>NULL</code> , but <code>materiality</code> should be specified in that case.

prior	whether to use a prior distribution when planning. Defaults to FALSE for frequentist planning. If TRUE, the prior distribution is updated by the specified likelihood. Chooses a conjugate gamma distribution for the Poisson likelihood, a conjugate beta distribution for the binomial likelihood, and a conjugate beta-binomial distribution for the hypergeometric likelihood.
kPrior	the prior parameter $\alpha$ (number of errors in the assumed prior sample).
nPrior	the prior parameter $\beta$ (total number of observations in the assumed prior sample).
increase	the desired increase step for the sample size calculation.
maxSize	the maximum sample size that is considered for calculations. Defaults to 5000 for efficiency. Increase this value if the sample size cannot be found due to it being too large (e.g., for a low materiality).

## Details

This section elaborates on the available likelihoods and corresponding prior distributions for the likelihood argument.

- **poisson:** The Poisson likelihood is used as a likelihood for monetary unit sampling (MUS). Its likelihood function is defined as:

$$p(x) = \frac{\lambda^x e^{-\lambda}}{x!}$$

The conjugate  $gamma(\alpha, \beta)$  prior has probability density function:

$$f(x; \alpha, \beta) = \frac{\beta^\alpha x^{\alpha-1} e^{-\beta x}}{\Gamma(\alpha)}$$

- **binomial:** The binomial likelihood is used as a likelihood for record sampling *with* replacement. Its likelihood function is defined as:

$$p(x) = \binom{n}{k} p^k (1-p)^{n-k}$$

The conjugate  $beta(\alpha, \beta)$  prior has probability density function:

$$f(x; \alpha, \beta) = \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1}$$

- **hypergeometric:** The hypergeometric likelihood is used as a likelihood for record sampling *without* replacement. Its likelihood function is defined as:

$$p(x = k) = \frac{\binom{K}{k} \binom{N-K}{n-k}}{\binom{N}{n}}$$

The conjugate  $beta-binomial(\alpha, \beta)$  prior (Dyer and Pierce, 1993) has probability density function:

$$f(k|n, \alpha, \beta) = \binom{n}{k} \frac{B(k + \alpha, n - k + \beta)}{B(\alpha, \beta)}$$

**Value**

An object of class `jfaPlanning` containing:

<code>confidence</code>	the confidence level for the desired population statement.
<code>expectedError</code>	the specified number of errors as a fraction or as a number.
<code>likelihood</code>	the specified likelihood.
<code>N</code>	the population size (only returned in case of a hypergeometric likelihood).
<code>materiality</code>	the value of the specified materiality. Can be <code>NULL</code> .
<code>minPrecision</code>	The minimum precision to be obtained. Can be <code>NULL</code> .
<code>sampleSize</code>	the resulting sample size.
<code>errorType</code>	whether the expected errors were specified as a percentage or as an integer.
<code>expectedSampleError</code>	the number of full errors that are allowed to occur in the sample.
<code>expectedBound</code>	a value specifying the expected upper bound if the sample goes according to plan.
<code>expectedPrecision</code>	a value specifying the expected precision if the sample goes according to plan.
<code>populationK</code>	the assumed population errors (only returned in case of a hypergeometric likelihood).
<code>prior</code>	an object of class <code>jfaPrior</code> that represents the prior distribution.
<code>expectedPosterior</code>	an object of class <code>jfaPosterior</code> that represents the expected posterior distribution.

**Author(s)**

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**References**

Dyer, D. and Pierce, R.L. (1993). On the Choice of the Prior Distribution in Hypergeometric Sampling. *Communications in Statistics - Theory and Methods*, 22(8), 2125 - 2146.

**See Also**

[auditPrior sampling evaluation](#)

**Examples**

```
library(jfa)

# Using the binomial distribution, calculates the required sample size for a
# materiality of 5% when 2.5% mistakes are expected to be found in the sample.

# Frequentist planning with binomial likelihood:

p1 <- planning(confidence = 0.95, expectedError = 0.025, likelihood = "binomial",
               materiality = 0.05)
print(p1)
```

```

# -----
#               jfa Planning Summary (Frequentist)
# -----
# Input:
#
# Confidence:           95%
# Materiality:          5%
# Minimum precision:    Not specified
# Likelihood:           binomial
# Expected sample errors: 6
# -----
# Output:
#
# Sample size:          234
# -----
# Statistics:
#
# Expected upper bound:  5%
# Expected precision:    2.43%
# -----

# Bayesian planning with prior:

prior <- auditPrior(confidence = 0.95, likelihood = "binomial", method = "arm",
                    expectedError = 0.025, materiality = 0.05, cr = 0.6)

p3 <- planning(confidence = 0.95, expectedError = 0.025, materiality = 0.05,
               prior = prior)
print(p3)

# -----
#               jfa Planning Summary (Bayesian)
# -----
# Input:
#
# Confidence:           95%
# Materiality:          5%
# Minimum precision:    Not specified
# Likelihood:           binomial
# Prior distribution:    beta(2.275, 50.725)
# Expected sample errors: 4.23
# -----
# Output:
#
# Sample size:          169
# Posterior distribution: beta(6.5, 215.5)
# -----
# Statistics:
#
# Expected upper bound:  4.99%
# Expected precision:    2.49%
# Expected Bayes factor-+: 9.32
# -----

```

sampling

*Sampling from Audit Populations***Description**

This function takes a data frame and performs sampling according to one of three popular algorithms: random sampling, cell sampling, or fixed interval sampling. Sampling is done in combination with one of two sampling units: records or monetary units. The returned object is of class `jfaSampling` and can be used with associated `print()` and `plot()` methods.

**Usage**

```
sampling(population, sampleSize, units = "records", algorithm = "random",
          bookValues = NULL, intervalStartingPoint = 1, ordered = TRUE,
          ascending = TRUE, withReplacement = FALSE, seed = 1)
```

**Arguments**

<code>population</code>	a data frame containing the population the auditor wishes to sample from.
<code>sampleSize</code>	the number of observations that need to be selected from the population. Can also be an object of class <code>jfaPlanning</code> .
<code>units</code>	can be either <code>records</code> (default) for record sampling, or <code>mus</code> for monetary unit sampling.
<code>algorithm</code>	can be either one of <code>random</code> (default) for random sampling, <code>cell</code> for cell sampling, or <code>interval</code> for fixed interval sampling.
<code>bookValues</code>	a character specifying the name of the column containing the book values (as in the population data).
<code>intervalStartingPoint</code>	the starting point in the interval (used only in fixed interval sampling)
<code>ordered</code>	if <code>TRUE</code> (default), the population is first ordered according to the value of their book values.
<code>ascending</code>	if <code>TRUE</code> (default), order the population in ascending order.
<code>withReplacement</code>	whether sampling should be performed with replacement. Defaults to <code>FALSE</code> .
<code>seed</code>	seed to reproduce results. Default is 1.

**Details**

This first part of this section elaborates on the possible options for the `units` argument:

- `records`: In record sampling, each observation in the population is seen as a sampling unit. An observation of \$5000 is therefore equally likely to be selected as an observation of \$500.
- `mus`: In monetary unit sampling, each monetary unit in the population is seen as a sampling unit. An observation of \$5000 is therefore ten times more likely to be selected as an observation of \$500.

This second part of this section elaborates on the possible options for the `algorithm` argument:

- random: In random sampling each sampling unit in the population is drawn with equal probability.
- cell: In cell sampling the sampling units in the population are divided into a number (equal to the sample size) of intervals. From each interval one sampling unit is selected with equal probability.
- interval: In fixed interval sampling the sampling units in the population are divided into a number (equal to the sample size) of intervals. From each interval one sampling unit is selected according to a fixed starting point (`intervalStartingPoint`).

## Value

An object of class `jfaSampling` containing:

<code>population</code>	a data frame containing the input population.
<code>sample</code>	a data frame containing the selected observations.
<code>units</code>	the sampling units that were used for sampling.
<code>algorithm</code>	the algorithm that was used for sampling.
<code>bookValues</code>	if specified, the name of the specified book value column.

## Author(s)

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## References

Wampler, B., & McEacharn, M. (2005). Monetary-unit sampling using Microsoft Excel. *The CPA journal*, 75(5), 36.

## See Also

[auditPrior planning evaluation](#)

## Examples

```
library(jfa)
set.seed(1)

# Generate some audit data (N = 1000).
population <- data.frame(ID = sample(1000:100000, size = 1000, replace = FALSE),
                          bookValue = runif(n = 1000, min = 700, max = 1000))

# Draw a custom sample of 100 from the population (via random record sampling):

s1 <- sampling(population = population, sampleSize = 100, algorithm = "random",
               units = "records", seed = 1)
print(s1)

# -----
#                               jfa Selection Summary
# -----
# Input:
#
# Population size:           1000
# Requested sample size:    100
```

```

# Sampling units:      Records
# Algorithm:           Random sampling
# -----
# Output:
#
# Obtained sample size: 100
# -----
# Statistics:
#
# Proportion n/N:      0.1
# -----

# Use the result from the planning stage in the sampling stage:

p1 <- planning(materiality = 0.05, confidence = 0.95, expectedError = 0.025,
               likelihood = "binomial")

# Draw a sample via random monetary unit sampling:
s2 <- sampling(population = population, sampleSize = p1, algorithm = "random",
               units = "mus", seed = 1, bookValues = "bookValue")
print(s2)

# -----
#                               jfa Selection Summary
# -----
# Input:
#
# Population size:      1000
# Requested sample size: 234
# Sampling units:      Monetary units
# Algorithm:           Random sampling
# -----
# Output:
#
# Obtained sample size: 234
# -----
# Statistics:
#
# Proportion n/N:      0.23
# Percentage of value:  23.06%
# -----

```



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