

# Package: InPlotSampling (via r-universe)

May 26, 2026

**Type** Package

**Title** Easing the Application of Ranked Set Sampling in Practice

**Version** 0.1.0

**Date** 2021-02-09

**Description** The InPlotSampling package provides a way for researchers to easily implement Ranked Set Sampling in practice. Ranked Set Sampling was originally described by McIntyre (1952) (reprinted in 2005) <[doi:10.1198/000313005X54180](https://doi.org/10.1198/000313005X54180)>. This package takes work by Omer and Kravchuk (2021) <<https://doi.org/10.1007/s13253-021-00439-1>> and enables easy use of the methods.

**License** MIT + file LICENSE

**URL** <https://aagi-aus.github.io/InPlotSampling/>

**BugReports** <https://github.com/AAGI-AUS/InPlotSampling/issues>

**Depends** R (>= 3.5.0)

**Imports** Rcpp, ggplot2, SDaA, FNN

**Suggests** covr, testthat, parallel

**LinkingTo** Rcpp

**Encoding** UTF-8

**LazyData** true

**Roxygen** list(markdown = TRUE)

**RoxygenNote** 7.3.2

**Repository** <https://aagi-aus.r-universe.dev>

**Date/Publication** 2026-03-27 06:43:03 UTC

**RemoteUrl** <https://github.com/AAGI-AUS/InPlotSampling>

**RemoteRef** HEAD

**RemoteSha** 5a444efee2ec3af20769a732b67d3e569735ac87

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bootstrap_sample	<i>Generate bootstrap sample on the provided population.</i>
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---

### Description

Generate bootstrap sample on the provided population.

### Usage

```
bootstrap_sample(pop, n, n_bootstraps)
```

### Arguments

pop	Population data
n	Sample sizes (SBS sample size, PPS sample size).
n_bootstraps	Number of bootstrap samples.

### Value

A summary data frame of the estimator.

coombe2019

*Coombe Vineyard Data from 2019 Season.***Description**

Measurements taken on the Coombe Research Vineyard, University of Adelaide Waite Campus after the 2019 season.

**Usage**

coombe2019

**Format**

A data frame with 352 rows and 11 variables:

**vine\_id** Identifier for the individual vines (1-352)

**rootstock** The rootstock that the vine is growing on (8 levels)

**row** The row in the vineyard (10-20)

**panel** The panel of the vines. A pair of vines is grouped into a panel (1-16)

**trunk\_circ\_18** The trunk circumference of the vine in 2018, measured in cm at watering height (~20 cm above the ground).

**trunk\_circ\_19** The trunk circumference of the vine in 2019, measured in cm at watering height (~20 cm above the ground).

**count\_shoots** Check this?

**non\_count\_shoots** Check this?

**total\_shoots** Sum of count\_shoot and non\_count\_shoot.

**pruning\_weight** Weight of the material removed during pruning in Kg. Check this?

**cordon\_length** The length of the cordon in cm

emergence\_ranks

*Ranks for Seed Emergence.***Description**

Contains the ranks given by 5 rankers of the number of seeds emerged in a sample of 15 plots.

**Usage**

emergence\_ranks

**Format**

A data frame with 6 variables: seed\_emergence, ranker1, ranker2, ranker3, ranker4 and ranker5.

---

InPlotSampling      *InPlotSampling: Easing the Application of Ranked Set Sampling in Practice*

---

### Description

The InPlotSampling package provides a way for researchers to easily implement Ranked Set Sampling in practice. Ranked Set Sampling was originally described by McIntyre (1952) (reprinted in 2005) doi:[10.1198/000313005X54180](https://doi.org/10.1198/000313005X54180). This package takes work by Omer and Kravchuk (2021) <https://doi.org/10.1007/s13253-021-00439-1> and enables easy use of the methods.

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### See Also

Useful links:

- <https://aagi-aus.github.io/InPlotSampling/>
- Report bugs at <https://github.com/AAGI-AUS/InPlotSampling/issues>

---

jps\_estimate      *Computes the estimator for JPS data*

---

### Description

Computes the estimator for JPS data

### Usage

```
jps_estimate(data, set_size, replace = TRUE, model_based, N, alpha)
```

**Arguments**

data	The data to use for estimation.
set_size	Set size for each ranking group.
replace	Logical (default TRUE). Sample with replacement?
model_based	An inference mode: <ul style="list-style-type: none"> <li>• FALSE: design based inference</li> <li>• TRUE: model based inference using super population model</li> </ul>
N	The population size.
alpha	The significance level.

**Value**

A data.frame with the point estimates provided by JPS estimators along with standard error and confidence intervals.

---

jps_sample	<i>Generate JPS sampling on the provided population.</i>
------------	--

---

**Description**

Generate JPS sampling on the provided population.

**Usage**

```
jps_sample(pop, n, H, tau, K, replace = FALSE, with_index = FALSE)
```

**Arguments**

pop	Population that will be sampled.
n	Sample size.
H	Set size for each ranking group.
tau	A parameter which controls ranking quality.
K	Number of rankers.
replace	A boolean which specifies whether to sample with replacement or not.
with_index	A boolean which specifies whether to return the index of the sampled population.

**Value**

A matrix with ranks from each ranker.

**Examples**

```

set.seed(112)
population_size <- 600
# the number of samples to be ranked in each set
H <- 3

with_replacement <- FALSE
sigma <- 4
mu <- 10
n_rankers <- 3
# sample size
n <- 10

rhos <- rep(0.75, n_rankers)
taus <- sigma * sqrt(1 / rhos^2 - 1)

population <- qnorm((1:population_size) / (population_size + 1), mu, sigma)
jps_sample(population, n, H, taus, n_rankers, with_replacement)
#>
#>           Y R1 R2 R3
#> [1,]  6.384461  1  2  2
#> [2,]  1.485141  1  1  1
#> [3,] 13.640711  2  3  2
#> [4,] 15.809136  3  3  2
#> [5,]  6.769463  2  2  1
#> [6,] 14.355524  3  3  3
#> [7,] 10.729740  2  1  3
#> [8,]  6.152453  1  1  1
#> [9,]  8.701285  2  1  2
#> [10,] 13.323884  3  3  3

```

---

population

*Seed Emergence Population.*

---

**Description**

The entire population of actual and estimated seed emergence.

**Usage**

population

**Format**

A data frame with 2640 rows of 2 variables: `actual_seed_emergence` and `estimated_seed_emergence` giving the actual and estimated number of seeds emerged.

---

rss\_jps\_estimate      *Estimate means from RSS or JPS sample.*

---

### Description

Estimate means from RSS or JPS sample.

### Usage

```
rss_jps_estimate(
  data,
  set_size,
  method,
  confidence = 0.95,
  replace = TRUE,
  model_based = FALSE,
  pop_size = NULL
)
```

### Arguments

data	A data frame of JPS or RSS rankings.
set_size	The set size of the ranks.
method	A method used to sample: <ul style="list-style-type: none"> <li>• "JPS": Judgment-post stratified sampling</li> <li>• "RSS": Ranked set sampling</li> </ul>
confidence	The confidence level to use.
replace	Logical (default TRUE). Sample with replacement?
model_based	An inference mode: <ul style="list-style-type: none"> <li>• FALSE: design based inference</li> <li>• TRUE: model based inference using super population model</li> </ul>
pop_size	The population size. Must be provided if <ul style="list-style-type: none"> <li>• sampling without replacement, or</li> <li>• model_based is TRUE.</li> </ul>

### Value

A data.frame with the point estimates provided by different types of estimators along with standard error and confidence intervals.

**Examples**

```

# JPS estimator
set.seed(112)
population_size <- 600
# the number of samples to be ranked in each set
H <- 3

with_replacement <- FALSE
sigma <- 4
mu <- 10
n_rankers <- 3
# sample size
n <- 30

rhos <- rep(0.75, n_rankers)
taus <- sigma * sqrt(1 / rhos^2 - 1)
population <- qnorm((1:population_size) / (population_size + 1), mu, sigma)

data <- InPlotSampling::jps_sample(population, n, H, taus, n_rankers, with_replacement)
data <- data[order(data[, 2]), ]

rss_jps_estimate(
  data,
  set_size = H,
  method = "JPS",
  confidence = 0.80,
  replace = with_replacement,
  model_based = FALSE,
  pop_size = population_size
)
#>      Estimator Estimate Standard Error 80% Confidence intervals
#> 1      UnWeighted   9.570         0.526      8.88,10.26
#> 2      Sd.Weighted   9.595         0.569      8.849,10.341
#> 3 Aggregate Weight   9.542         0.500      8.887,10.198
#> 4      JPS Estimate   9.502         0.650      8.651,10.354
#> 5      SRS estimate   9.793         0.783      8.766,10.821
#> 6      Minimum      9.542         0.500      8.887,10.198

# RSS estimator
set.seed(112)
population_size <- 600
# the number of samples to be ranked in each set
H <- 3

with_replacement <- FALSE
sigma <- 4
mu <- 10
n_rankers <- 3
# sample size
n <- 30

population <- qnorm((1:population_size) / (population_size + 1), mu, sigma)

```

```

rho <- 0.75
tau <- sigma * sqrt(1 / rho^2 - 1)
x <- population + tau * rnorm(population_size, 0, 1)

population <- cbind(population, x)
data <- InPlotSampling::rss_sample(population, n, H, n_rankers, with_replacement)
data <- data[order(data[, 2]), ]

rss_estimates <- rss_jps_estimate(
  data,
  set_size = H,
  method = "RSS",
  confidence = 0.80,
  replace = with_replacement,
  model_based = FALSE,
  pop_size = population_size
)

print(rss_estimates)
#>           Estimator point.est St.error 80% Confidence Interval
#> 1           RSS-1      9.153   0.766           8.148,10.158
#> 2 Aggregate Weighted    9.064   0.652           8.209,9.919

```

---

 rss\_sample

*Generate ranked set sampling (RSS) on the population provided.*


---

### Description

Generate ranked set sampling (RSS) on the population provided.

### Usage

```
rss_sample(pop, n, H, K, replace = FALSE)
```

### Arguments

pop	Population that will be sampled with an auxiliary parameter in the second column.
n	Sample size.
H	Set size for each ranking group.
K	Number of rankers.
replace	A boolean which specifies whether to sample with replacement or not.

### Value

A matrix with ranks from each ranker.

**Examples**

```

set.seed(112)
population_size <- 600
# the number of samples to be ranked in each set
H <- 3

replace <- FALSE
sigma <- 4
mu <- 10
n_rankers <- 3
# sample size
n <- 9

population <- qnorm((1:population_size) / (population_size + 1), mu, sigma)
rho <- 0.75
tau <- sigma * sqrt(1 / rho^2 - 1)
x <- population + tau * rnorm(population_size, 0, 1)

population <- cbind(population, x)
rss_sample(population, n, H, n_rankers, replace)
#>      [,1] [,2] [,3] [,4]
#> [1,]  8.910625  1  3  1
#> [2,] 12.317145  2  2  1
#> [3,] 11.619746  3  3  2
#> [4,]  8.307549  1  2  1
#> [5,]  5.089992  2  2  3
#> [6,]  7.233575  3  3  3
#> [7,] 11.601654  1  2  2
#> [8,]  9.134107  2  1  1
#> [9,] 12.960431  3  3  1

```

---

sbs\_pps\_estimate

*Compute an estimator for SBS PPS sampled data.*


---

**Description**

Compute an estimator for SBS PPS sampled data.

**Usage**

```

sbs_pps_estimate(
  population,
  n,
  y,
  sample_matrix,
  n_bootstraps = 100,
  alpha = 0.05,
  n_cores = getOption("n_cores", 1)
)

```

**Arguments**

population	Population data frame to be sampled with 4 columns. <ol style="list-style-type: none"> <li>Halton numbers</li> <li>X1-coordinate of population unit</li> <li>X2-coordinate of population unit</li> <li>Size measurements of population units</li> </ol>
n	Sample sizes (SBS sample size, PPS sample size).
y	Sample response values.
sample_matrix	Sample data frame to be sampled with 6 columns. <ol style="list-style-type: none"> <li>Halton numbers</li> <li>X1-coordinate of population unit</li> <li>X2-coordinate of population unit</li> <li>Size measurement of population unit</li> <li>Weight</li> <li>Inclusion probability</li> </ol>
n_bootstraps	Number of bootstrap samples.
alpha	The significance level.
n_cores	The number of cores to be used for computational tasks (specify 0 for max). This can also be set by calling options, e.g., options(n_cores = 2).

**Value**

A summary data frame of the estimator.

**Examples**

```
set.seed(112)

# SBS sample size, PPS sample size
sample_sizes <- c(5, 5)

n_population <- 233
k <- 0:(n_population - 1)
x1 <- sample(1:13, n_population, replace = TRUE) / 13
x2 <- sample(1:8, n_population, replace = TRUE) / 8
y <- (x1 + x2) * runif(n = n_population, min = 1, max = 2) + 1
measured_sizes <- y * runif(n = n_population, min = 0, max = 4)

population <- matrix(cbind(k, x1, x2, measured_sizes), ncol = 4)
sample_result <- sbs_pps_sample(population, sample_sizes)

# estimate the population mean and construct a confidence interval
df_sample <- sample_result$sample
sample_id <- df_sample[, 1]
y_sample <- y[sample_id]
```

```
sbs_pps_estimates <- sbs_pps_estimate(  
  population, sample_sizes, y_sample, df_sample,  
  n_bootstrap = 100, alpha = 0.05  
)  
print(sbs_pps_estimates)  
#>   n1 n2 Estimate St.error 95% Confidence intervals  
#> 1  5  5    2.849 0.1760682          2.451,3.247
```

---

sbs\_pps\_heatmap      *Generate a heat map of SBS PPS sample of the provided population.*

---

### Description

Generate a heat map of SBS PPS sample of the provided population.

### Usage

```
sbs_pps_heatmap(pop, sbs_indices, pps_indices)
```

### Arguments

pop	Population data frame to be sampled with 5 columns. <ol style="list-style-type: none"><li>1. Halton numbers</li><li>2. X1-coordinate of population unit</li><li>3. X2-coordinate of population unit</li><li>4. Size measurements of population units</li><li>5. Inclusion probabilities</li></ol>
sbs_indices	Indices of SBS sample.
pps_indices	Indices of PPS sample.

### Value

Heat map of the sample.

---

sbs_pps_sample	<i>Generate probability proportional to size (PPS) and spatially balanced sampling on the population provided.</i>
----------------	--

---

### Description

Generate probability proportional to size (PPS) and spatially balanced sampling on the population provided.

### Usage

```
sbs_pps_sample(population, n, n_cores = getOption("n_cores", 1))
```

### Arguments

population	Population data frame to be sampled with 4 columns. <ol style="list-style-type: none"> <li>1. Halton numbers</li> <li>2. X1-coordinate of population unit</li> <li>3. X2-coordinate of population unit</li> <li>4. Size measurements of population units</li> </ol>
n	Sample sizes (SBS sample size, PPS sample size).
n_cores	The number of cores to be used for computational tasks (specify 0 for max). This can also be set by calling options, e.g., options(n_cores = 2).

### Value

A named list of:

- heatmap: heat map of the sample
- sample: SBS PPS sample of the population

### Examples

```
set.seed(112)

# SBS sample size, PPS sample size
sample_sizes <- c(5, 5)

n_population <- 233
k <- 0:(n_population - 1)
x1 <- sample(1:13, n_population, replace = TRUE) / 13
x2 <- sample(1:8, n_population, replace = TRUE) / 8
y <- (x1 + x2) * runif(n = n_population, min = 1, max = 2) + 1
measured_sizes <- y * runif(n = n_population, min = 0, max = 4)

population <- matrix(cbind(k, x1, x2, measured_sizes), ncol = 4)
sample_result <- sbs_pps_sample(population, sample_sizes)
```

```
print(sample_result$sample)
#>   sbs_pps_indices      x1      x2      size      weight inclusion_probability
#> 1          87 0.4615385 0.625 0.4665423 0.000000000 0.02319163
#> 2          88 0.1538462 0.625 1.7389902 0.000000000 0.02790409
#> 3          89 0.8461538 0.625 7.0815547 0.000000000 0.04749104
#> 4          90 0.6923077 0.750 9.5428032 0.000000000 0.05640733
#> 5          91 0.2307692 0.750 5.1375136 0.000000000 0.04039996
#> 6         173 0.1538462 0.500 6.6400168 0.005024130 0.04588620
#> 7          26 0.6153846 0.500 4.3146186 0.003264631 0.03738898
#> 8         232 0.8461538 0.750 12.0057856 0.009084108 0.06526583
#> 9         171 0.6153846 0.750 6.9029083 0.005223046 0.04684225
#> 10         29 0.8461538 0.375 4.6324720 0.003505133 0.03855377
```

---

single\_bootstrap      *Generate a single bootstrap sample on the provided population.*

---

### Description

Generate a single bootstrap sample on the provided population.

### Usage

```
single_bootstrap(pop, n)
```

### Arguments

pop	Population data
n	Sample sizes (SBS sample size, PPS sample size).

### Value

A summary data frame of the estimator.

---

two\_stage\_cluster\_sample      *Generate two-stage cluster sampling on the population provided.*

---

### Description

Generate two-stage cluster sampling on the population provided.

**Usage**

```
two_stage_cluster_sample(
  pop,
  sampling_strategies,
  n,
  H,
  replace,
  ni,
  Hi,
  replace_i
)
```

**Arguments**

pop	Population that will be sampled with these ordered columns: <ol style="list-style-type: none"> <li>1. Parent id: an index to denotes the parent of the record</li> <li>2. Parent auxiliary parameter: an auxiliary parameter for ranking parents</li> <li>3. Child auxiliary parameter: an auxiliary parameter for ranking children</li> </ol>
sampling_strategies	(first stage sampling strategy, second stage sampling strategy), e.g., c('srs', 'jps'). <ul style="list-style-type: none"> <li>• 'srs': simple random sampling without replacement</li> <li>• 'jps': JPS sampling</li> </ul>
n	Number of samples in the first stage.
H	Set size for each ranking group in the first stage.
replace	A boolean which specifies whether to sample with replacement or not in the first stage (applicable only for JPS sampling).
ni	Number(s) of samples in the second stage. Can be a single number or a vector of n numbers.
Hi	Set size for each ranking group in the second stage. Can be a single number or a vector of n numbers.
replace_i	A boolean which specifies whether to sample with replacement or not in the second stage (applicable only for JPS sampling).

**Value**

A matrix with ranks from each ranker.

**Examples**

```
set.seed(112)
parent_size <- 300
child_size <- 50
# the number of samples to be ranked in each set
H <- 3
```

```

sampling_strategies <- c("jps", "jps")
replace <- FALSE
mu <- 10
sigma <- 4
n <- 4

parent_indices <- rep(1:parent_size, child_size)
parent_aux <- abs(qnorm(1:parent_size / (parent_size + 1), mu, sigma) + 5 * rnorm(parent_size, 0, 1))
child_aux <- abs(parent_aux + 10 * rnorm(parent_size * child_size, 0, 1))

population <- cbind(parent_indices, rep(parent_aux, child_size), child_aux)
two_stage_cluster_sample(population, sampling_strategies, n, H, replace, 6, 3, FALSE)
#>      parent_id parent_rank child_id  child_aux child_rank
#> [1,]      201          1     7101  2.2349453          1
#> [2,]      201          1    12801  9.7175545          3
#> [3,]      201          1     6501  7.9207230          1
#> [4,]      201          1     9801  5.7644835          2
#> [5,]      201          1    10701 13.8089335          3
#> [6,]      201          1     3501  0.3598331          1
#> [7,]      254          2     8654 17.3059292          3
#> [8,]      254          2    11354 15.0837335          2
#> [9,]      254          2     9254  6.0103919          2
#> [10,]     254          2     2954 12.7011502          2
#> [11,]     254          2    14954  5.1158133          2
#> [12,]     254          2    13754  5.8931551          1
#> [13,]      74          1     8474  4.3393349          1
#> [14,]      74          1     9674 15.0512523          2
#> [15,]      74          1     6674 12.9022479          3
#> [16,]      74          1      674  2.9209174          2
#> [17,]      74          1     7274  7.2500468          3
#> [18,]      74          1     6374  7.0925954          1
#> [19,]     223          3     9223 28.4694257          3
#> [20,]     223          3      223  4.4001977          1
#> [21,]     223          3     9823 22.8676415          3
#> [22,]     223          3    11923 26.4531048          3
#> [23,]     223          3      823 20.8714211          2
#> [24,]     223          3     9523  8.1783058          1

```

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