# Package: SparseMSE (via r-universe)

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Title 'Multiple Systems Estimation for Sparse Capture Data'

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Author Lax Chan [aut, cre], Bernard Silverman [aut], Kyle Vincent [aut]

Maintainer Lax Chan <laxchan77@gmail.com>

Description Implements the routines and algorithms developed and analysed in ``Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists" Chan, L, Silverman, B. W., Vincent, K (2019) <https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748> and in ``Bootstrapping multiple systems estimates to account for model selection" Silverman, B. W., Chan, L, Vincent, K (2023)<https://doi.org/10.1007/s11222-023-10346-9>. This package explicitly handles situations where there are pairs of lists which have no observed individuals in common. It deals correctly with parameters whose estimated values can be considered as being negative infinity. It also addresses other possible issues of non-existence and non-identifiability of maximum likelihood estimates.

URL https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748

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```
ancestors
```

*Find the "ancestors" of a given capture history* 

## Description

Given any encoded capture history and the number of lists, find all the encoded capture histories that are included in the original capture history

#### Usage

ancestors(k, nlists = 10)

## Arguments

k	An encoded capture history
nlists	The total number of lists

#### Value

a vector giving the encoded versions of the ancestors

#### References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

ancestors(2,10)
ancestors(1,5)

Artificial\_3

#### Description

This is a simple data set based on three lists, which gives examples of models that fail on one or the other of the criteria tested by checkident. This is Table 2 in Chan, Silverman and Vincent (2021).

#### Usage

Artificial\_3

## Format

An object of class data. frame with 4 rows and 4 columns.

## Details

If all three two-list effects are included in the fitted model then the linear program in checkident yields a strictly positive value but the matrix A is not of full column rank, so the parameters are not identifiable. If the model contains AB either alone or in conjunction with one of AC and BC, then the linear program result is zero, so the MLE does not exist. If only main effects are considered, or if either or both of AC and BC, but not AB are included, then the model passes both tests.

#### References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116(535)**, 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

assemble\_bic

Models BICS, abundance and maxorder.

## Description

This routine sorts the models in increasing order according to their BICs, returns the sorted models with their corresponding BICs and abundance. The original data as well as the maxorder of the models considered are returned as well.

## bcaconfvalues

#### Usage

```
assemble_bic(
  xdata,
  maxorder = dim(xdata)[2] - 2,
  checkexist = T,
  removeFRfail = T,
  ...
)
```

#### Arguments

xdata	The original data matrix with capture histories and counts.
maxorder	Maximum order of models to be included
checkexist	If T then the Fienberg-Rinaldo condition is checked for each model
removeFRfail	If checkexist is T then models which fail the FR condition are removed from the results
	Parameters to be fed to gethiermodels.

## Value

A list with the following components

res A matrix with the models considered, their abundance, BIC and their order, sorted into increasing order of BIC

xdata The original data matrix with capture histories and counts.

maxorder Order parameter that was feed into gethiermodels

## Examples

```
data(hiermodels)
data(Korea)
assemble_bic(Korea, checkexist=T)
```

bcaconfvalues BCa confidence intervals

# Description

The BCa confidence intervals use percentiles of the bootstrap distribution of the population size , but adjust the percentile actually used. The adjusted percentiles depend on an estimated bias parameter, and the quantile function of the estimated bias parameter is the proportion of the bootstrap estimates that fall below the estimate from the original data, and an estimated acceleration factor, which derivation depends on a jackknife approach. This routine is called internally by estimatepopulation.

## Usage

```
bcaconfvalues(
    bootreps,
    popest,
    ahat,
    alpha = c(0.025, 0.05, 0.1, 0.16, 0.84, 0.9, 0.95, 0.975)
)
```

## Arguments

bootreps	Point estimates of total population sizes from each bootstrap sample
popest	A point estimate of the total population of the original data set.
ahat	the estimated acceleration factor
alpha	Bootstrap quantiles of interests

## Value

BCa confidence intervals

#### References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116(535)**, 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

DiCiccio, T. J. and Efron, B. (1996). Bootstrap Confidence Intervals. *Statistical Science*, **40**(3), 189-228.

Efron, B. (1987). Better Bootstrap Confidence Intervals. *Journal of the American Statistical Association*, **82(397)**, 171-185.

BICandbootstrapsim Comparison of BIC approach and BCa approach

## Description

This routine carries out the simulation study as detailed in Section 3.4 of Chan, Silverman and Vincent (2021). If the original data set has low counts, so that there is a possibility of a simulated data set containing empty lists, then it may be advisable to use the option noninformativelist=TRUE.

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## BICandbootstrapsim

## Usage

```
BICandbootstrapsim(
   zdat,
   nsims = 100,
   nboot = 100,
   pthresh = 0.02,
   iseed = 1234,
   alpha = c(0.025, 0.05, 0.1, 0.16, 0.5, 0.84, 0.9, 0.95, 0.975),
   noninformativelist = F,
   verbose = F,
   ...
)
```

## Arguments

zdat	Data matrix with $t + 1$ columns. The first t columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last
	B are constructed if not supplied. Where a capture history is not explicitly
	listed, it is assumed that it has zero count.
nsims	Number of simulations to be carried out.
nboot	Number of bootstrap replications for each simulation
pthresh	p-value threshold used in estimatepopulation.0.
iseed	seed for initialization.
alpha	bootstrap quantiles of interests.
noninformativel	list
	if noninformativelist=TRUE then each generated data set in the simulation study (including all bootstrap replications) will be passed to removen on informativelists
verbose	If verbose=FALSE, then the progress of the simulation will not show. If verbose=TRUE, then the progress of the simulation will be shown.
	other arguments.

## Value

A list with components as below

popest Total population point estimate from the original data using estimatepopulation.0 with default threshold.

BICmodels The best model chosen by the BIC at each simulation.

BICvals Point estimates of the total population and standard error of the best model chosen by the BIC at each simulation.

simreps Counts associated to each capture history at each simulation.

modelmat A full capture history matrix excluding the row corresponding to the dark figure.

popestsim Total population estimate given by the BCa method in each simulation.

BCaquantiles bootstrap confidence intervals given by the BCa method.

BICconf confidence interval given by the BIC method.

## References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116(535)**, 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

DiCiccio, T. J. and Efron, B. (1996). Bootstrap Confidence Intervals. *Statistical Science*, **40**(**3**), 189-228.

Rivest, L-P. and Baillargeon, S. (2014) Rcapture. CRAN package. Available from Available from https://CRAN.R-project.org/package=Rcapture.

#### Examples

```
zdat=UKdat_5
BICandbootstrapsim(zdat,nsims=1000, nboot=100, pthresh=0.02, iseed=1234, noninformativelist=T)
```

bicktopahatcal Acceleration factor calculation

#### Description

Returns acceleration factors for each value of ntop for given modelorder

#### Usage

```
bicktopahatcal(z, modelorder)
```

## Arguments

Z	The output of assemble_bic, bootstrapcal and jackknifecal
modelorder	the order of the considered models

## Value

the estimated acceleration factor

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## BICrank\_tiebreak

## Examples

```
data(Korea)
z=assemble_bic(Korea, checkexist=T)
z=jackknifecal(z, checkexist=T)
nmods= dim(z$jackabund)[1]
modscores = (1:nmods)
modelorder= order(modscores)
bicktopahatcal(z,modelorder)
```

BICrank\_tiebreak Return a BIC rank matrix breaking ties with each roles

## Description

This routine takes the ouput of find\_bic\_rank\_matrix and produce a BIC rank matrix with a specified order of rank breaking ties with each previous column.

## Usage

```
BICrank_tiebreak(BICmatrix_prop, k)
```

## Arguments

BICmatrix\_prop The output from find\_bic\_rank\_matrix k The order of BIC rank

#### Value

A BIC rank matrix with the last column breaking ties with the previous ones.

#### References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

```
data(Korea)
z=assemble_bic(Korea)
z=jackknifecal(z)
BICmatrix_prop= find_bic_rank_matrix(z)
BICmatrix_break = BICrank_tiebreak(BICmatrix_prop, 2)
```

bootstrapcal

## Description

This routine takes the output from assemble\_bic or subsetmat and returns bootstrap abundance matrix and BIC matrix. This version makes use of the checkident.2.

#### Usage

```
bootstrapcal(
    z,
    nboot = 1000,
    iseed = 1234,
    checkexist = T,
    saveinterval = Inf,
    savefile = "bootout.Rdata"
)
```

#### Arguments

Z	Results from assemble_bic or subsetmat
nboot	The number of bootstrap replications.
iseed	Integer seed to allow for replicability.
checkexist	If checkexist=TRUE, check for existence, else it does not check for existence.
saveinterval	If this is set to a finite value, the output list z will be saved every time the number of replications is a multiple of it. A message will be printed every time it is saved.
savefile	The file to which the output will be saved if saveinterval is set to a finite value.

## Value

The original input list z with the additional components

bootabund Bootstrap abundance matrix

bootbic Bootstrap BIC matrix

If there are already components with these names they will be overwritten.

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

#### bootstrap\_mse

## Examples

```
z=assemble_bic(Korea)
bootstrapcal(z, checkexist=T, saveinterval=50, savefile="Koreabootresults.Rdata")
```

bootstrap\_mse A routine for naive user

## Description

A routine for naive user

## Usage

```
bootstrap_mse(
   zdat,
   maxorder = dim(zdat)[2] - 2,
   nboot = 10000,
   iseed = 1234,
   alpha = c(0.025, 0.1, 0.9, 0.975)
)
```

## Arguments

zdat	Data matrix with $t + 1$ columns. The first t columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last column is the count of cases with that particular capture history.
maxorder	Maximum order of models to be included.
nboot	Number of bootstrap replications.
iseed	seed for initialisation
alpha	Levels of confidence intervals to be constructed and assessed

## Value

A list with the following components

confvals BCa confidence intervals for each considered model

probests Corresponding probabilities of the estimates

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

```
data(Korea)
bootstrap_mse(Korea)
```

boundary\_captures Given a vector of captures, find those which are not in the vector but all of whose parents are

### Description

Call the resulting set the "boundary". Supposing that the current set of captures is a hierarchical model, that property will be preserved if a capture in the boundary is added to it. The routine is called internally by find\_neighbour\_hierarchies.

## Usage

```
boundary_captures(kcap, nlists)
```

## Arguments

kcap	An encoded capture history that corresponds to the row number of the capture history data set
nlists	The total number of lists

## Value

a vector giving the encoded versions of the descendants

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

```
modelstr = "[12,23]"
nlists=4
zhierroots = convert_from_hierarchy(modelstr, F)
zhier = unique(ancestors(zhierroots, nlists))
boundary_captures(zhier,nlists)
```

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#### Description

For multiple systems estimation model corresponding to a specified set of two-list effects, set up the GLM model formula and data matrix.

## Usage

buildmodel(zdat, mX)

#### Arguments

zdat	Data matrix with $t + 1$ columns. The first t columns, each corresponding to a
	particular list, are 0s and 1s defining the capture histories observed. The last
	column is the count of cases with that particular capture history. List names A,
	B, are constructed if not supplied. Where a capture history is not explicitly listed, it is assumed that it has zero counts.
mΧ	A $2 \times k$ matrix giving the k two-list effects to be included in the model. Each column of mX contains the numbers of the corresponding pair of lists. If mX = 0, then all two-list effects are included. If mX = NULL, no such effects are included and the main effects model is fitted.

#### Value

A list with components as below.

datamatrix A matrix with all possible capture histories, other than those equal to or containing non-overlapping pairs indexed by parameters that are within the model specified by mX. A nonoverlapping pair is a pair of lists (i, j) such that no case is observed in both lists, regardless of whether it is present on any other lists. If (i, j) is within the model specified by mX, all capture histories containing both i and j are then excluded.

model form The model formula suitable to be called by the Generalized Linear Model function glm. Model terms corresponding to non-overlapping pairs are not included, because they are handled by removing appropriate rows from the data matrix supplied to glm. The list of non-overlapping pairs are provided in emptyoverlaps. See Chan, Silverman and Vincent (2021) for details.

emptyoverlaps A matrix with two rows, whose columns give the indices of non-overlapping pairs of lists where the parameter indexed by the pair is within the specified model. The column names give the names of the lists corresponding to each pair.

#### References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116(535)**, 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

## Examples

```
data(NewOrl)
buildmodel(NewOrl, mX=NULL)
#Build a matrix that contains all two-list effects
m=dim(NewOrl)[2]-1
mX = t(expand.grid(1:m, 1:m)); mX = mX[ , mX[1,]<mX[2,]]
# With one two-list effect
buildmodel(NewOrl, mX=mX[,1])
#With three two-list effects
buildmodel(NewOrl, mX=mX[,1:3])
```

buildmodelmatrix

Build the model matrix based on particular data, as required to check for identifiability and existence of the maximum likelihood estimate

## Description

This routine builds a model matrix as required by the linear program check checkident and checks if the matrix is of full rank. In addition, for each individual list, and for each pair of lists included in the model, it returns the total count of individuals appearing on the specific list or lists whether or not in combination with other lists.

#### Usage

buildmodelmatrix(zdat, mX = NULL)

## Arguments

zdat	Data matrix with $t + 1$ columns. The first t columns, each corresponding to a
	particular list, are 0s and 1s defining the capture histories observed. The last
	column is the count of cases with that particular capture history. List names A,
	B, are constructed if not supplied. Where a capture history is not explicitly
	listed, it is assumed that it has zero count.
mХ	A $2 \times k$ matrix giving the k two-list parameters to be included in the model.
	Each column of mX contains the numbers of the corresponding pair of lists. If mX
	= 0, then all two-list parameters are included. If mX = NULL, no such parameters
	are included and the main effects model is fitted.

## Value

A list with components as below

modmat The matrix that maps the parameters in the model (excluding any corresponding to nonoverlapping lists) to the log expected value of the counts of capture histories that do not contain non-overlapping pairs in the data.

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## checkallmodels

tvec A vector indexed by the parameters in the model, excluding those corresponding to nonoverlapping pairs of lists. For each parameter the vector contains the total count of individuals in all the capture histories that include both the lists indexed by that parameter.

rankdef The column rank deficiency of the matrix modmat. If rankdef = 0, the matrix has full column rank.

## Examples

```
data(NewOrl)
buildmodelmatrix(NewOrl, mX=NULL)
```

checkallmodels Check all possible models for existence and identifiability

## Description

This routine efficiently checks whether every possible model satisfies the conditions for the existence and identifiability of an extended maximum likelihood estimate. For identifiability it is only necessary to check the model containing all two-list effects. For existence, the approach set out in Chan, Silverman and Vincent (2021) is used. This uses the linear programming approach described in a more general and abstract context by Fienberg and Rinaldo (2012).

## Usage

```
checkallmodels(zdat, nreport = 1024)
```

#### Arguments

zdat	Data matrix with $t + 1$ columns. The first $t$ columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last column is the count of cases with that particular capture history. List names A, B, are constructed if not supplied. Where a capture history is not explicitly listed, it is assumed that it has zero count.
nreport	A message is printed for every nreport models checked. It gives the number of top level models to be checked altogether, and also the number of models found so far for which the estimate does not exist.

#### Details

If the extended maximum likelihood estimator exists for a particular model, then it will still exist if one or more overlapping pairs are removed from the model. This allows the search to be carried out efficiently, as follows:

1. Search over 'top-level' models which contain all overlapping pairs. The number of such models is  $2^k$ , where k is the number of non-overlapping pairs, because there are  $2^k$  possible choices of the set of non-overlapping pairs to include in the model.

2. For each top-level model, if the estimate exists there is no need to consider that model further. If the estimate does not exist, then a branch search is carried out over the tree structure obtained by successively leaving out overlapping pairs.

#### Value

The routine prints a message if the model with all two-list parameters is not identifiable. As set out above, it gives regular progress reports in the case where there are a large number of models to be checked.

If all models give estimates which exist, then a message is printed to that effect.

If there are models for which the estimate does not exist, the routine reports the number of such models and returns a matrix whose rows give the parameters included in them.

## References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116**(**535**), 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

Fienberg, S. E. and Rinaldo, A. (2012). Maximum likelihood estimation in log-linear models. Ann. Statist. 40, 996-1023. Supplementary material: Technical report, Carnegie Mellon University. Available from http://www.stat.cmu.edu/~arinaldo/Fienberg\_Rinaldo\_Supplementary\_Material.pdf.

#### Examples

data(Artificial\_3)
data(Western)
checkallmodels(Artificial\_3)
checkallmodels(Western)

checkident

Check a model for the existence and identifiability of the maximum likelihood estimate

#### Description

Apply the linear programming test as derived by Fienberg and Rinaldo (2012), and a calculation of the rank of the design matrix, to check whether a particular model yields an identifiable maximum likelihood estimate based on the given data. The linear programming problem is as described on page 3 of Fienberg and Rinaldo (2012), with a typographical error corrected. Further details are given by Chan, Silverman and Vincent (2021).

## Usage

checkident(zdat, mX = 0, verbose = FALSE)

## checkident

#### Arguments

zdat	Data matrix with $t + 1$ columns. The first $t$ columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last column is the count of cases with that particular capture history. List names A, B, are constructed if not supplied. Where a capture history is not explicitly listed, it is assumed that it has observed count zero.
mX	A $2 \times k$ matrix giving the k two-list parameters to be included in the model. Each column of mX contains the numbers of the corresponding pair of lists. If mX = 0, then all two-list interactions are included. If mX = NULL, no two-list parameters are included and the main effects model is fitted.
verbose	Specifies the output. If FALSE then the error code is returned. If TRUE then in addition the routine prints an error message if the model/data fail either of the two tests, and also returns both the error code and the lp object.

#### Value

If verbose=FALSE, then return the error code ierr which is 0 if there are no errors, 1 if the linear program test shows that the maximum likelihood estimate does not exist, 2 if it is not identifiable, and 3 if both tests are failed.

If verbose=TRUE, then return a list with components as below

ierr As described above.

zlp Linear programming object, in particular giving the value of the objective function at optimum.

## References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116**(**535**), 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

Fienberg, S. E. and Rinaldo, A. (2012). Maximum likelihood estimation in log-linear models. Ann. Statist. 40, 996-1023. Supplementary material: Technical report, Carnegie Mellon University. Available from http://www.stat.cmu.edu/~arinaldo/Fienberg\_Rinaldo\_Supplementary\_Material.pdf.

```
data(Artificial_3)
#Build a matrix that contains all two-list effects
m=dim(Artificial_3)[2]-1
mX = t(expand.grid(1:m, 1:m)); mX = mX[ , mX[1,]<mX[2,]]
# When the model is not identifiable
checkident(Artificial_3,mX=mX, verbose=TRUE)
# When the maximum likelihood estimate does not exist
checkident(Artificial_3, mX=mX[,1],verbose=TRUE)
#Model passes both tests
checkident(Artificial_3, mX=mX[,2:3],verbose=TRUE)</pre>
```

checkident.1

## Description

This routine performs the Fienberg-Rinaldo linear program check in the framework of extended maximum likelihood estimates, the parameters estimates exist if and only if the return value of the check is nonzero

## Usage

checkident.1(parset, datlist)

## Arguments

parset	Either the hierarchical representation of the model, or the vector of the corre-
	sponding capture histories to the model.
datlist	The output of ingest_data on the data set

#### Value

The value of the linear program. The parameter estimates within the extended ML framework exist if and only if this value is nonzero.

## References

Silverman, B. W., Chan, L. and Vincent, K., (2022). Bootstrapping Multiple Systems Estimates to Account for Model Selection

Fienberg, S. E. and Rinaldo, A. (2012). Maximum likelihood estimation in log-linear models. Ann. Statist. 40, 996-1023. Supplementary material: Technical report, Carnegie Mellon University. Available from http://www.stat.cmu.edu/~arinaldo/Fienberg\_Rinaldo\_Supplementary\_Material.pdf.

```
data(Korea)
datlist = ingest_data(Korea)
parset ="[0,0,1]"
checkident.1(parset,datlist)
```

checkident.2

Carry out the Fienberg-Rinaldo procedure on an array of data vectors and a vector of models

## Description

Suppose we have a collection of different data outcomes on the same set of capture histories and a vector of models. Typically the data outcomes will be bootstrap replications. This routine finds the unique support patterns among the data and hence economises the task of finding which model/data combinations satisfy the Fienberg-Rinaldo condition

## Usage

checkident.2(x, xcap, zmods)

## Arguments

х	a matrix of data observations for a common capture matrix
хсар	the incidence matrix of the capture histories corresponding to the rows of x
zmods	a vector of models

## Value

a matrix with rows corresponding to the models and columns to the columns of x, with elements taking the value T if the FR linear program for a vector of 0s and 1s with the same zero pattern as the x data yields a strictly positive value

#### References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

```
data(Korea)
z=assemble_bic(Korea)
z=bootstrapcal(z)
n2=dim(z$xdata)[2]
checkident.2(z$bootreplications, z$xdata[,-n2], dimnames(z$res)[[1]])
```

checkthetasubset

#### Description

This routine leaves out a particular set of parameters corresponding to the two-list effects from the parameter set theta. For the resulting model, it constructs the linear programming problem to check whether the extended maximum likelihood estimates of the parameters exists. It is called internally by checkallmodels.

## Usage

checkthetasubset(zset, amat, tvec, nlists)

#### Arguments

zset	set of indices that is not included, numbered among the two-list effects only
amat	a design matrix
tvec	vector of sufficient statistics
nlists	number of lists in the original capture-recapture matrix

## Value

If the return result is TRUE, the linear program shows that the extended maximum likelihood estimate does not exist. If the return result is FALSE, the estimate exists.

#### References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistcal Association*, **116**(**535**), 1297-1306, Available from https://www.tandfonline.com/ doi/full/10.1080/01621459.2019.1708748.

child\_captures Find the "children" of a given capture history

## Description

Given any encoded capture history that corresponds to the row number of the capture history data set and the number of lists, find the encoded capture histories which are obtained by adding one more list in turn

#### Usage

```
child_captures(k, nlists)
```

#### Arguments

k	An encoded capture history that corresponds to the row number of the capture
	history data set
nlists	The total number of lists

## Value

a vector giving the encoded versions of the children

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

#### Examples

```
child_captures(2,5)
child_captures(1,4)
```

convert\_from\_hierarchy

Find the vector of captures corresponding to a given hierarchical model

## Description

Given a hierarchical model, find the vector of all the corresponding encoded captures

## Usage

```
convert_from_hierarchy(modelstr, findancestors = T)
```

#### Arguments

modelstr	A given hierarchical model
findancestors	If T then find all the captures. If F then just return the encoded defining histories of the hierarchy

#### Value

The encoded capture histories that corresponds to the row number of the capture history data set

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

```
modelstr = "[12,23]"
convert_from_hierarchy(modelstr)
modelstr = "[12,3]"
convert_from_hierarchy(modelstr, findancestors=F)
```

convert\_to\_hierarchy Find hierarchical representation of a vector of captures

## Description

Given a vector of encoded captures defining a hierarchical model, re-express it in hierarchical model form

## Usage

```
convert_to_hierarchy(kcap, nlists)
```

## Arguments

kcap	A vector of captures
nlists	The number of lists

## Value

A hierarchical representation of the vector of encoded captures.

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

```
kcap=c(1,2,3,5,4)
nlists=3
convert_to_hierarchy(kcap, nlists)
```

22

count\_triples

## Description

The routine counts the number of subsets of size three of lists such that every pair of lists in the triple overlaps. If the number is zero, then the model with all two-list effects is unidentifiable.

#### Usage

```
count_triples(zdat)
```

## Arguments

## zdat

Data matrix with t + 1 columns. The first t columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last column is the count of cases with that particular capture history. List names A, B, ... are constructed if not supplied. Where a capture history is not explicitly listed, it is assumed that it has zero count.

#### Value

a count of subsets of size three of lists such that every pair of lists in the triple overlaps.

### Examples

```
data(Western)
data(Artificial_3)
count_triples(Western)
count_triples(Artificial_3)
```

decode\_capture Decode capture history

#### Description

Given a capture history as a number and the number of lists, decode it into a logical vector giving presence or absence in the capture history.

#### Usage

```
decode_capture(k, nlists)
```

descendants

#### Arguments

k	The capture history to be decoded
nlists	The number of lists

## Value

A logical vector of length nlists giving presence or absence in the capture history

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

#### Examples

```
decode_capture(2,5)
decode_capture(1,4)
```

descendants

Find the "descendants" of a given capture history

#### Description

Given any encoded capture history, find all the encoded capture histories that include the original capture history and any other lists

### Usage

```
descendants(k, nlists, omitk = FALSE)
```

## Arguments

k	An encoded capture history
nlists	The total number of lists
omitk	Determine whether the original capture history is included as a descendant of itself. If omitk=T it is not.

## Value

a vector giving the encoded versions of the descendants

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

descendants(2,5)
descendants(5,10)

downhill\_bootstrapcal Bootstrap downhill

## Description

Construct bootstrap replications and use the downhill fit method to obtain point estimates of total population sizes from each bootstrap sample.

## Usage

```
downhill_bootstrapcal(
   xdata,
   nboot = 1000,
   iseed = 1234,
   checkid = T,
   verbose = F,
   maxorder = dim(xdata)[2] - 2
)
```

## Arguments

xdata	original data matrix
nboot	number of bootstrap replicates
iseed	random seed
checkid	If it is T, then checkident.1 is called and it performs the Fienberg-Rinaldo linear program check for the existence of the estimates
verbose	If TRUE, return the list of extra output from downhill_fit
maxorder	Maximum order of models to be included

## Value

Point estimates of total population sizes from each bootstrap sample.

#### References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

data(Korea)
downhill\_bootstrapcal(Korea)

downhill\_fit

## Description

Find a local optimum by downhill search among hierarchical models

## Usage

```
downhill_fit(
  counts,
  desmat,
  maxorder = dim(desmat)[2] - 1,
  checkid = T,
  niter = 20,
  verbose = F
)
```

## Arguments

counts	Observed counts for the capture histories defined by desmat
desmat	Incidence matrix defining the capture histories observed with counts given by counts
maxorder	Maximum order of models to be included
checkid	If it is T, then checkident.1 is called and it performs the Fienberg-Rinaldo linear program check for the existence of the estimates
niter	Number of iterations
verbose	Specifies the output, if F then only returns the best value, if T, returns a more detailed list of objects

#### Value

A list with the following components

optimum\_hierarchy Optimal hierarchical model

minimum\_value hierarchical model with the minimum value

hierarchies\_considered hierarhical models considered

function\_values Values of function

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## downhill\_funs

## Examples

```
data(Korea)
xdata=Korea
counts = xdata[,dim(xdata)[2]]
desmat = xdata[,1:(dim(xdata)[2]-1)]
downhill_fit(counts, desmat)
```

downhill\_funs Downhill wrapper

#### Description

The routine carries out pointwise, bootstrap and jackknife BCa confidence interval calculation and it calls the following routines downhill\_fit, downhill\_bootstrapcal, downhill\_jackknifecal and bcaconfvalues.

## Usage

```
downhill_funs(
   xdata,
   maxorder = dim(xdata)[2] - 2,
   checkid = T,
   verbose = F,
   nboot = 1000,
   alpha = c(0.025, 0.05, 0.95, 0.975)
)
```

## Arguments

xdata	The data matrix
maxorder	Maximum order of models to be included
checkid	if TRUE, do the Fienburg-Renaldo test
verbose	if TRUE, return the list of extra output from <code>downhill_fit</code>
nboot	The number of bootstrap replications
alpha	Bootstrap quantiles of interests.

## Value

A list with the following components

point\_est A point estimate using the downhill fit approachboot\_res Point estimates of total population sizes from each bootstrap samplejack\_res The estimated acceleration factorBCaconf\_int BCa confidence intervals

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

data(Korea) xdata=Korea downhill\_funs(xdata)

downhill\_jackknifecal Jackknife downhill

## Description

It uses the downhill approach to calculate the jackknife abundance and returns the estimated acceleration factor

#### Usage

```
downhill_jackknifecal(xdata, checkid = T, maxorder = dim(xdata)[2] - 2)
```

## Arguments

xdata	original data matrix
checkid	If it is T, then checkident.1 is called and it performs the Fienberg-Rinaldo linear program check for the existence of the estimates
maxorder	Maximum order of models to be included

## Value

the estimated acceleration factor

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

data(Korea)
downhill\_jackknifecal(Korea)

encode\_capture Encode capture history

#### Description

Given a 0/1 capture history, encode it as number that corresponds to the row number of the capture history data set

## Usage

```
encode_capture(z)
```

#### Arguments

Ζ

The capture history to be encoded, as a logical vector or a vector of 0s and 1s

#### Value

The capture history encoded as a number that corresponds to the row number of the capture history data set

#### Examples

```
encode_capture(c(1,0,0,0,0))
encode_capture(c(1,1,1,1,0))
encode_capture(c(T,F,T,F))
```

estimatepopulation Bootstrapping to evaluate confidence intervals using BCa methods

#### Description

This routine implements the bootstrapping and jacknife approach as detailed in Section 3.3 of Chan, Silverman and Vincent (2021). It calls the routine estimatepopulation.0 and so is the preferred routine to be called if a user wishes to estimate the population and obtain BCa confidence intervals.

## Usage

```
estimatepopulation(
   zdat,
   nboot = 1000,
   pthresh = 0.02,
   iseed = 1234,
   alpha = c(0.025, 0.05, 0.1, 0.16, 0.84, 0.9, 0.95, 0.975),
   ...
)
```

## Arguments

zdat	Data matrix with $t + 1$ columns. The first $t$ columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last column is the count of cases with that particular capture history. List names A, B, are constructed if not supplied. Where a capture history is not explicitly listed, it is assumed that it has zero count.
nboot	Number of bootstrap replications.
pthresh	p-value threshold used in estimatepopulation.0.
iseed	seed for initialisation.
alpha	Bootstrap quantiles of interests.
	other arguments which will be passed to estimatepopulation.0

## Value

A list with components as below:

popest point estimate of the total population of the original data set

MSEfit model fitted to the data, in the format described in modelfit

bootreps point estimates of total population sizes from each bootstrap sample

ahat the estimated acceleration factor

BCaquantiles BCa confidence intervals

## References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116**(**535**), 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

DiCiccio, T. J. and Efron, B. (1996). Bootstrap Confidence Intervals. *Statistical Science*, **40**(**3**), 189-228.

```
data(Korea)
zdat=Korea
estimatepopulation(zdat, nboot=10)
```

estimatepopulation.0 Estimate the total population including the dark figure. If the user wishes to find bootstrap confidence intervals then the routine estimatepopulation should be used instead.

## Description

This routine estimates the total population size, which includes the dark figure, together with confidence intervals as specified. It also returns the details of the fitted model. The user can choose whether to fit main effects only, to fit a particular model containing specified two-list parameters, or to choose the model using the stepwise approach described by Chan, Silverman and Vincent (2021).

## Usage

```
estimatepopulation.0(
  zdat,
  method = "stepwise",
  quantiles = c(0.025, 0.975),
  mX = NULL,
  pthresh = 0.02
)
```

#### Arguments

zdat	Data matrix with $t + 1$ columns. The first $t$ columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last column is the count of cases with that particular capture history. List names A, B, are constructed if not supplied. Where a capture history is not explicitly listed, it is assumed that it has zero count.
method	If method = "stepwise" the stepwise method implemented in stepwisefit is used. If method = "fixed" then a specified fixed model is used; the model is then given by mX. If method = "main" then main effects only are fitted.
quantiles	Quantiles of interest for confidence intervals.
mX	A $2 \times k$ matrix giving the k two-list parameters to be included in the model if method = "fixed". Each column of mX contains the numbers of the correspond- ing pair of lists. If mX = 0, then all two-list parameters are included. If mX = NULL, no two-list parameters are included and the main effects model is fitted. If only one two-list parameter is to be fitted, it is sufficient to specify it as a vector of length 2, e.g mX=c(1,3) for the parameter indexed by lists 1 and 3. If method is equal to "stepwise" or "main", then mX is ignored.
pthresh	Threshold p-value used if method = "stepwise".

#### Value

A list with components as below

estimate Point estimate and confidence interval estimates corresponding to specified quantiles. MSEfit The model fitted to the data in the format described in modelfit.

## References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116**(**535**), 1297-1306, Available from https://www.tandfonline.com/ doi/full/10.1080/01621459.2019.1708748.

## Examples

```
data(NewOrl)
data(NewOrl_5)
estimatepopulation.0(NewOrl, method="stepwise", quantiles=c(0.025,0.975))
estimatepopulation.0(NewOrl_5, method="main", quantiles=c(0.01, 0.05,0.95, 0.99))
```

find\_bic\_rank\_matrix Find BIC rank matrix up to a specified number of bic ranks for a given data set

#### Description

Find BIC rank matrix up to a specified number of bic ranks for a given data set

## Usage

```
find_bic_rank_matrix(zsortbic, nbicranks = 5)
```

#### Arguments

zsortbic	Output from applying assemble_bic to the data
nbicranks	The number of bic ranks to be propagated

#### Value

A bic rank matrix encoding how many steps the models considered need to get to the optimal one.

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

#### Examples

```
data(Korea)
zsortbic=assemble_bic(Korea, checkexist=T)
find_bic_rank_matrix(zsortbic, nbicranks=5)
```

## 32

find\_neighbour\_hierarchies

Find all neighbouring hierarchical model to a given one

## Description

Find all neighbouring hierarchical model to a given one

## Usage

```
find_neighbour_hierarchies(
  modelstr,
  nlists = NA,
  keepmaineffects = T,
  maxorder = nlists - 1
)
```

## Arguments

modelstr	Model string written in hierarchical form
nlists	Number of lists.
keepmaineffects	:
	If T, keep the main effects. If F remove.
maxorder	Maximum order of models to be included

## Value

neighbour hierarchical models

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

```
modelstr = "[12,23]"
find_neighbour_hierarchies(modelstr)
```

find\_unique\_patterns Find unique patterns in matrix columns

## Description

Given a matrix (for example of bootstrap replications) construct the matrix of unique patterns of non-zeroes, together with a vector of pointers back to that matrix.

## Usage

```
find_unique_patterns(x)
```

## Arguments

x a matrix

### Value

The original data x with the additional components

yuniq matrix of unique patterns of non-zeroes/zeroes in the columns of x

**pointers** vector of length dim(x)[2] giving the column of yuniq corresponding to each column of x

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

data(Korea)
z=assemble\_bic(Korea)
z=bootstrapcal(z)
find\_unique\_patterns(z\$bootreplications)

fit\_hier\_model

## Description

Fit a hierarchical model taking account of possible sparsity

## Usage

```
fit_hier_model(xdatin, hiermod, bicRcap = T, checkid = F)
```

## Arguments

xdatin	data obtained using ingest_data
hiermod	hierarchical model to fit
bicRcap	if T then use the Rcapture convention that the BIC sample size is the number of cases observed. Otherwise use the number of cells in the Poisson log linear model.
checkid	if T then checkident.1 is called inside the routine

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

```
data(Korea)
xdatin = ingest_data(Korea)
fit_hier_model(xdatin,"[12,23]")
fit_hier_model(xdatin,"[12,3]")
```

gethiermodels	Get a list of all hierarchical models for given number of lists and max-
	imum order

## Description

Extracts from a larger vector of hierarchical models the ones satisfying the given criterion

## Usage

```
gethiermodels(nlists, maxorder = nlists - 1, modelvec = hiermodels)
```

hiermodels

#### Arguments

nlists	Number of lists
maxorder	Maximum order of models to be returned (defaults to nlists-1)
modelvec	vector of hierarchical models (defaults to hiermodels)

## Value

A list of models satisfying the given criteria

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

```
data(hiermodels)
# Five lists with maximum order of 4
gethiermodels(nlists=5,maxorder=4)
# Five lists with maximum order of 2
gethiermodels(nlists=5, maxorder=2)
```

hiermodels Hierarchical models

## Description

Hierarchical models from two lists to six lists

## Usage

hiermodels

## Format

An object of class character of length 39783.

## Details

These data provide hierarchical models constructed from two lists to six lists of maximal order 2.

ingest\_data

## Description

Perform various preprocessing tasks on the data

#### Usage

```
ingest_data(xdat)
```

## Arguments

xdat Data matrix of the usual kind

## Value

A list with the following elements

- nobs Numbers of observations indexed by encoded histories
- nstar For each capture history, total number of observations for that capture history and all its descendants

**nlists** Total number of lists

listnames Names of the lists, constructed to be A, B, ... if necessary

- **data** The input data matrix
- **notestimable** A vector indicating which parameters are not estimable, because they are strict descendants of parameters which would be already estimated to be  $-\infty$  if they are included in the model

masterdesign The inclusion matrix as constructed by make\_master\_design

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

```
#Korea data
data(Korea)
ingest_data(Korea)
#Kosovo data
data(Kosovo)
ingest_data(Kosovo)
```

investigateAIC Plot of simulation study

## Description

This routine reproduces Figure 1 of Chan, Silverman and Vincent (2021).

#### Usage

investigateAIC(nsim = 10000, Nsamp = 1000, seed = 1001)

#### Arguments

nsim	The number of simulation replications
Nsamp	The expected value of the total population size within each simulation
seed	The random number seed

#### Details

Simulations are carried out for two different three-list models. In one model, the probabilities of capture are 0.01, 0.04 and 0.2 for the three lists respectively, while in the other the probability is 0.3 on all three lists. In both cases, captures on the lists occur independently of each other, so there are no two-list effects. The first model is chosen to be somewhat more typical of the sparse capture case, of the kind which often occurs in the human trafficking context, while the second, more reminiscent of a classical mark-recapture study.

The probability of an individual having each possible capture history is first evaluated. Then these probabilities are multiplied by Nsamp = 1000 and, for each simulation replicate, Poisson random values with expectations equal to these values are generated to give a full set of observed capture histories; together with the null capture history the expected number of counts (population size) is equal to Nsamp. Inference was carried out both for the model with main effects only, and for the model with the addition of a correlation effect between the first two lists. The reduction in deviance between the two models was determined. Ten thousand simulations were carried out.

Checking for compliance with the conditions for existence and identifiability of the estimates shows that a very small number of the simulations for the sparse model (two out of ten thousand) fail the checks for existence even within the extended maximum likelihood context. Detailed investigation shows that in neither of these cases is the dark figure itself not estimable; although the parameters themselves cannot all be estimated, there is a maximum likelihood estimate of the expected capture frequencies, and hence the deviance can still be calculated.

The routine produces QQ-plots of the resulting deviance reductions against quantiles of the  $\chi_1^2$  distribution, for nsim simulation replications.

#### Value

An  $nsim \times 2$  matrix giving the changes in deviance for each replication for each of the two models.

## jackknifecal

## References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116**(**535**), 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

jackknifecal

#### Jackknife abundance and Jackknife bic

#### Description

This routine takes the output from subsetmat or from assemble\_bic and returns the jackknife abundance matrix and jackknife BIC matrix.

## Usage

jackknifecal(z, checkexist = T)

#### Arguments

checkexist If checkexist=TRUE, check for existence in cases where the jackknife intro- duces an additional zero, else it does not check for existence. Note that in the current version it is assume that models for which the fit doesn't exist for the original data have already been excluded.	Z	Results from assemble_bic or subsetmat.
	checkexist	If checkexist=TRUE, check for existence in cases where the jackknife intro- duces an additional zero, else it does not check for existence. Note that in the current version it is assume that models for which the fit doesn't exist for the original data have already been excluded.

## Value

A list with the following components

jackabund Jackknife abundance matrix

jackbic Jackknife BIC matrix

countsobserved Capture counts in the same order as the columns of jackabund and jackbic

#### References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

```
data(Korea)
z=assemble_bic(Korea)
jackknifecal(z,checkexist=T)
```

Korea

#### Description

Korean woman held in sexual slavery by the Japanese military

#### Usage

Korea

## Format

An object of class matrix (inherits from array) with 7 rows and 4 columns.

#### Details

These data are collected into three lists. Full details are given in Figure 1 and the Data section of Ball et al. (2018)

## References

Ball, P., Shin, E. H-S. and Yang, H. (2018). There may have been 14 undocumented Korean "comfort women" in Palembang, Indonesia. Technical Report, Humans Rights Data Analysis Group. Available from https://hrdag.org/wp-content/uploads/2018/12/KP-Palemban-ests.pdf

Kosovo

Kosovo data

#### Description

Data on 4400 observed killings in the Kosovo war between 20 March and 22 June 1999

## Usage

Kosovo

#### Format

An object of class data. frame with 15 rows and 5 columns.

## Details

These data give the numbers of cases on each possible combination of four lists. The lists are labelled as follows: EXH = exhumations; ABA = American Bar Association Central and East European Law Initiative; OSCE = Organization for Security and Cooperation in Europe; HRW = Human Rights Watch. All 15 combinations have a nonzero count.

## ktopBCa

## References

Ball, P., W. Betts, F. Scheuren, J. Dudukovich, and J. Asher (2002). Killings and Refugee Flow in Kosovo March-June 1999. American Association for the Advancement of Science. A Report to the International Criminal Tribunal for the Former Yugoslavia.

ktopBCa

Find BCa confidence intervals using ktop idea.

## Description

Find BCa confidence intervals using ktop idea.

## Usage

```
ktopBCa(
    z,
    BICmatrix_break,
    alpha = c(0.025, 0.05, 0.1, 0.16, 0.2, 0.5, 0.8, 0.84, 0.9, 0.95, 0.975),
    maxorder = Inf,
    BCaFD = FALSE
)
```

#### Arguments

Z	The output of assemble_bic, bootstrapcal and jackknifecal
BICmatrix_break	< compared with the second s
	output from BICrank_tiebreak
alpha	Levels of confidence intervals to be constructed and assessed
maxorder	Maximum order of models to be included
BCaFD	If T, return also the probabilities of the estimates

## Value

A list with the following components

confvals BCa confidence intervals for each considered model

probests Corresponding probabilities of the estimates

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

```
data(Korea)
z=assemble_bic(Korea)
z=bootstrapcal(z)
z=jackknifecal(z)
BICmatrix_prop= find_bic_rank_matrix(z)
BICmatrix_break = BICrank_tiebreak(BICmatrix_prop, 2)
ktopBCa(z,BICmatrix_break)
```

make\_master\_design Set up the inclusion matrix for all possible capture histories

#### Description

This is the master design matrix which maps parameters to observations. Rows correspond to observations and columns to parameters.

#### Usage

```
make_master_design(nlists)
```

## Arguments

nlists The number of lists

## Value

A matrix whose (i, j) element is 1 if the expected log of observation *i* depends on parameter *j*, in other words if *j* is an ancestor of *i*.

#### References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

```
#Create master design matrix with 5 lists
make_master_design(5)
#Create master design matrix with 3 lists
make_master_design(3)
```

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modelfit

## Description

This routine fits a specified model to multiple systems estimation data, taking account of the possibility of empty overlaps between pairs of observed lists.

#### Usage

modelfit(zdat, mX = NULL, check = TRUE)

#### Arguments

zdat	Data matrix with $t + 1$ columns. The first $t$ columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last column is the count of cases with that particular capture history. List names A, B, are constructed if not supplied. Where a capture history is not explicitly listed, it is assumed that it has zero count.
mX	A $2 \times k$ matrix giving the k two-list parameters to be included in the model. Each column of mX contains the numbers of the corresponding pair of lists. If mX = 0, then all two-list parameters are included. If mX = NULL, no two-list parameters are included and the main effects model is fitted. If only one two-list parameter is to be fitted, it may be specified as a vector of length 2, e.g mX=c(1,3) for the parameter corresponding to lists 1 and 3.
check	If check = TRUE check first of all if the maximum likelihood estimate exists and is identifiable, using the routine checkident. If either condition fails, print an appropriate error message and return the error code.

## Value

A list with components as below

fit Details of the fit of the specified model as output by glm. The Akaike information criterion is adjusted to take account of the number of parameters corresponding to non-overlapping pairs.

emptyoverlaps Matrix with two rows, giving the list pairs within the model for which no cases are observed in common. Each column gives the indices of a pair of lists, with the names of the lists in the column name.

poisspempty the Poisson p-values of the parameters corresponding to non-overlapping pairs.

```
data(NewOrl)
modelfit(NewOrl,mX= c(1,3), check=TRUE)
```

modelorder

## Description

This routine returns the order of the model given the model character string

#### Usage

```
modelorder(x)
```

## Arguments

х

a model character string

## Value

the order of the model character string

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

```
x="[1,2,3,4,5]"
modelorder(x)
y="[12,24,25,35,45]"
modelorder(y)
```

Ned

The Netherlands data

## Description

Victims related to human trafficking in the Netherlands

## Usage

Ned

## Format

An object of class data. frame with 24 rows and 7 columns.

## $Ned_5$

## Details

These data are collected into six lists. Full details are given in Table 2 of Silverman (2020).

#### References

Silverman, B. W. (2020). Model fitting in Multiple Systems Analysis for the quantification of Modern Slavery: Classical and Bayesian approaches *Journal of Royal Statistical Society: Series A*, **183(3)**, 691-736, Available from https://rss.onlinelibrary.wiley.com/doi/full/10.1111/rssa.12505

Ned\_5

Netherlands data five list version

## Description

Netherlands data consolidated into five lists

#### Usage

Ned\_5

## Format

An object of class data. frame with 17 rows and 6 columns.

## Details

This reduces the Netherlands data Ned into five lists, constructed by combining the two smallest lists I and O into a single list.

New0rl

New Orleans data

#### Description

Victims related to human trafficking in Greater New Orleans

## Usage

New0rl

## Format

An object of class data. frame with 19 rows and 9 columns.

## Details

These data are collected into 8 lists. For reasons of confidentiality the lists are only labelled as A, B, ..., H. Full details are given in Bales, Murphy and Silverman (2020).

## References

K. Bales, L. Murphy and B. W. Silverman (2020). How many trafficked people are there in Greater New Orleans? *Journal of Human Trafficking*, **6**(**4**), 375-384, available from https://www.tandfonline.com/doi/full/10.1080/23322705.2019.1634936

NewOrl\_5

New Orleans data five list version

#### Description

New Orleans data consolidated into five lists

## Usage

NewOrl\_5

## Format

An object of class data. frame with 14 rows and 6 columns.

#### Details

This reduces the New Orleans data NewOrl into five lists, constructed by combining the four smallest lists B, E, F and G into a single list.

ntopBCa

Find BCa confidence intervals for all possible ntop

## Description

Find BCa confidence intervals for all possible ntop

## Usage

```
ntopBCa(
    z,
    alpha = c(0.025, 0.05, 0.1, 0.16, 0.2, 0.5, 0.8, 0.84, 0.9, 0.95, 0.975),
    maxorder = Inf,
    BCaFD = FALSE
)
```

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#### ordercaptures

#### Arguments

Z	The output of $\verb"assemble_bic", \verb"bootstrapcal" and jackknifecal$
alpha	Levels of confidence intervals to be constructed and assessed
maxorder	Maximum order of models to be included
BCaFD	If T, return also the probabilities of the estimates

## Value

A list with the following components

confvals BCa confidence intervals for each considered model

probests Corresponding probabilities of the estimates

## References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/ 10.1007/s11222-023-10346-9.

## Examples

```
data(Korea)
z=assemble_bic(Korea, checkexist=T)
z=bootstrapcal(z, checkexist=T)
z=jackknifecal(z,checkexist=T)
ntopBCa(z)
```

ordercaptures Order capture histories

## Description

Given a matrix with capture histories only, the routine orders the capture histories first by the number of 1s in the capture history and then lexicographically by columns.

## Usage

```
ordercaptures(zmat)
```

#### Arguments

zmat

Data matrix with t columns. The t columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. Where a capture history is not explicitly listed, it is assumed that it has observed count zero.

#### Value

A data matrix that is ordered first by the number of 1s in the capture history and then lexicographically by columns.

## References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116**(**535**), 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

## Examples

```
#UK 5 list data
data(UKdat_5)
# Obtain matrix with capture histories only
UKdat_5_cap <- UKdat_5[, 1:dim(UKdat_5)[2]-1]
ordercaptures(UKdat_5_cap)
```

parent\_captures Find the "parents" of a given capture history

#### Description

Given any encoded capture history and the number of lists, find the encoded capture histories which are obtained by leaving out just one list in turn

#### Usage

```
parent_captures(k, nlists = 10)
```

## Arguments

k	An encoded capture history that corresponds to the row number of the capture
	history data set
nlists	The total number of lists

#### Value

a vector giving the encoded versions of the parents

#### References

Silverman, B. W., Chan, L. and Vincent, K., (2024). Bootstrapping Multiple Systems Estimates to Account for Model Selection *Statistics and Computing*, **34**(**44**), Available from https://doi.org/10.1007/s11222-023-10346-9.

## Examples

parent\_captures(2,10)
parent\_captures(1,4)

removenoninformativelists

Remove non-informative list

## Description

The routine cleans up the data set by removing any lists that contain no data, any lists which contain all the observed data, and any list whose results duplicate those of another list. If as a result the original data set has no list left, it returns a matrix with the value of the total count.

#### Usage

removenoninformativelists(zdat)

#### Arguments

zdat Data matrix with t + 1 columns. The first t columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last column is the count of cases with that particular capture history.

## Value

data matrix that contains no duplicate lists, no lists with no data, and no lists that contain all the observed data. If all lists are removed, the total count is returned.

## References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116**(**535**), 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

sortmodelsbic

## Description

This routine sorts the models in increasing order according to their BICs, returns the sorted models with their corresponding BICs and abundance. The original data as well as the maxorder of the models considered are returned as well.

## Usage

```
sortmodelsbic(
  xdata,
  maxorder = dim(xdata)[2] - 2,
  checkexist = T,
  removeFRfail = T,
  ...
)
```

#### Arguments

xdata	The original data matrix with capture histories and counts.
maxorder	Maximum order of models to be included
checkexist	If T then the Fienberg-Rinaldo condition is checked for each model
removeFRfail	If checkexist is T then models which fail the FR condition are removed from the results
	Parameters to be fed to gethiermodels.

## Value

A list with the following components

- **res** A matrix with the models considered, their abundance, BIC and their order, sorted into increasing order of BIC
- xdata The original data matrix with capture histories and counts.

maxorder Order parameter that was feed into gethiermodels

```
data(hiermodels)
data(Korea)
sortmodelsbic(Korea, checkexist=T)
```

stepwisefit

#### Description

Starting with a model with main effects only, two-list parameters are added one by one. At each stage the parameter with the lowest p-value is added, provided that p-value is lower than pthresh, and provided that the resulting model does not fail either of the tests in checkident.

#### Usage

stepwisefit(zdat, pthresh = 0.02)

## Arguments

zdat	Data matrix with $t + 1$ columns. The first t columns, each corresponding to a
	particular list, are 0s and 1s defining the capture histories observed. The last
	column is the count of cases with that particular capture history. List names A,
	B, are constructed if not supplied. Where a capture history is not explicitly
	listed, it is assumed that it has zero count.
pthresh	this is the threshold below which the p-value of the newly added parameter needs to be in order to be included in the model. If pthresh = 0 then the model with main effects only is returned.

#### Details

For each candidate two-list parameter for possible addition to the model, the p-value is calculated as follows. The total of cases occurring on both lists indexed by the parameter (regardless of whether or not they are on any other lists) is calculated. On the null hypothesis that the effect is not included in the model, this statistic has a Poisson distribution whose mean depends on the parameters within the model. The one-sided Poisson p-value of the observed statistic is calculated.

#### Value

A list with components as below

fit Details of the fit of the specified model as output by glm. The Akaike information criterion is adjusted to take account of the number of parameters corresponding to non-overlapping pairs.

emptyoverlaps Matrix with two rows, giving the list pairs within the model for which no cases are observed in common. Each column gives the indices of a pair of lists, with the names of the lists in the column name.

poisspempty the Poisson p-values of the parameters corresponding to non-overlapping pairs.

```
data(NewOrl)
stepwisefit(NewOrl, pthresh=0.02)
```

subsetmat

#### Description

The idea of this routine is to reduce either or both of ntopmodels and maxorder without the need to recalculate any actual model fits.

#### Usage

subsetmat(z, ntopmodels = Inf, maxorder = Inf)

#### Arguments

Z	output from assemble_bic or a list output from applying assemble_bic, jackknifecal and bootstrapcal.
ntopmodels	number of top models. If (taking into account any change in the maximum order of models) there are fewer than ntopmodels in the data supplied, then it will be reduced to that value. If it is not specified then there will be no reduction in the number.
maxorder	the maximum order of the models to be considered. If not specified, it will be set to the corresponding value in the input data.

#### **Details**

The routine subsets the results matrix as part of the output given by assemble\_bic based on specified parameters ntopmodels and maxorder. It returns the subsetted matrix, original data matrix with capture histories and counts and the new actual value of maxorder (reducing it from the input value if necessary or if the default input value of  $\infty$  is used).

#### Value

A list with the following components

- **res** a matrix containing models being considered, abundance, BIC and their ordered after being subsetted by maxorder and ntopmodels
- xdata Original data matrix with counts and capture histories

maxorder The maximum order of models considered after subsetting

jackabund Jackknife abundance matrix, subsetted by maxorder and ntopmodels

jackbic Jackknife BIC matrix, subsetted by maxorder and ntopmodels

countsobserved Capture counts in the same order as the columns of jackabund and jackbic

bootabund Bootstrap abundance matrix, subsetted by maxorder and ntopmodels

bootbic Bootstrap BIC matrix, subsetted by maxorder and ntopmodels

If the input only has the output from assemble\_bic, the last five items of the list do not appear.

#### subsetsearch

## Examples

```
data(UKdat_5)
#Example 1
z=assemble_bic(UKdat_5)
subsetmat(z,ntopmodels=100,maxorder=2)
subsetmat(z)
#Example 2
z=assemble_bic(UKdat_5)
z=jackknifecal(z)
z=bootstrapcal(z)
subsetmat(z,ntopmodels=100,maxorder=2)
```

subsetsearch

Search subsets for a property which is inherited in a particular way

### Description

The following routine returns a list of all subsets of a given set for which a specified property is TRUE. It is assumed that if the property is FALSE for any particular subset, it is also FALSE for all supersets of that subset. This enables a branch search strategy to be used to obviate the need to search over supersets of subsets already eliminated from consideration. It is used within the hierarchical search step of checkallmodels.

## Usage

```
subsetsearch(n, checkfun, testnull = TRUE, ...)
```

## Arguments

n	an integer such that the search is over all subsets of $\{1,, n\}$
checkfun	a function which takes arguments zset, a subset of $\{1,, n\}$ and The function returns the value TRUE or FALSE. It needs to have the property that if it is FALSE for any particular zset, it is also FALSE for all supersets of zset.
testnull	If TRUE, then include the null subset in the search. If FALSE, do not test the null subset.
	other arguments

#### Value

A list of all subsets for which the value is TRUE

## References

Chan, L., Silverman, B. W., and Vincent, K. (2021). Multiple Systems Estimation for Sparse Capture Data: Inferential Challenges when there are Non-Overlapping Lists. *Journal of American Statistical Association*, **116**(**535**), 1297-1306, Available from https://www.tandfonline.com/doi/full/10.1080/01621459.2019.1708748.

```
tidylists
```

## Description

This routine finds rows with the same capture history and consolidates them into a single row whose count is the sum of counts of the relevant rows. If includezerocounts = TRUE then it also includes rows for all the capture histories with zero count; otherwise these are all removed.

#### Usage

```
tidylists(zdat, includezerocounts = FALSE)
```

## Arguments zdat

Data matrix with t + 1 columns. The first t columns, each corresponding to a particular list, are 0s and 1s defining the capture histories observed. The last column is the count of cases with that particular capture history. List names A, B, ... are constructed if not supplied. Where a capture history is not explicitly listed, it is assumed that it has zero count.

includezerocounts

If FALSE then remove rows corresponding to capture histories with zero count. If TRUE then include all possible capture histories including those with zero count, excluding the all-zero row corresponding to the dark figure.

## Value

A data matrix in the form specified above, including all capture histories with zero counts if includezerocounts=TRUE.

## Examples

data(NewOrl)
tidylists(NewOrl,includezerocounts=TRUE)

UKdat

UK data

## Description

Data from the UK 2013 strategic assessment

#### Usage

UKdat

## UKdat\_5

## Format

An object of class data. frame with 25 rows and 7 columns.

## Details

This is a table of six lists used in the research published by the Home Office as part of the strategy leading to the Modern Slavery Act 2015. The data are considered in six lists, labelled as follows: LA–Local authorities; NG–Non-government organisations; PF–Police forces; GO–Government organisations; GP–General public; NCA–National Crime Agency. Each of the first six columns in the data frame corresponds to one of these lists. Each of the rows of the data frame corresponds to a possible combination of lists, with value 1 in the relevant column if the list is in that particular combination. The last column of the data frame states the number of cases observed in that particular combination of lists. Combinations of lists for which zero cases are observed are omitted.

#### References

https://www.gov.uk/government/publications/modern-slavery-an-application-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-of-multiple-systems-estimation-

UKdat\_5

UK data five list version

#### Description

UK data consolidated into five lists

#### Usage

UKdat\_5

## Format

An object of class data. frame with 18 rows and 6 columns.

## Details

This reduces the UK data UKdat into five lists, constructed by combining the PF and NCA lists into a single PFNCA list

## Western

## Description

These data are collected into 5 lists. For reasons of confidentiality the lists are only labelled as A, B, C, D and E. Full details are given in Farrell, Dank, Kfafian, Lockwood, Pfeffer, Hughes and Vincent (2019).

## Usage

Western

## Format

An object of class data. frame with 13 rows and 6 columns.

#### References

Farrell, A., Dank, M., Kfafian, M., Lockwood, S., Pfeffer, R., Hughes, A., and Vincent, K. (2019). Capturing human trafficking victimization through crime reporting. Technical Report 2015-VF-GX-0105, National Institute of Justice. Available from https://www.ncjrs.gov/pdffiles1/nij/grants/252520.pdf.

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