# Package 'CompositeRegressionEstimation'

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### Description

ΑK

Consider a sequence of monthly samples  $(S_m)_{m\in\{1,\dots,M\}}$ . In the CPS, a sample  $S_m$  is the union of 8 rotation groups:  $S_m = S_{m,1} \cup S_{m,2} \cup S_{m,3} \cup S_{m,4} \cup S_{m,5} \cup S_{m,6} \cup S_{m,7} \cup S_{m,8}$ , where two consecutive samples are always such that  $S_{m,2} = S_{m-1,1}, S_{m,3} = S_{m-1,2}, S_{m,4} = S_{m-1,3}, S_{m,6} = S_{m-1,5}, S_{m,7} = S_{m-1,6}, S_{m,8} = S_{m-1,7}$ , and one year appart samples are always such that  $S_{m,5} = S_{m-12,1}, S_{m,6} = S_{m-12,2}, S_{m,7} = S_{m-12,3}, S_{m,8} = S_{m-12,4}$ .

AK Estimator (recursive version)

The subsamples  $S_{m,g}$  are called rotation groups, and rotation patterns different than the CPS rotation pattern are possible.

For each individual k of the sample m, one observes the employment status  $Y_{k,m}$  (A binary variable) of individual k at time m, and the survey weight  $w_{k,m}$ , as well as its "rotation group".

The AK composite estimator is defined in "CPS Technical Paper (2006), [section 10-11]":

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For m = 1,  $\hat{t}_{Y_{.,1}} = \sum_{k \in S_1} w_{k,m} Y_{k,m}$ . For m > 2,

$$\hat{t}_{Y_{.,m}} = (1 - K) \times \left( \sum_{k \in S_m} w_{k,m} Y_{k,m} \right) + K \times (\hat{t}_{Y_{.,m-1}} + \Delta_m) + A \times \hat{\beta}_m$$

where

$$\Delta_m = \eta_0 \times \sum_{k \in S_m \cap S_{m-1}} (w_{k,m} Y_{k,m} - w_{k,m-1} Y_{k,m-1})$$

and

$$\hat{\beta}_m = \left(\sum_{k \notin S_m \cap S_{m-1}} w_{k,m} Y_{k,m}\right) - \eta_1 \times \left(\sum_{k \in S_m \cap S_{m-1}} w_{k,m} Y_{k,m}\right)$$

For the CPS,  $\eta_0$  is the ratio between the number of rotation groups in the sample and the number of overlaping rotation groups between two month, which is a constant  $\eta_0 = 4/3$ ;  $\eta_1$  is the ratio between the number of non overlaping rotation groups the number of overlaping rotation groups between two month, which is a constant of 1/3.

In the case of the CPS, the rotation group one sample unit belongs to in a particular month is a function of the number of times it has been selected before, including this month, and so the rotation group of an individual in a particular month is called the "month in sample" variable.

For the CPS, in month m the overlap  $S_{m-1} \cap S_m$  correspond to the individuals in the sample  $S_m$  with a value of month in sample equal to 2,3,4, 6,7 or 8. The overlap  $S_{m-1} \cap S_m$  correspond to the individuals in the sample  $S_m$  with a value of month in sample equal to 2,3,4, 6,7 or 8. as well as individuals in the sample  $S_{m-1}$  with a value of month in sample equal to 1,2,3, 5,6 or 7. When parametrising the function, the choice would be group\_1=c(1:3,5:7) and group0=c(2:4,6:8).

Computing the estimators recursively is not very efficient. At the end, we get a linear combinaison of month in sample estimates The functions AK3, and WSrg computes the linear combination directly and more efficiently.

#### Usage

```
AK(
list.tables,
w,
list.y,
id = NULL,
groupvar = NULL,
groups_1 = NULL,
groups0 = NULL,
A = 0,
K = 0,
dft0.y = NULL,
eta0 = 0,
eta1 = 0
)
```

AK

#### **Arguments**

list.tables	a list of tables
W	a character string: name of the weights variable (should be the same in all tables)
list.y	a vector of variable names
id	a character string: name of the identifier variable (should be the same in all tables)
groupvar	a character string: name of the rotation group variable (should be the same in all tables)
groups_1	a character string:
groups0	if groupvar is not null, a vector of possible values for L[[groupvar]]
eta0	a numeric value
eta1	a numeric value

#### **Details**

the function is based on the more general function CompositeRegressionEstimation::composite

#### References

"CPS Technical Paper (2006). Design and Methodology of the Current Population Survey. Technical Report 66, U.S. Census Bureau.", "Gurney, M. and Daly, J. F. (1965). A multivariate approach to estimation in periodic sample surveys. In Proceedings of the Social Statistics Section, American Statistical Association, volume 242, page 257."

### See Also

CompositeRegressionEstimation::composite

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AK\_est

AK estimation on array of month in sample estimates

### Description

AK estimation on array of month in sample estimates

### Usage

```
AK_est(
   Y,
   month = names(dimnames(Y))[1],
   group = names(dimnames(Y))[2],
   variable = names(dimnames(Y))[3],
   S,
   S_1 = S - 1,
   a,
   k,
   groups = dimnames(Y)[[group]],
   eta0 = length(groups)/length(S),
   eta1 = eta0 - 1
)
```

### Arguments

Υ	an array of named dimensions with 3 dimensions: 1 for the month, 1 for the month in sample, 1 for the variable name
month	: name of the month dimension (by default the name of the first dimension of Y names(dimnames(dim(Y)))[1])
group	: name of the group dimenstion of $Y$ (by default the name of the second dimension of $Y$ names(dim(names(dim(Y)))[2])
S	a vector of integers, subvector of 1:ngroup, to be passed to W.ak, indicating the rotation group numbers this month that were present the previous months (for CPS, $c(2:4,6:8)$ )
a	a numeric value
k	a numeric value
eta0	a numeric value to be passed to W.ak
eta1	a numeric value to be passed to W.ak

#### Value

an array

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#### **Examples**

```
library(dataCPS)
period=200501:200512
list.tables<-lapply(data(list=paste0("cps",period),package="dataCPS"),get);
names(list.tables)<-period
Y<-WSrg(list.tables,weight="pwsswgt",list.y="pemlr",rg="hrmis")
dimnames(Y);month="m";group="mis";variable="y";
S=c(2:4,6:8);eta0=1.3;eta1=.3
A=W.ak(months = dimnames(Y)[[month]],groups = dimnames(Y)[[group]],S=c(2:4,6:8),a=.5,k=.5,eta0,eta1)
a=.5;
k=.5;
ngroup=dim(Y)[group];eta0=ngroup/length(S);eta1=eta0-1;
AK_est(Y,month="m",group="mis",a=.5,k=.6,eta0,eta1)</pre>
```

CoeffS2

Compute the coefficients for Multivariate Blue

#### **Description**

Compute the coefficients for Multivariate Blue

#### Usage

```
CoeffS2(nmonth)
```

#### **Arguments**

Sigma a p x p matrix
X an n x p matrix

Xplus: a general inverse of X

#### Value

the coefficients matrix \$W\$ such that \$WY\$ is the best unbiased linear estimator of \$\beta\$ where \$E[Y]=X\beta\$

```
A=array(rnorm(prod(2:5)),2:5); M=a2m(A,2); dim(A); dim(M); dim(a2m(A))
```

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CoeffYF

Compute Yansaneh-Fuller coefficient for CPS, matrix version

#### **Description**

Compute Yansaneh-Fuller coefficient for CPS, matrix version

### Usage

```
CoeffYF(Sigma, nmonth = dim(Sigma)[[1]])
```

#### **Arguments**

Sigma

a Variance covariance array

#### Value

a matrix.

#### **Examples**

CoeffYF(var())

CoeffYF.matrix

Compute the coefficients for Multivariate Blue

#### **Description**

Compute the coefficients for Multivariate Blue

### Usage

```
CoeffYF.matrix(Sigma, X, Xplus = MASS::ginv(X))
```

### Arguments

Sigma a p x p matrix
X an n x p matrix
Xplus: a general inverse of X

#### Value

the coefficients matrix  $W\$  such that  $WY\$  is the best unbiased linear estimator of  $\theta\$  where  $E[Y]=X\$ 

```
A=array(rnorm(prod(2:5)),2:5); \\ M=a2m(A,2); \\ dim(A); \\ dim(M); \\ dim(a2m(A))
```

8 composite

composite

Linear Composite Estimator from overlap and non overlapping consecutive subsamples direct totals

#### **Description**

Consider a sequence of monthly samples  $(S_m)_{m \in \{1,\dots,M\}}$ . For each individual k of the sample m, one observes the employment status  $Y_{k,m}$  (A binary variable) of individual k at time m, and the survey weight  $w_{k,m}$ . The following program allows to compute recursively for  $m=1,\dots,M$  the Census composite estimator of the total of  $Y_{...m}$  with coefficients defined recursively as follows:

For 
$$m = 1$$
,  $\hat{t}_{Y_{.,1}} = \sum_{k \in S_1} w_{k,m} Y_{k,m}$ .

For m > 2,

$$\hat{t}_{Y,m} = \begin{bmatrix} \hat{t}_{Y,m-1} \\ \sum_{k \in S_m} w_{k,m} Y_{k,m} \\ \sum_{k \in S_{m-1} \cap S_m} w_{k,m-1} Y_{k,m-1} \\ \sum_{k \in S_{m-1} \cap S_m} w_{k,m} Y_{k,m} \\ \sum_{k \in S_m \setminus S_{m-1}} w_{k,m} Y_{k,m} \end{bmatrix}^{T} \times \begin{bmatrix} \alpha_{(-1)} \\ \alpha_{0} \\ \beta_{(-1)} \\ \beta_{0} \\ \gamma_{0} \end{bmatrix}$$

This function computes the estimators for given values of  $\alpha, \beta, \gamma$ .

An example of use of such estimate is the Census Bureau AK estimator: it is a special case of this estimator, with the values of  $\alpha$ ,  $\beta$ ,  $\gamma$  that are given as a function of two parameters A and K:

$$\begin{bmatrix} \alpha_{(-1)} \\ \alpha_0 \\ \beta_{(-1)} \\ \beta_0 \\ \gamma_0 \end{bmatrix} = \begin{bmatrix} K \\ 1 - K \\ -4 K/3 \\ (4K - A)/3 \\ A \end{bmatrix}$$

for more references, please refer to the function CompositeRegressionEstimation::AK.

See "CPS Technical Paper (2006). Design and Methodology of the Current Population Survey. Technical Report 66, U.S. Census Bureau."

$$\begin{array}{lll} \hat{t}_{Y_{.,m}} = & K & \times \hat{t}_{Y_{.,m-1}} \\ & + & (1-K) & \times \sum_{k \in S_m} w_{k,m} Y_{k,m} \\ & + & (-4K/3) & \times \sum_{k \in S_{m-1} \cap S_m} w_{k,m-1} Y_{k,m-1} \\ & + & (4K-A)/3 & \times \sum_{k \in S_{m-1} \cap S_m} w_{k,m} Y_{k,m} \\ & + & A & \times \sum_{k \in S_m \backslash S_{m-1}} w_{k,m} Y_{k,m} \end{array}$$

Computing the estimators recursively is not very efficient. At the end, we get a linear combinaison of month in sample estimates The functions AK3, and WSrg computes the linear combination directly and more efficiently.

For the CPS, in month m the overlap  $S_{m-1} \cap S_m$  correspond to the individuals in the sample  $S_m$  with a value of month in sample equal to 2,3,4, 6,7 or 8. The overlap  $S_{m-1} \cap S_m$  correspond to the individuals in the sample  $S_m$  with a value of month in sample equal to 2,3,4, 6,7 or 8. as well as individuals in the sample  $S_{m-1}$  with a value of month in sample equal to 1,2,3, 5,6 or 7. When parametrising the function, the choice would be group\_1=c(1:3,5:7) and group0=c(2:4,6:8).

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#### Usage

```
composite(
  list.tables,
  w,
  list.y,
  id = NULL,
  groupvar = NULL,
  groups_1 = NULL,
  groups0 = NULL,
  Coef = c(alpha_1 = 0, alpha0 = 1, beta_1 = 0, beta0 = 0, gamma0 = 0),
  dft0.y = NULL
)
```

#### **Arguments**

list.tables	a list of tables
W	a character string: name of the weights variable (should be the same in all tables)
list.y	a vector of variable names
id	a character string: name of the identifier variable (should be the same in all tables)
groupvar	a character string: name of the rotation group variable (should be the same in all tables)
groups_1	a character string:
groups0	if groupvar is not null, a vector of possible values for L[[groupvar]]

#### See Also

CompositeRegressionEstimation::AK

CPS\_AK

Gives A,K coefficient for unemployed used by the Census

#### **Description**

Gives A,K coefficient for unemployed used by the Census

#### Usage

```
CPS_AK()
```

#### Value

```
The vector c(a1=CPS_A_u(),a2=CPS_A_e(),a3=0,k1=CPS_K_u(),k2=CPS_K_e(),k3=0)
```

CPS\_AK\_coeff.array.fl Empirical variance of a collection of arrays.

### Description

Empirical variance of a collection of arrays.

#### Usage

```
CPS_AK_coeff.array.fl(
  nmonth,
  ak = list(c(a_1 = 0, a_2 = 0, a_3 = 0, k_1 = 0, k_2 = 0, k_3 = 0)),
  simplify = TRUE,
  statuslabel = c("0", "1", "_1")
)
```

#### **Arguments**

nmonth a strictly positive integer

ak, a list of numeric vectors of length 6.

simplify a boolean

statuslabel : a character vector of dimension 3 indicating the label for unemployed, em-

ployed, not in the labor force.

```
CPS_AK_coeff.array.fl()
```

CPS\_AK\_est

CPS AK est	Gives the variance of the AK estimators from the A,K coefficients and
CI 3_AIX_C3 C	
	the variance covariance matrix of the month in sample estimates

#### **Description**

Gives the variance of the AK estimators from the A,K coefficients and the variance covariance matrix of the month in sample estimates

### Usage

```
CPS_AK_est(
  mistotals,
  coeff = CPS_AK_coeff.array.fl(dim(mistotals)[1], ak, simplify = FALSE),
  ak = CPS_AK()
)
```

#### **Arguments**

mistotals An array of dimension nmonth x 8 x 3. mistotals[i,j,k] is the month in sample

direct estimate for month i, month in sample j rotation group, and variable k.

coeff An array of coefficients W[ak,y2,m2,y1,mis1,m1] such that AK estimate for co-

efficients ak, month m2 and employment status y2 is sum(W[ak,y2,m2,,,])\*Y[,,]) where mistotals[y1,mis1,m1] is direct estimate on mis mis1 for emp stat y1 at

month m1.

ak: an ak coefficients vector or a list of ak coefficients.

#### Value

The variance of the AK estimators from the A,K coefficients and the variance covariance matrix .

### Examples

CPS\_A\_e

Gives K coefficient for unemployed used by the Census

#### **Description**

Gives K coefficient for unemployed used by the Census

#### Usage

```
CPS_A_e()
```

CPS\_K\_u

### Value

.4

 $\mathsf{CPS}\_\mathsf{A}\_\mathsf{u}$ 

Gives K coefficient for unemployed used by the Census

### Description

Gives K coefficient for unemployed used by the Census

### Usage

```
CPS_A_u()
```

#### Value

.3

CPS\_K\_e

Gives K coefficient for unemployed used by the Census

### Description

Gives K coefficient for unemployed used by the Census

### Usage

```
CPS_K_e()
```

#### Value

.7

CPS\_K\_u

Gives K coefficient for unemployed used by the Census

### Description

Gives K coefficient for unemployed used by the Census

### Usage

#### Value

.4

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CPS\_Xplus\_array

Compute the Moore penrose general inverse of a the Yansaneh Fuller X matrix for CPS, array version

#### **Description**

Compute the Moore penrose general inverse of a the Yansaneh Fuller X matrix for CPS, array version

#### Usage

```
CPS_Xplus_array(X)
```

#### **Arguments**

nmonth

an integer

#### Value

a matrix.

### **Examples**

```
CPS_Xplus_matrix(10)
```

CPS\_Xplus\_matrix

Compute the Moore penrose general inverse of a the Yansaneh Fuller X matrix for CPS

### Description

Compute the Moore penrose general inverse of a the Yansaneh Fuller X matrix for CPS

#### Usage

```
CPS_Xplus_matrix(X)
```

#### **Arguments**

nmonth

an integer

#### Value

a matrix.

```
CPS_Xplus_matrix(10)
```

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CPS\_X\_array

Compute Yansaneh-Fuller X matrix for CPS, array version

### Description

Compute Yansaneh-Fuller X matrix for CPS, array version

### Usage

```
CPS_X_array(nmonth)
```

#### **Arguments**

nmonth

an integer

### Value

a matrix.

### **Examples**

```
CPS_X_matrix(10)
```

 ${\tt CPS\_X\_matrix}$ 

Compute Yansaneh-Fuller X matrix for CPS

### Description

Compute Yansaneh-Fuller X matrix for CPS

### Usage

```
CPS_X_matrix(nmonth)
```

### Arguments

nmonth

an integer

#### Value

a matrix.

```
CPS_X_matrix(10)
```

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douuble

Compute weighted sums

### Description

Compute weighted sums

#### Usage

```
douuble(list.tables, w, id, y)
```

#### **Arguments**

list.tables A list of dataframes, order matters.

w either a real number of a character string indicating the name of the weight

variable.

id primary key of the tables, used to merge tables together.

y a string indicating the name of a factor variable common to all tables of list.tables.

#### Value

a list of three arrays.

#### **Examples**

```
\label{list.tables=lapply(1:10,function(x){cbind(id=1:nrow(Orange),Orange)[sample(nrow(Orange),30),]}), w="ciral content for the content for
```

empirical.var

Empirical variance of a collection of arrays.

### Description

Empirical variance of a collection of arrays.

### Usage

```
empirical.var(A, MARGIN, n)
```

#### **Arguments**

A An array of dimension d\_1 x ... d\_p

MARGIN a vector of integers

the array of dimension a\_1 x ... x a\_n  $Y[i_1,...,i_n]=sum(W[i_1,...,i_n,...,])$ 

```
empirical.var()
```

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factorisedf

Convert variables to numeric in dataframe.

### Description

Convert variables to numeric in dataframe.

#### Usage

```
factorisedf(dfr, list.y)
```

#### **Arguments**

dfr

A dataframe

list.y

character vector containing the names of the variables to be converted.

#### Value

a dataframe

#### **Examples**

```
factorisedf(Orange,names(Orange))
```

MR

Regression Composite estimation

### Description

Regression Composite estimation

### Usage

```
MR(
   list.tables,
   w,
   id,
   list.xMR = NULL,
   list.x1 = NULL,
   list.x2 = NULL,
   list.y = NULL,
   calibmethod = "linear",
   Alpha = 0.75,
   theta = 3/4,
   list.dft.x2 = NULL,
   dft0.xMR = NULL,
```

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```
mu0 = NULL,
Singh = TRUE,
dispweight = FALSE,
analyse = FALSE
)
```

### Arguments

list.tables	A list of dataframes
113t.table3	A list of dataframes
W	either a real number of a character string indicating the name of the weight variable.
id	an identifier
list.xMR	list of variables used to compute proxy composite regression variable
list.x1	list of auxiliary variables used in the cablibration, whose calibrated weighted total has to be equal to initially weithed total
list.x2	id list of auxiliary variables used in the cablibration, whose calibrated weighted total has to be equal to initially weithed total
Alpha	a vector of alpha values. if alpha="01", this will compute MR3
theta	a numerical value
list.dft.x2	id list of auxiliary variables used in the cablibration, whose calibrated weighted total has to be equal to initially weithed total
mu0	a numerical value
Singh	a boolean
dispweight	a boolean
analyse	a boolean
list.y:	list of variables whose weighted sum needs to be computed. It can be factor or character variables.

### Value

a dataframe.

```
\label{list.tables} \mbox{\tt MR(list.tables<-plyr::dlply(CRE\_data,.variables="time),w="Sampling.weight",list.xMR="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Status",id="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Identifier",list.xml="Ident
```

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Gives the variance of the AK estimators from the A,K coefficients and the variance covariance matrix of the month in sample estimates

### Description

Gives the variance of the AK estimators from the A,K coefficients and the variance covariance matrix of the month in sample estimates

#### Usage

```
varAK3(ak, Sigma)
```

#### **Arguments**

ak	A set of 3 A, K coefficients, of the form c(a1=.3,a2=.4,a3=0,k1=.4,k2=.7,k3=0).
Sigma	An array of dimension 3 x 8 (number of rotation groups) x number of months x
	3 x 8 (number of rotation groups) x number of months.

#### Value

The variance of the AK estimators from the A,K coefficients and the variance covariance matrix .

#### **Examples**

```
varAK3(ak=c(a1=.3,a2=.4,a3=0,k1=.4,k2=.7,k3=0), Sigma=array(drop(stats::rWishart(1,df=3*10*8,diag(3*10*8))),report varAK3diff

Gives the variance of the consecutive differences of AK estimators from the A,K coefficients and the variance covariance matrix of the month in sample estimates
```

### Description

Gives the variance of the consecutive differences of AK estimators from the A,K coefficients and the variance covariance matrix of the month in sample estimates

#### Usage

```
varAK3diff(ak, Sigma)
```

#### **Arguments**

ak	A set of 3 A, K coefficients, of the form c(a1=.3,a2=.4,a3=0,k1=.4,k2=.7,k3=0).
Sigma	An array of dimension 3 x 8 (number of rotation groups) x number of months x
	3 x 8 (number of rotation groups) x number of months.

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#### Value

The variance of the consecutive differences of the AK estimators from the A,K coefficients and the variance covariance matrix.

#### **Examples**

```
 varAK3diff(ak=c(a1=.3,a2=.4,a3=0,k1=.4,k2=.7,k3=0), \ Sigma=array(drop(stats::rWishart(1,df=3*10*8,diag(3*10*8))), \ add(10, 1)
```

varAK3diffrat Gives the variance of the unemployment rate estimates derived from

AK estimators from the A,K coefficients and the variance covariance

matrix of the month in sample estimates

#### Description

Gives the variance of the unemployment rate estimates derived from AK estimators from the A,K coefficients and the variance covariance matrix of the month in sample estimates

### Usage

```
varAK3diffrat(ak, Sigma, Scomppop, what = c(unemployed = "0", employed = "1"))
```

#### **Arguments**

ak A set of 3 A, K coefficients, of the form c(a1=.3,a2=.4,a3=0,k1=.4,k2=.7,k3=0).

Sigma An array of dimension 3 x 8 (number of rotation groups) x number of months x

3 x 8 (number of rotation groups) x number of months.

Scomppop An array of dimension number of months x 3.

#### Value

The variance of the the unemployment rate estimates derived from the AK estimators from the A,K coefficients and the variance covariance matrix .

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erived from
covariance

### Description

Gives the variance of the unemployment rate estimates derived from AK estimators from the A,K coefficients and the variance covariance matrix of the month in sample estimates

#### Usage

```
varAK3rat(ak, Sigma, Scomppop, what = c(unemployed = "0", employed = "1"))
```

#### **Arguments**

ak	A set of 3 A, K coefficients, of the form c(a1=.3,a2=.4,a3=0,k1=.4,k2=.7,k3=0).
Sigma	An array of dimension 3 x 8 (number of rotation groups) x number of months x 3 x 8 (number of rotation groups) x number of months.
Scomppop	An array of dimension number of months x 3.

#### Value

The variance of the the unemployment rate estimates derived from the AK estimators from the A,K coefficients and the variance covariance matrix .

### Examples

```
varAK3rat(ak=c(a1=.3,a2=.4,a3=0,k1=.4,k2=.7,k3=0), Sigma=array(drop(stats::rWishart(1,df=3*10*8,diag(3*10*8)))
```

var lin	Gives the variance of an array Y that is a linear transformation AX of
vai _IIII	Gives the variance of an array I that is a tinear transformation AX of
	an array X from the coefficients of A and Sigma=Var[X]

### Description

Gives the variance of an array Y that is a linear transformation AX of an array X from the coefficients of A and Sigma=Var[X]

#### Usage

```
var_lin(A, Sigma)
```

W.ak

#### **Arguments**

```
Sigma An array of dimension b_1 x \dots x b_p x b_1 x \dots x b_p coeff An array of dimension a_1 x \dots x a_n x b_1 x \dots x b_p
```

#### Value

The variance of the AK estimators from the A,K coefficients and the variance covariance matrix .

### **Examples**

```
 a=c(2,4); b=c(3,10,8); A<-array(rnorm(prod(a)*prod(b)), c(a,b)); \\ dimnames(A)[1:2]<-lapply(a,function(x){letters[1:x]}); names(dimnames(A))[1:2]<-c("d1","d2"); \\ Sigma=array(drop(stats::rWishart(1,df=prod(b),diag(prod(b)))),rep(b,2)); \\ var_lin(A,Sigma)
```

W.ak

general AK weights as a function of a and k parameters.

### Description

general AK weights as a function of a and k parameters.

#### Usage

```
W.ak(
   months,
   groups,
   S = c(2:4, 6:8),
   S_1 = S - 1,
   a,
   k,
   eta0 = length(groups)/length(S),
   eta1 = eta0 - 1,
   rescaled = F
)
```

#### **Arguments**

months	an integer, indicating number of months
groups	a vector of character strings or numeric string
S	a vector of integers indicating the indices of the rotation group in the sample that overlap with the previous sample: groups[S] are the overlapping rotation groups
S_1	a vector of integers indicating the indices of the corresponding rotation group of S in the previous month
a	a numeric value

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k a numeric value

rescaled a boolean (default FALSE) indicating whether these AK coefficient are to be

applied to rescaled or not rescaled month in sample weighted sums

nmonth an integer, indicating number of months

ngroup a vector of character strings or numeric string

#### Value

an array of AK coefficients W[m2,m1,mis1] such that Ak estimate for month m2 is sum(W[y2,,])\*Y) where Y[m1,mis1] is direct estimate on mis mis1 for emp stat y1 at month m1.

#### **Examples**

```
W<-W.ak(months=1:3,groups=1:8,a=.2,k=.5);dimnames(W)
W<-W.ak(months=2:4,groups=letters[1:8],a=.2,k=.5);dimnames(W);
Y<-WSrg(list.tables,weight="pwsswgt",list.y="pemlr",rg="hrmis")
dimnames(Y);month="m";group="mis";variable="y";
A=W.ak(months = dimnames(Y)[[month]],groups = dimnames(Y)[[group]],S=c(2:4,6:8),a=.5,k=.5)</pre>
```

W.multi.ak

general AK weights as a function of a and k parameters.

#### Description

general AK weights as a function of a and k parameters.

#### Usage

```
W.multi.ak(
  months,
  groups,
  S,
  S_1 = S - 1,
  ak,
  eta0 = length(groups)/length(S),
  eta1 = eta0 - 1,
  rescaled = F
)
```

#### **Arguments**

S a vector of integers indicating the indices of the rotation group in the sample

ak a list of 2-dimension vectors

nmonth an integer, indicating number of months

ngroups : number of groups

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#### Value

an array of AK coefficients W[m2,m1,mis1] such that Ak estimate for month m2 is sum(W[y2,,])\*Y) where Y[m1,mis1] is direct estimate on mis mis1 for emp stat y1 at month m1.

#### **Examples**

```
W.multi.ak(months=1:3,groups=1:8,S=c(2:4,6:8),ak=list(c(a=.2,k=.5),c(a=.2,k=.4)))
```

WS

Compute weighted sums

#### **Description**

Compute weighted sums

#### Usage

```
WS(list.tables, weight = 1, list.y = NULL, sep = "_n")
```

### Arguments

list.tables A list of dataframes

weight either a real number of a character string indicating the name of the weight

variable.

list.y: list of variables whose weighted sum needs to be computed. It can be factor or

character variables.

#### Value

a dataframe.

```
WS(plyr::dlply(CRE_data,.variables=~time), "Sampling.weight",c("Hobby","Status","State"));
WS(plyr::dlply(CRE_data,.variables=~time), "Sampling.weight",character(0));
```

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WSrg

Weighted sums by rotation groups

#### Description

Weighted sums by rotation groups

#### Usage

```
WSrg(list.tables, weight = 1, list.y = NULL, rg = "hrmis", rescale = F)
```

### Arguments

```
list.tables a named list of data frames

weight a character string indicating the variable name or a numerical value

list.y a vector of character strings indicating the study variables

rg a character string indicating the name of the rotation group.
```

#### Value

an array

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