

# Gustave: An R *package* for variance estimation in surveys

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# Why computing variance estimations of a survey indicator ?

- Variance estimation is important **for the data analyst** :
  - Enables to get a confidence interval
  - Enables to comment variations' significance
- Variance estimation is a way to **measure survey indicators' quality**
  - Included in the quality reports sent to Eurostat
  - Required in different surveys by the new European framework regulation for household surveys (*Integrated European social statistics*) under negotiation
- Variance estimation is an important step in the **production process of a survey**

What Gustave is supposed to do :

- Facilitates interactions between methodologists and data analysts
- Offers a computational framework to methodologists and data analysts with existing functionalities (linearization functions, domains...)
- Simplifies analytical variance estimation

What Gustave does not do :

- Does not provide a ready for use variance estimation function to methodologists
- Does not offer a computational framework for variance estimation with bootstrap

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- Variance estimation for domains included in Gustave
- Some linearization functions included in Gustave (*ratio*, *diff\_of\_ratio*, *mean*)
- Some variance functions already coded (Sen Yates Grundy variance estimator *varSYG*, Deville-Tille variance estimator *varDT*)
- A way to take into account calibration included in the package (*res\_cal*)

# The use of Gustave for the methodologist

What the methodologist produces :

- A variance function for a total

What the methodologist does not need to code :

- Formatting
- Estimation for domains
- Classic linearization functions

- **First step** : Data computation in order to prepare all variables which are necessary for the variance estimation (for instance, non-response units' probability)
- **Second step** : Coding of the variance function  
**Example** with a two-stage sample scheme with primary units' sampling before a dwellings' sampling, a reweighting for nonresponse adjustment and a calibration on simulated data of Labour Force Survey

```
varEec <- function(y, up, log, ind){  
  
  variance <- list()  
  
  # Etape 0 : Agrégation par logement  
  y <- sum_by(y, by = ind$idlog)  
  
  # Etape 1 : Prise en compte du calage  
  y <- add_zero(y, log$id[log$call])  
  y <- res_cal(y, precalc = log$res_cal_precalc)
```

```

(...)

# Etape 2 : Prise en compte de la non-réponse
variance[["nr"]] <- colSums(
  (1/log$piolog[log$cal]^2 - log$qlog[log$cal]) *
  (1 - log$pinr[log$cal]) * (y/log$pinr[log$cal])^2
)
y <- add_zero(y / log$pinr[log$cal], log$id)

# Etape 3 : Sélection des logements dans les up
variance[["log"]] <- varDT(
  y, w = 1/(log$piup^2) - log$qup,
  precalc = log$varDT_precalc
)

# Etape 4 : Sélection des up
y <- sum_by(y, by = log$idup, w = 1/log$piolog_up)
y <- add_zero(y, up$id)
variance[["up"]] <- varDT(y, precalc = up$precalc)

colSums(do.call(rbind, variance))
}

```



## A key functionality : the variance *wrapper* (1/2)

The **variance wrapper adds functionalities** (linearization functions, estimation for domains, ...) to the variance function for totals coded by the methodologist

Last part of the example :

```
# Création du wrapper de variance avec define_variance_wrapper()
precisionEec <- define_variance_wrapper(
  variance_function = varEec,
  technical_data = list(up = up, log = log, ind = ind),
  reference_id = technical_data$ind$id,
  reference_weight = technical_data$ind$w,
  default_id = quote(paste0(ident, noi))
)

# Utilisation du wrapper de variance (données du T4 2014)
precisionEec(z, acteu %in% 2)

##               call      est  variance      std      cv    lower
## 1 total(y = acteu %in% 2) 3001046 2158830156 46463.21 1.548234 2909980
##      upper
## 1 3092112
```

# A key functionality : the variance *wrapper* (2/2)

## Variance estimation of unemployment **rate** :

```
precisionEec(z, ratio(acteu %in% 2, acteu %in% c(1, 2)))  
##                                call  
## 1 ratio(num = acteu %in% 2, denom = acteu %in% c(1, 2))  
##      est      variance      std      cv      lower  
## 1 0.1044647 2.5918e-06 0.001609907 1.541101 0.1013094
```

## Variance estimation of unemployment rate for people **over 50** :

```
precisionEec(z,  
  ratio(acteu %in% 2, acteu %in% c(1, 2)),  
  where = age >= 50  
)  
##      est      variance      std      cv  
## 1 0.07047492 5.402617e-06 0.002324353 3.298128
```

## Variance estimation of unemployment rate **by region** :

```
precisionEec(z,  
  ratio(acteu %in% 2, acteu %in% c(1, 2)),  
  by = reg  
)  
##  by      est      variance      std      cv  
## 1 11 0.1003089 1.538408e-05 0.003922254 3.910175  
## 2 21 0.1130015 1.068723e-04 0.010337904 9.148463  
## 3 22 0.1220682 9.565600e-05 0.009780388 8.012235
```

# An expandable *package*

`define_variance_wrapper()` can handle **any variance function in input** :

- as many parameters as necessary for the variance function
- use of other *packages* to code the variance function (use of `require()` in the variance function)

**New linearization functions** with `define_statistic_wrapper()` :

```
# Définition du coefficient de gini à partir du package vardpoor
gini <- define_statistic_wrapper(
  statistic_function = function(y, weight){
    require(vardpoor)
    result <- lingini(Y = y, weight = weight)
    list(point = result$value$Gini, lin = result$lin$lin_gini)
  },
  arg_type = list(data = "y", weight = "weight", param = NULL)
)

# Utilisation pour calculer la précision dans l'enquête SRCV en 2014
precisionSrcv(r, gini(HX090))
```

## qvar() : a ready for use variance estimation function

- "What Gustave does not do : to provide a ready for use variance estimation function to methodologists"... but for **one specific case**
- **A ready for use variance estimation function** for the following case :
  - Stratified simple random sampling
  - Nonresponse adjustment by reweighting with response homogeneity groups
  - Calibration
- Useful for business surveys
- qvar() built as a variance estimation function, which is integrated in a wrapper

Examples of use at Insee :

- Labour Force Survey
- Statistics on Income and Living Conditions
- Some business surveys

Used for surveys with different issues :

- French master sample (household surveys, formula from Breidt, Chauvet, 2011 and Gros, Moussallam, 2015)
- Multiple stages for indicators on individuals (Cadre de vie et sécurité)
- Weight sharing (SILC)

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# Constraints for dissemination

- Warning concerning the dissemination of the variance function produced in Gustave :

The variance *wrapper* is a fully self-sufficient function : answers to the survey and information describing the sample scheme are given in input in the `technical_data` parameter and are kept in the function environment (it's a *closure*)

- Be careful to disseminate the variance function only to people who have the right to access your data and are aware of the rules on confidentiality

# Dissemination, unit testing and seamless integration

- *package* available on [Cran](#), last version : August 2018
- **Open source code** on several development platforms, in particular [github.com](#)
- possibility for users to suggest **upgrades** (to rectify bugs, to answer to specific requests, *pull requests*)
- More than **180 unitary tests** in Gustave to test Gustave's functionalities
- Tests which are automatically computed at each *package's* new version : **seamless integration**



Insee's Statistical Method Department developed an R *package* to **simplify variance estimation** :

- a solution that simplify methodologists' work
- production of variance estimation functions which are simple to use
- a documented *package*, with a vignette in progress

A *package* **used for Insee's household surveys** :

- to produce variance estimation joined to quality reports
- to check that surveys satisfy to IEES's precision constraints