

# Working with the **SubnationalCRVS** R Package

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## Contents

<b>1</b>	<b>Setup</b>	<b>1</b>
<b>2</b>	<b>Datasets Included with the Package: Tabulations from Ecuador</b>	<b>2</b>
<b>3</b>	<b>Example of Subnational Analysis: Provinces in Ecuador</b>	<b>3</b>
3.1	Demographic Data Quality Assessment (DDQA) . . . . .	3
3.2	DDM Estimation of Death Registration Completeness . . . . .	15
<b>4</b>	<b>Example of National Analysis: Ecuador</b>	<b>18</b>
4.1	Plot of sex ratios . . . . .	19
4.2	Plot of age ratios . . . . .	20
4.3	Plots related to age heaping . . . . .	21
4.4	Plot of DDM estimates . . . . .	24
	<b>References</b>	<b>26</b>

## 1 Setup

The **SubnationalCRVS** package, which is still under active development, is hosted on the GitHub page [www.github.com/ConVERGE-UNFPA/SubnationalCRVS](https://www.github.com/ConVERGE-UNFPA/SubnationalCRVS) rather than on CRAN. As a result, **SubnationalCRVS** cannot be installed with the usual `install.packages()` function. Instead, **SubnationalCRVS** can be installed with the `install_github()` function from the **devtools** package. The key dependency **DemoTools** (Riffe et al. 2019) is also hosted on GitHub instead of CRAN and can also be installed with `install_github()`.

```
install.packages("devtools")
library(devtools)
install_github("timriffe/DemoTools") # install the DemoTools dependency
install_github("ConVERGE-UNFPA/SubnationalCRVS") # install the SubnationalCRVS package
```

Now we can load **SubnationalCRVS**, specify the name of the local folder in which we will save the plots produced by the package, and create that local folder if it does not already exist.

```
library(SubnationalCRVS)
my_plots_dir <- "Plots/" # local folder where the plots should be saved
dir.create(my_plots_dir) # create the folder if it does not already exist
```

For this demonstration, we will also load the `dplyr` package to customize the display of some of the tables returned by `SunbationalCRVS`.

```
library(dplyr)
```

## 2 Datasets Included with the Package: Tabulations from Ecuador

The `SubnationalCRVS` package comes with tabulations of population and registered deaths in Ecuador – disaggregated by age, sex, and province – created from publicly available datasets from Ecuador (Peralta et al. 2019; INEC 2010) based on the Ecuador’s 2001 Census, 2010 Census, and annual counts of registered deaths from 2001 through 2010. The structure of the included example datasets is inspired by the data requirements of the `ddm()` function from the `DDM` package (Riffe, Lima, and Queiroz 2017).

The `ecuador_single_year_ages_combined` dataset reports the estimated populations in Ecuador by single-year ages (0, 1, 2, ... in the `age` column) from both the 2001 Census (`pop1` column) and the 2010 Census (`pop2` column), separately for males and females (`m` and `f` in the `sex` column) and province (in the `province` column).

```
head(ecuador_single_year_ages)
```

##	province_name	sex	age	pop1	pop2	year1	month1	day1	year2	month2	day2
## 1	Azuay	m	0	6086	6750	2001	11	25	2010	11	28
## 2	Azuay	m	1	6555	6984	2001	11	25	2010	11	28
## 3	Azuay	m	2	7232	7090	2001	11	25	2010	11	28
## 4	Azuay	m	3	7101	7095	2001	11	25	2010	11	28
## 5	Azuay	m	4	7083	6961	2001	11	25	2010	11	28
## 6	Azuay	m	5	6583	6895	2001	11	25	2010	11	28

The `ecuador_single_year_ages_combined` dataset appends rows to `ecuador_single_year_ages_combined` that report the sex- and single-year-age- disaggregated population estimates for the entire country (as opposed to a single province).

```
ecuador_single_year_ages_combined %>% filter(province_name == "National") %>%
  head()
```

##	province_name	sex	age	pop1	pop2	year1	month1	day1	year2	month2	day2
## 1	National	m	0	117897	123013	2001	11	25	2010	11	28
## 2	National	m	1	133280	141602	2001	11	25	2010	11	28
## 3	National	m	2	141869	141719	2001	11	25	2010	11	28
## 4	National	m	3	134793	143418	2001	11	25	2010	11	28
## 5	National	m	4	139268	143830	2001	11	25	2010	11	28
## 6	National	m	5	131491	136479	2001	11	25	2010	11	28

The `ecuador_five_year_ages` and `ecuador_five_year_ages_combined` datasets have the same variables as `ecuador_single_year_ages_combined` and `ecuador_single_year_ages_combined` with two exceptions: (1) the `age` variable now represents five-year age groups (in the `age` column, with 0-4 coded as 0, 5-9 coded as 5, 10-14 coded as 10, etc.) instead of single-year ages; and (2) there is an additional column called `deaths` that reports the registered deaths collected between 2001 and 2010.

```
head(ecuador_five_year_ages)
```

```
##   province_name sex age  pop1  pop2 deaths year1 month1 day1 year2 month2 day2
## 1      Azuay    m   0 34101 34886   772  2001     11   25  2010     11   28
## 2      Azuay    m   5 34996 36406   232  2001     11   25  2010     11   28
## 3      Azuay    m  10 34946 38125   223  2001     11   25  2010     11   28
## 4      Azuay    m  15 32387 37611   416  2001     11   25  2010     11   28
## 5      Azuay    m  20 25634 33665   480  2001     11   25  2010     11   28
## 6      Azuay    m  25 18606 28376   475  2001     11   25  2010     11   28
```

```
head(ecuador_five_year_ages_combined)
```

```
##   province_name sex age  pop1  pop2 deaths year1 month1 day1 year2 month2 day2
## 1      Azuay    m   0 34101 34886   772  2001     11   25  2010     11   28
## 2      Azuay    m   5 34996 36406   232  2001     11   25  2010     11   28
## 3      Azuay    m  10 34946 38125   223  2001     11   25  2010     11   28
## 4      Azuay    m  15 32387 37611   416  2001     11   25  2010     11   28
## 5      Azuay    m  20 25634 33665   480  2001     11   25  2010     11   28
## 6      Azuay    m  25 18606 28376   475  2001     11   25  2010     11   28
```

```
ecuador_five_year_ages_combined %>% filter(province_name == "National") %>%
  head()
```

```
##   province_name sex age  pop1  pop2 deaths year1 month1 day1 year2 month2
## 1   National    m   0 678280 743576 22338  2001     11   25  2010     11
## 2   National    m  10 679067 782559  4290  2001     11   25  2010     11
## 3   National    m  15 616725 712878  8293  2001     11   25  2010     11
## 4   National    m  20 569964 638042 13412  2001     11   25  2010     11
## 5   National    m  25 456230 585652 13301  2001     11   25  2010     11
## 6   National    m  30 422307 519493 11811  2001     11   25  2010     11
##   day2
## 1    28
## 2    28
## 3    28
## 4    28
## 5    28
## 6    28
```

### 3 Example of Subnational Analysis: Provinces in Ecuador

In this section, we describe the key functions of `SubnationalCRVS` in the context of visualizing outputs from a demographic data quality assessment (DDQA) and estimates of death registration completeness (Riffe, Lima, and Queiroz 2017) within provinces of Ecuador. Later in this tutorial, we also show how `SubnationalCRVS` provides corresponding visualizations for national-level results.

#### 3.1 Demographic Data Quality Assessment (DDQA)

##### 3.1.1 Sex ratios: `PlotSexRatios()`

A key step in this demographic data quality assessment (DDQA) process is to use the `PlotSexRatios()` function to compute and plot sex ratios within each combination of province, sex, and single-year age for

the 2001 and 2010 data stored in `ecuador_single_year_ages_combined`.

To use `PlotSexRatios()`, we are required to provide our tabulated data frame in the `data` argument and a few additional required arguments that describe the variable names and values in `data`. These arguments and the expected format of the specified dataset are motivated by the data structure enforced in the DDM package (Riffe, Lima, and Queiroz 2017).

- `name.disaggregations` is the name of variable representing the subnational disaggregation (apart from sex, which is required,) in `data`. Here we specify `name.disaggregations="province_name"`.
- `name.sex` is the name of variable representing sex. Here we specify `name.sex="sex"`
- `name.age` is the name of variable representing age. Here we specify `name.age="age"`
- `name.year1` is the name of variable that provides the year of the earlier of the two time periods represented in `data`. Here we specify `name.year1="year1"` (the value of this variable, "2001" was the year of Ecuador's 2001 Census)
- `name.month1` is the name of variable that provides the month of the earlier of the two time periods represented in `data`. Here we specify `name.month1="month1"` (the value of this variable, "11" was the numerical month, November, of Ecuador's 2001 Census)
- `name.day1` is the name of variable that provides the day of the earlier of the two time periods represented in `data`. Here we specify `name.day1="day1"` (the value of this variable, "25" was the day of Ecuador's 2001 Census)
- `name.year2` is the name of variable that provides the year of the later of the two time periods represented in `data`. Here we specify `name.year2="year2"` (the value of this variable, "2010" was the year of Ecuador's 2010 Census)
- `name.month2` is the name of variable that provides the month of the later of the two time periods represented in `data`. Here we specify `name.month2="month2"` (the value of this variable, "11" was the numerical month, November, of Ecuador's 2010 Census)
- `name.day2` is the name of variable that provides the day of the later of the two time periods represented in `data`. Here we specify `name.day2="day2"` (the value of this variable, "28" was the day of Ecuador's 2010 Census)
- `name.population.year1` is the name of variable representing the population in the earlier of the two time periods represented in the dataset. Here we specify `name.population.year2="pop1"`
- `name.population.year2` is the name of variable representing the population in the earlier of the two time periods represented in the dataset. Here we specify `name.population.year2="pop2"`
- `name.male` is the name of value of the `name.sex` variable that represents males. Here we specify `name.males="m"`
- `name.female` is the name of value of the `name.sex` variable that represents females. Here we specify `name.females="f"`

In addition, we specify two optional arguments, `plots.dir` and `name.national`. We set `plots.dir=my_plots_dir` so that the plots will save in the local folder `Plots`; if the `plots.dir` argument is omitted, the plots will be saved in the same working directory of the R script. We also set `name.national="National"` to reflect the fact that national-level datasets are provided in rows where the `name.disaggregations` variable has the value "National". Specifying `name.national` produces visualizations catered specifically to national-level analysis (instead of accommodated further disaggregation) that will be presented later in the tutorial. If `name.national` is not provided, those national-level visualizations are simply not produced.

```
s <- PlotSexRatios(data=ecuador_single_year_ages_combined,
  name.disaggregations="province_name",
  name.males="m",
  name.females="f",
  name.age="age",
  name.sex="sex",
  name.population.year1="pop1",
  name.population.year2="pop2",
  name.year1="year1",
  name.month1="month1",
  name.day1="day1",
  name.year2="year2",
  name.month2="month2",
  name.day2="day2",
  name.national="National",
  label.subnational.level="Province",
  plots.dir=my_plots_dir)
```

The plots of sex ratios are saved in the `Plots/` sub-folder we specified with the argument `plots.dir=my_plots_dir`; `plots.dir` is an optional argument and, if we do not specify a value for it, the plots will be saved the working directory. We also specified `label.subnational.level="Province"` so that the disaggregations are labeled `Province` instead of the less clear `province_name`.

The sex ratios for all levels of subnational disaggregation are overlaid in the following “combined” plots separately for each data year.

```
knitr::include_graphics(path=paste0(my_plots_dir,
  "sex_ratios_combined_province_name_",
  Sys.Date(),
  ".pdf"))
```

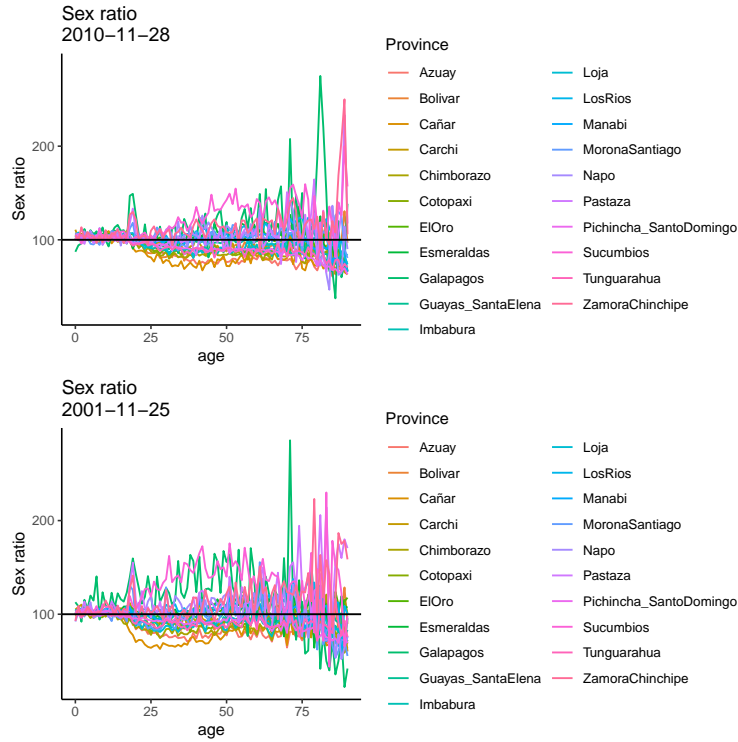


Figure 1: Sex ratios in Ecuador by province, combined plot

Additionally, the sex ratios are plotted in separate figures for each level of subnational disaggregation in the following “disaggregated” plots.

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "sex_ratios_by_province_name_",
                                     Sys.Date(),
                                     ".pdf"))
```

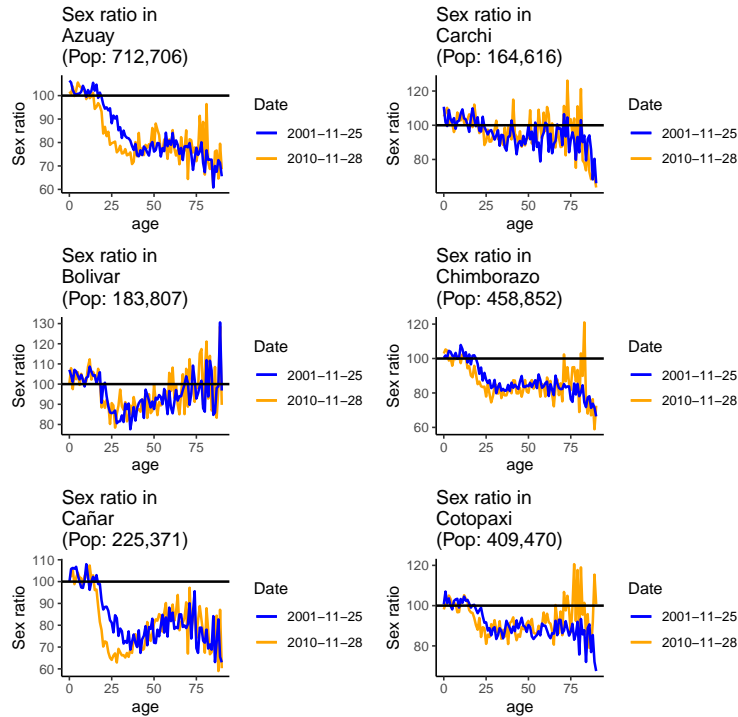


Figure 2: Sex ratios in Ecuador by province, disaggregated plots (only first page shown)

### 3.1.2 View sex ratios in table

The object returned by `PlotSexRatios()` is a table that shows us the sex ratios for each combination of province, sex, and single-year age in the `sex_ratio_1` column (for the 2001 Census) and the `sex_ratio_2` column (for the 2010 Census).

```
s %>% select(province_name, age, pop1, pop2, sex_ratio_1, sex_ratio_2) %>%
  head()
```

##	province_name	age	pop1	pop2	sex_ratio_1	sex_ratio_2
## 1	Azuay	0	12073	13092	101.7	106.4
## 2	Azuay	1	13060	13596	100.8	105.6
## 3	Azuay	2	14195	14014	103.9	102.4
## 4	Azuay	3	14217	14124	99.8	100.9
## 5	Azuay	4	14012	13911	102.2	100.2
## 6	Azuay	5	12815	13736	105.6	100.8

### 3.1.3 Age ratios: `PlotAgeRatios()`

Another step in our demographic data quality assessment is using the `PlotAgeRatios()` function in the `SubnationalCRVS` package to compute and plot age ratios within each combination of province and sex for the 2001 and 2010 data stored in `ecuador_five_year_ages_combined`. The arguments we provide to `PlotAgeRatios()` are actually identical to those we specified for `PlotSexRatios()`, except now we are using the tabulation with five-year age groups (`ecuador_five_year_ages_combined`) instead of the tabulation with single-year ages.

```
a <- PlotAgeRatios(data=ecuador_five_year_ages_combined,
  name.disaggregations="province_name",
  name.males="m",
  name.females="f",
  name.age="age",
  name.sex="sex",
  name.population.year1="pop1",
  name.population.year2="pop2",
  name.year1="year1",
  name.month1="month1",
  name.day1="day1",
  name.year2="year2",
  name.month2="month2",
  name.day2="day2",
  label.subnational.level="Province",
  name.national="National",
  plots.dir=my_plots_dir)
```

### 3.1.4 View age ratios in combined plot

The following “combined” plots, saved in the Plots/ folder, show the age ratios for all levels of subnational disaggregation, separately for males and females in each data year.

```
knitr::include_graphics(path=paste0(my_plots_dir,
  "age_ratios_combined_province_name_",
  Sys.Date(),
  ".pdf"))
```



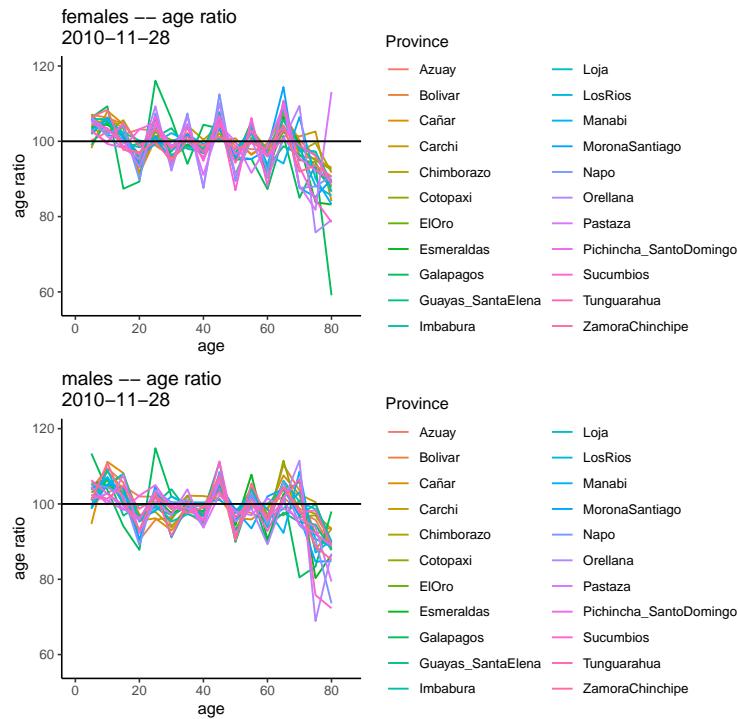


Figure 3: Age ratios in Ecuador by province, combined plot

In addition, `PlotAgeRatios()` also creates the following “disaggregated” plots, saved in the `Plots/` folder, where the age ratios for each level of disaggregation are shown in separate plots, with different sexes and data years overlaid within each plot.

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "age_ratios_by_province_name_",
                                     Sys.Date(),
                                     ".pdf"))
```

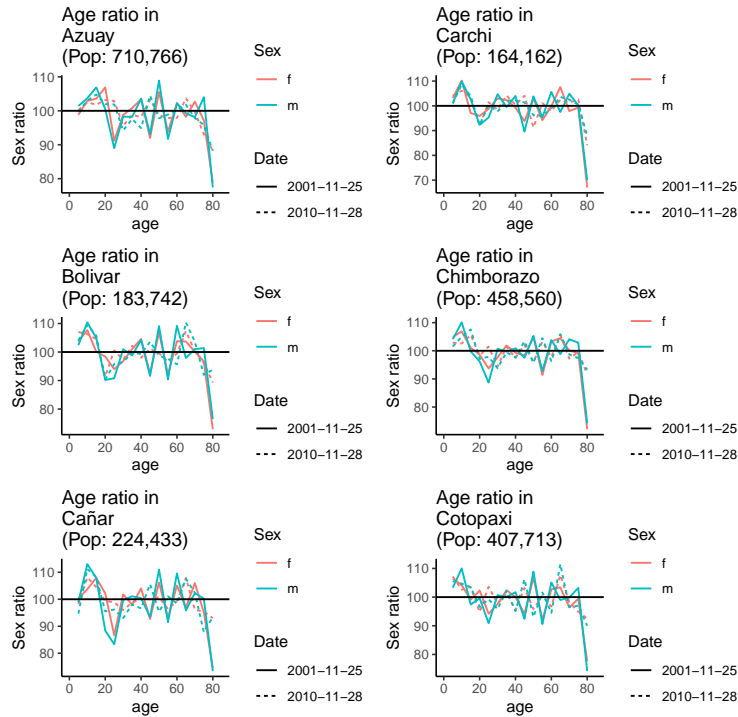


Figure 4: Age ratios in Ecuador by province, disaggregated plots (only first page shown)

Just as the `PlotSexRatios()` function returns a table of disaggregated sex ratios, `PlotAgeRatios()` returns a table of disaggregated age ratios.

```
a %>% select(province_name, age, pop1, pop2, age_ratio_1, age_ratio_2) %>%
  head()
```

```
## province_name age pop1 pop2 age_ratio_1 age_ratio_2
## 1 Azuay 0 33491 33876 NA NA
## 2 Azuay 5 33817 35701 98.8 100.2
## 3 Azuay 10 34975 37366 102.9 102.5
## 4 Azuay 15 34181 37215 103.6 101.8
## 5 Azuay 20 31000 35753 106.9 103.2
## 6 Azuay 25 23844 32054 91.2 102.9
```

### 3.1.5 Potential age heaping: `PlotPotentialAgeHeaping()`

To give us a sense of whether “age-heaping” is occurring within the levels of disaggregation present in our `ecuador_single_year_ages_combined` dataset, we turn to the `PlotPotentialAgeHeaping()` function and actually provide the same arguments we used in the `PlotSexRatios()` function.

```
PlotPotentialAgeHeaping(data=ecuador_single_year_ages_combined,
  name.disaggregations="province_name",
  name.males="m",
  name.females="f",
  name.age="age",
  name.sex="sex",
```

```

name.population.year1="pop1",
name.population.year2="pop2",
name.year1="year1",
name.month1="month1",
name.day1="day1",
name.year2="year2",
name.month2="month2",
name.day2="day2",
name.national="National",
label.subnational.level="Province",
plots.dir=my_plots_dir)

```

### 3.1.6 View potential age heaping in combined plot

The following “combined” plots show us estimated population counts by single-year ages with different provinces represented with different overlaid colors, and separate plots for each sex and data year.

```

knitr::include_graphics(path=paste0(my_plots_dir,
                                     "potential_age_heaping_combined_province_name_",
                                     Sys.Date(),
                                     ".pdf"))

```

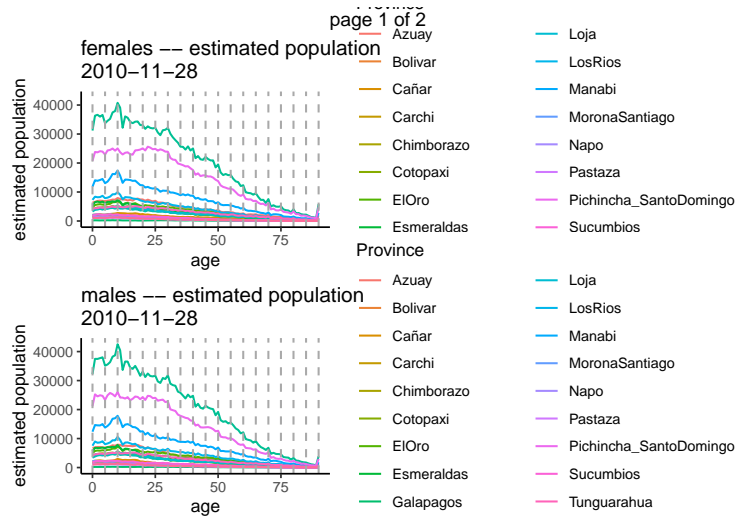


Figure 5: Population counts in Ecuador by single-year age, combined plot

PlotPotentialAgeHeaping() present separate plots of population counts for each province in Ecuador, with different sexes and data years overlaid within each plot.

```

knitr::include_graphics(path=paste0(my_plots_dir,
                                     "potential_age_heaping_by_province_name_",
                                     Sys.Date(),
                                     ".pdf"))

```

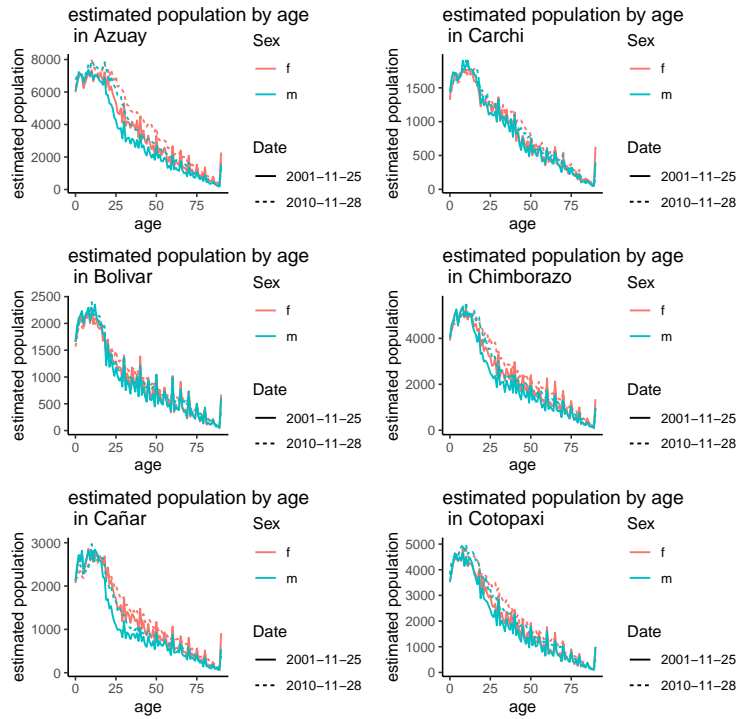


Figure 6: Population counts in Ecuador by single-year age, disaggregated plots

### 3.1.7 Age heaping indices: `PlotAgeHeapingScores()`

As a more concise summary of potential age-heaping suggested by the visualizations from `PlotPotentialAgeHeaping`, we now use the `PlotAgeHeapingScores` function with the same arguments we provided to the `PlotAgeRatios` function.

```
ageheaping <- PlotAgeHeapingScores(data=ecuador_single_year_ages_combined,
  name.disaggregations="province_name",
  name.males="m",
  name.females="f",
  name.age="age",
  name.sex="sex",
  name.population.year1="pop1",
  name.population.year2="pop2",
  name.year1="year1",
  name.month1="month1",
  name.day1="day1",
  name.year2="year2",
  name.month2="month2",
  name.day2="day2",
  name.national="National",
  label.subnational.level="Province",
  plots.dir=my_plots_dir)
```

One set of plots returned by `PlotAgeHeapingScores()` shows the values of three age-heaping indices within each combination of province, sex, and data year: Roughness, Whipple, and Myers. The indices are com-

puted with the `check_heaping_roughness()`, `check_heaping_Whipple()`, and `check_heaping_myers()` functions, respectively, from the `DemoTools` package (Riffe et al. 2019).

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "age_heaping_scores_combined_province_name_",
                                     Sys.Date(),
                                     ".pdf"))
```

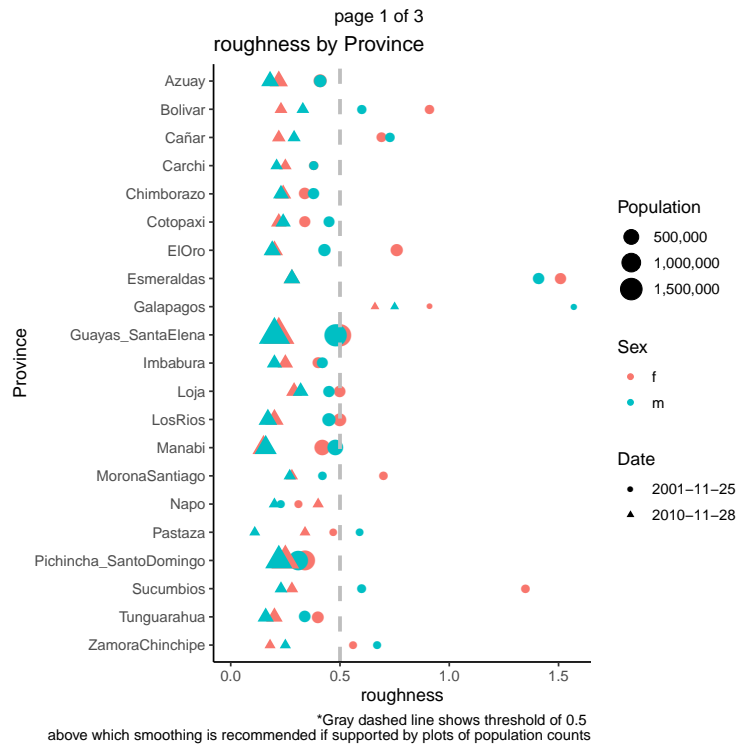


Figure 7: Roughness, Whipple, and Myers indices in Ecuador by province (only first page shown)

Further, `PlotAgeHeapingScores()` also creates plots showing the Noubbissi index in each province computed for each terminal digit from 0-9, using the `check_heaping_Noubbissi()` function from the `DemoTools` package (Riffe et al. 2019).

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "Noubbissi_by_province_name_",
                                     Sys.Date(),
                                     ".pdf"))
```

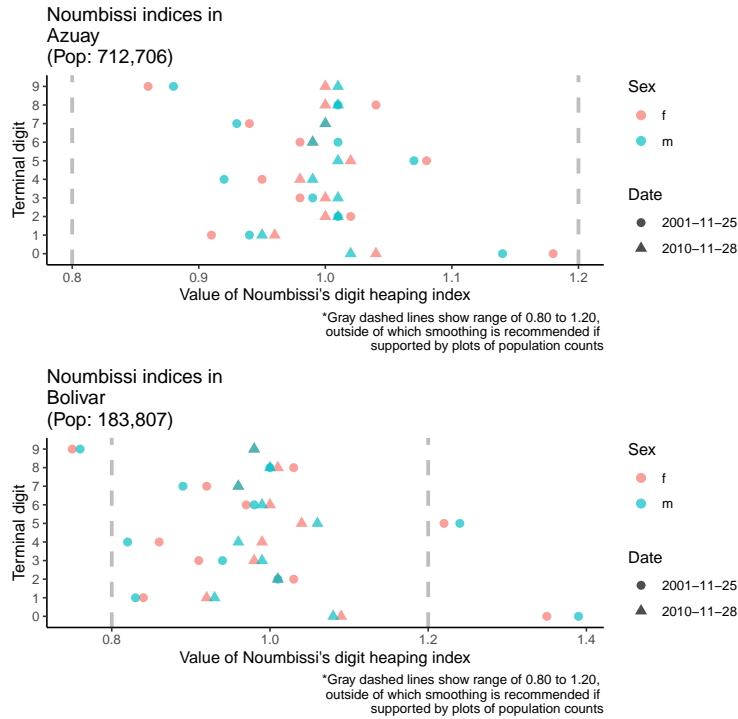


Figure 8: Noubbissi indices in Ecuador by province (only first page shown)

We can also view the age-heaping indices in the table returned by `PlotAgeHeapingScores()`

```
head(ageheaping)
```

##	province_name	date	total_pop	sex	roughness	Whipple	Myers	Noubbissi_0
## 1	Azuay	2001-11-25	319983	f	0.41	1.18	4.21	1.18
## 2	Bolivar	2001-11-25	86256	f	0.91	1.37	7.39	1.35
## 3	Cañar	2001-11-25	112041	f	0.69	1.22	4.89	1.22
## 4	Carchi	2001-11-25	77172	f	0.38	1.18	3.75	1.15
## 5	Chimborazo	2001-11-25	213106	f	0.34	1.25	5.44	1.23
## 6	Cotopaxi	2001-11-25	180328	f	0.34	1.27	5.99	1.25
##	Noubbissi_1	Noubbissi_2	Noubbissi_3	Noubbissi_4	Noubbissi_5	Noubbissi_6		
## 1	0.91	1.02	0.98	0.95	1.08	0.98		
## 2	0.84	1.03	0.91	0.86	1.22	0.97		
## 3	0.90	1.02	0.98	0.91	1.11	0.98		
## 4	0.91	1.05	0.96	0.92	1.11	0.98		
## 5	0.88	1.01	0.97	0.89	1.12	0.98		
## 6	0.88	1.02	0.95	0.90	1.18	0.98		
##	Noubbissi_7	Noubbissi_8	Noubbissi_9					
## 1	0.94	1.04	0.86					
## 2	0.92	1.03	0.75					
## 3	0.95	1.03	0.80					
## 4	0.96	1.04	0.86					
## 5	0.94	1.05	0.82					
## 6	0.91	1.06	0.80					

## 3.2 DDM Estimation of Death Registration Completeness

The structure of the `ecuador_five_year_ages` dataset is inspired by the requirements for the `ddm()` function from the DDM package (Riffe, Lima, and Queiroz 2017), which uses established Death Distribution Methods (DDM) to estimate death registration completeness for adults (aged 15+) between two consecutive Censuses. Essentially, the `SubnationalCRVS` package offers its `EstimateDDM()` function only as a convenient wrapper to `DDM::ddm()` to perform DDM estimation of death registration completeness using the same pipeline used to perform the DDQA.

### 3.2.1 Compute DDM estimates: `EstimateDDM()`

```
ddm_results <- EstimateDDM(data=ecuador_five_year_ages_combined,
  name.disaggregations="province_name",
  name.age="age",
  name.sex="sex",
  name.males="m",
  name.females="f",
  name.year1="year1",
  name.month1="month1",
  name.day1="day1",
  name.year2="year2",
  name.month2="month2",
  name.day2="day2",
  name.population.year1="pop1",
  name.population.year2="pop2",
  name.national="National",
  name.deaths="deaths",
  deaths.summed=TRUE)
```

```
## [1] "performing GGB-SEG estimation within each of 20 possible age ranges..."
```

We note `EstimateDDM()` uses exactly the same arguments as `PlotAgeRatios()`, for example, from the DDQA with two additional required arguments:

- `name.deaths`, which provides the name of the variable representing the count of registered deaths between the two dates represented in `name.date1` and `name.date2`
- `deaths.summed`, which should be set to `TRUE` when the `name.deaths` variable represents the total number of registered deaths `name.date1` and `name.date2` and set to `FALSE` when the `name.deaths` variable represents the average number of registered deaths between the two dates.

### 3.2.2 Plot DDM estimates: `PlotDDM()`

We can plot the estimated adult death registration completeness (using the “hybrid” GGB-SEG method) with the `PlotDDM` function:

```
PlotDDM(ddm_results=ddm_results,
  label.subnational.level="Province",
  plots.dir=my_plots_dir)
```

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "ggbseg_point_estimates_combined_province_name_",
                                     Sys.Date(),
                                     ".pdf"))
```

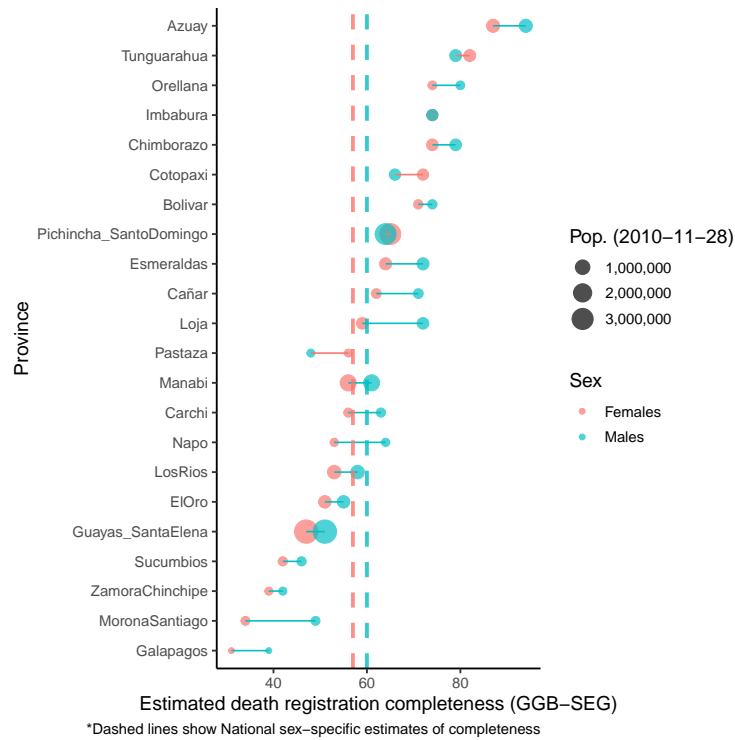


Figure 9: Point estimates of death registration completeness in Ecuador from 2001-2010, using the GGB-SEG method

The `EstimateDDM` function also returns a list, in which the `ddm_estimates` object contains the estimated death registration completeness with the GGB-SEG approach

```
head(ddm_results$ggbseg_estimates)
```

##	cod	sex	ggbseg	lower_age_range	upper_age_range	total_pop1	total_pop2
## 1	Azuay	Females	0.87	15	50	599313	710766
## 2	Azuay	Males	0.94	15	50	599313	710766
## 3	Bolivar	Females	0.71	15	65	170696	183742
## 4	Bolivar	Males	0.74	15	65	170696	183742
## 5	Cañar	Females	0.62	15	65	206346	224433
## 6	Cañar	Males	0.71	15	65	206346	224433

Additionally, `PlotDDM` presents a visualization

of the point estimates of completeness and the corresponding root mean squared errors (RMSEs) for each of the permitted values of the age range that is selected `DDM::ddm()` as part of its fitting procedure.



```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "ggbseg_sensitivity_province_name_",
                                     Sys.Date(),
                                     ".pdf"))
```

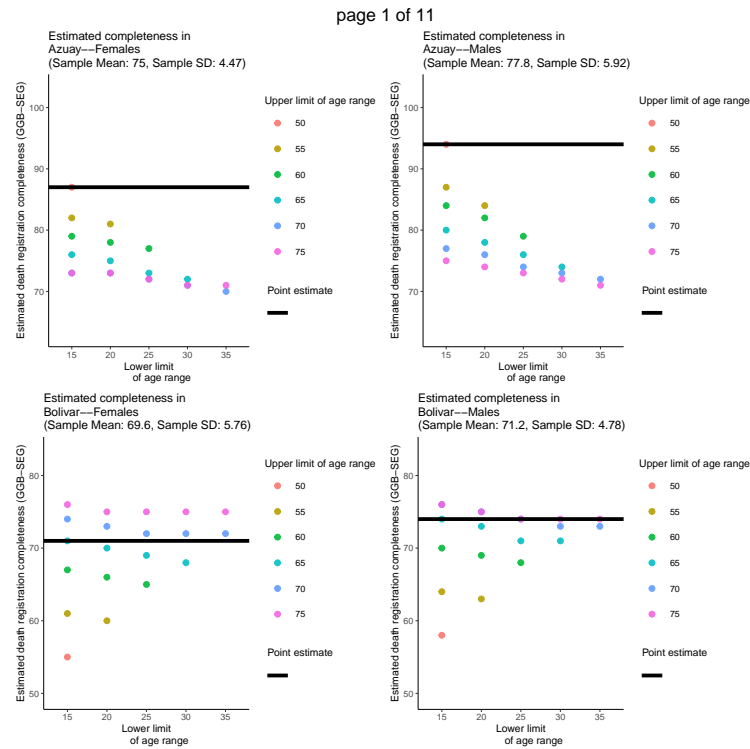


Figure 10: Sensitivity of point estimates of death registration completeness in Ecuador from 2001-2010 to choice of age-range parameter in the GGB-SEG method

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "ggbseg_RMSE_province_name_",
                                     Sys.Date(),
                                     ".pdf"))
```

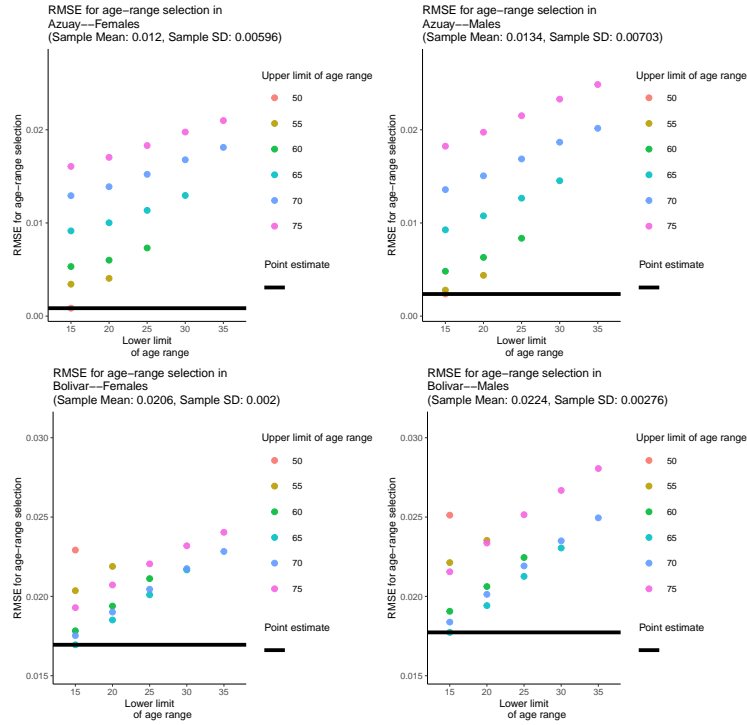


Figure 11: Sensitivity of point estimates of death registration completeness in Ecuador from 2001-2010 to choice of age-range parameter in the GGB-SEG method

The sensitivity estimates are also returned in table by `EstimateDDM`, in the `sensitivity_ddm_estimates` element of its list.

```
head(ddm_results$sensitivity_ggbseg_estimates, n=5)
```

```
##      cod      sex ggbseg lower_age_range upper_age_range total_pop1 total_pop2
## 1 Azuay Females  0.87          15          50      599313      710766
## 2 Azuay Females  0.82          15          55      599313      710766
## 3 Azuay Females  0.81          20          55      599313      710766
## 4 Azuay Females  0.79          15          60      599313      710766
## 5 Azuay Females  0.78          20          60      599313      710766
##      RMSE
## 1 0.0008509
## 2 0.0034260
## 3 0.0040540
## 4 0.0053210
## 5 0.0060060
```

## 4 Example of National Analysis: Ecuador

As discussed, specifying the argument `name.national="National"` during the previous DDQA and estimation of death registration completeness permitted `SubnationalCRVS` to save plots specifically designed to display national-level (as opposed to subnational-level) visualizations. These

national-level plots are can also be produced by specifying a national-level-only dataset, for example `ecuador_five_year_ages_national` (shown below), as long as the corresponding `name.national` argument is also specified (e.g. `name.national="National"` here)

```
head(ecuador_five_year_ages_national)
```

```
## province_name sex age pop1 pop2 deaths year1 month1 day1 year2 month2
## 1 National m 0 678280 743576 22338 2001 11 25 2010 11
## 2 National m 10 679067 782559 4290 2001 11 25 2010 11
## 3 National m 15 616725 712878 8293 2001 11 25 2010 11
## 4 National m 20 569964 638042 13412 2001 11 25 2010 11
## 5 National m 25 456230 585652 13301 2001 11 25 2010 11
## 6 National m 30 422307 519493 11811 2001 11 25 2010 11
## day2
## 1 28
## 2 28
## 3 28
## 4 28
## 5 28
## 6 28
```

```
tail(ecuador_five_year_ages_national)
```

```
## province_name sex age pop1 pop2 deaths year1 month1 day1 year2 month2
## 31 National f 60 149508 203789 12500 2001 11 25 2010 11
## 32 National f 65 126310 166559 14635 2001 11 25 2010 11
## 33 National f 70 99469 123647 18139 2001 11 25 2010 11
## 34 National f 75 73819 86449 21616 2001 11 25 2010 11
## 35 National f 80 52335 62304 23629 2001 11 25 2010 11
## 36 National f 85 72515 55087 52329 2001 11 25 2010 11
## day2
## 31 28
## 32 28
## 33 28
## 34 28
## 35 28
## 36 28
```

## 4.1 Plot of sex ratios

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "sex_ratios_national_",
                                     Sys.Date(),
                                     ".pdf"))
```

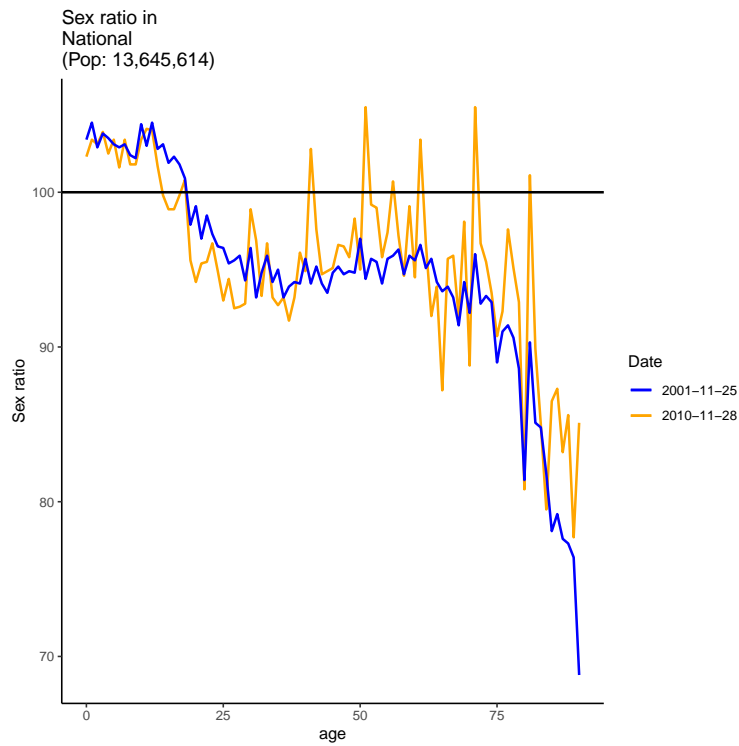


Figure 12: Sex ratios in Ecuador

## 4.2 Plot of age ratios

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "age_ratios_national_",
                                     Sys.Date(),
                                     ".pdf"))
```

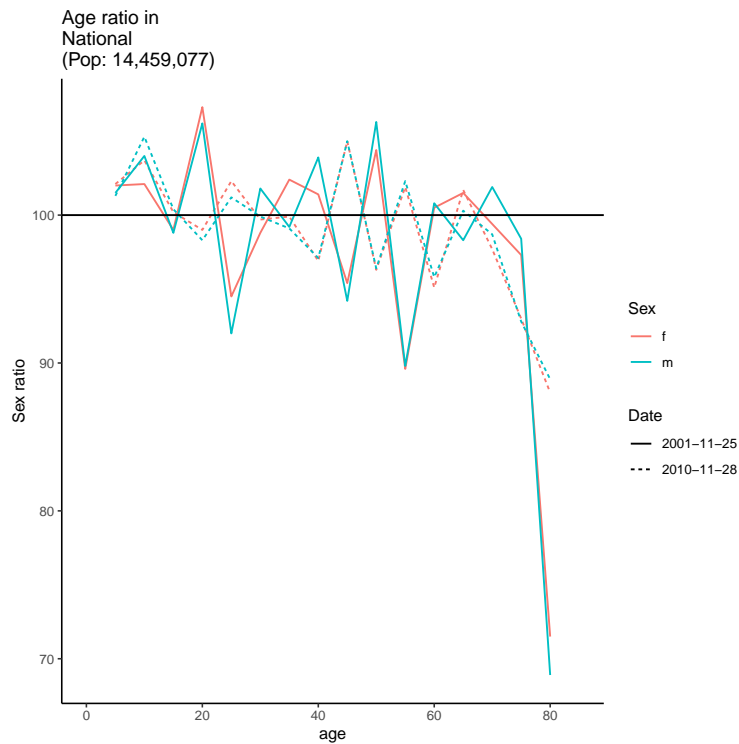


Figure 13: Age ratios in Ecuador

## 4.3 Plots related to age heaping

### 4.3.1 Plot of potential age heaping

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "potential_age_heaping_National_",
                                     Sys.Date(),
                                     ".pdf"))
```

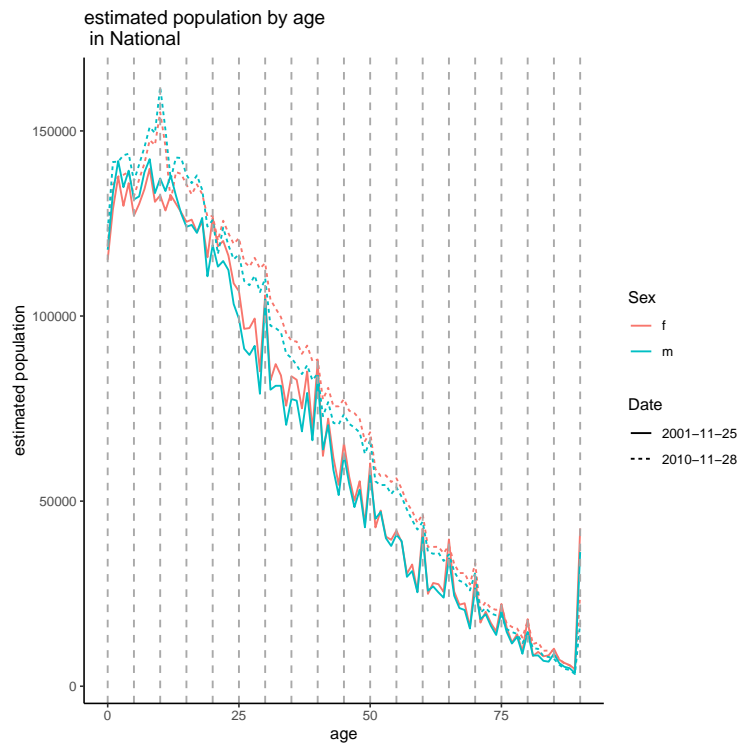


Figure 14: Population counts in Ecuador by single-year age

#### 4.3.2 Plot of age heaping indices

```
knitr::include_graphics(path=paste0(my_plots_dir,
                                     "age_heaping_scores_National_",
                                     Sys.Date(),
                                     ".pdf"))
```

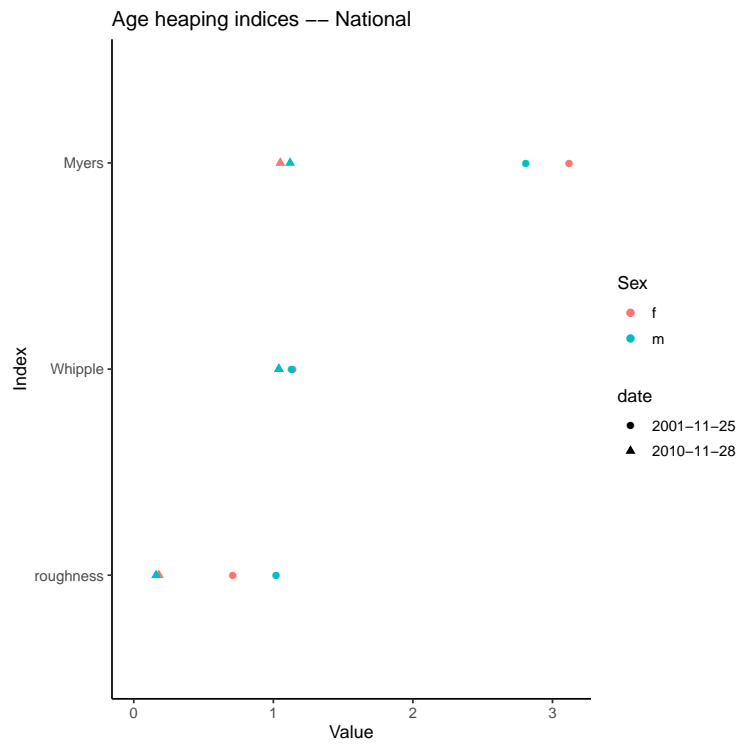


Figure 15: Roughness, Whipple, and Myers indices in Ecuador

### 4.3.3 Plot of Noumbissi indices

```
knitr::include_graphics(path=paste0(my_plots_dir,
  "Noumbissi_National_",
  Sys.Date(),
  ".pdf"))
```

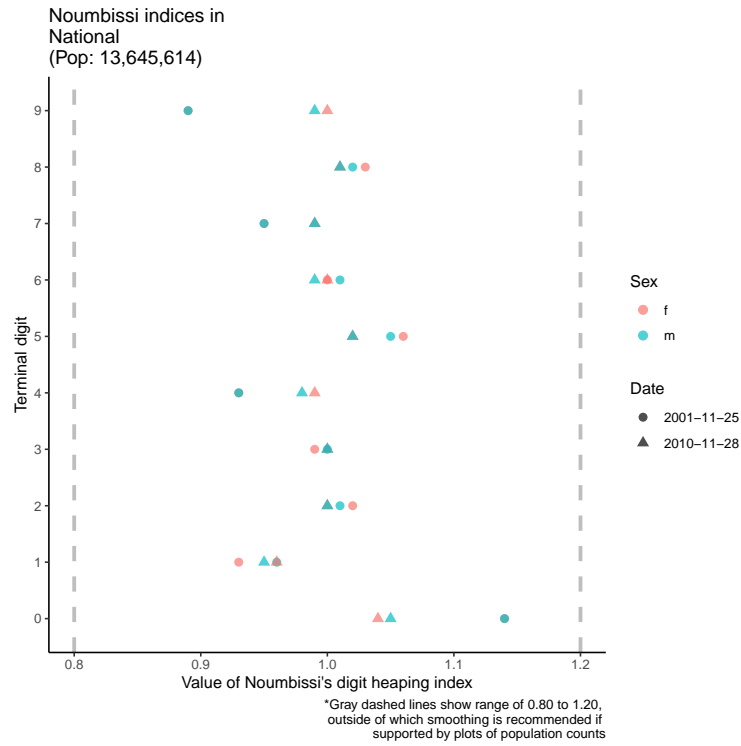


Figure 16: Noumbissi indices in Ecuador

#### 4.4 Plot of DDM estimates

```
knitr::include_graphics(path=paste0(my_plots_dir,
  "ggbsseg_sensitivity_National_",
  Sys.Date(),
  ".pdf"))
```



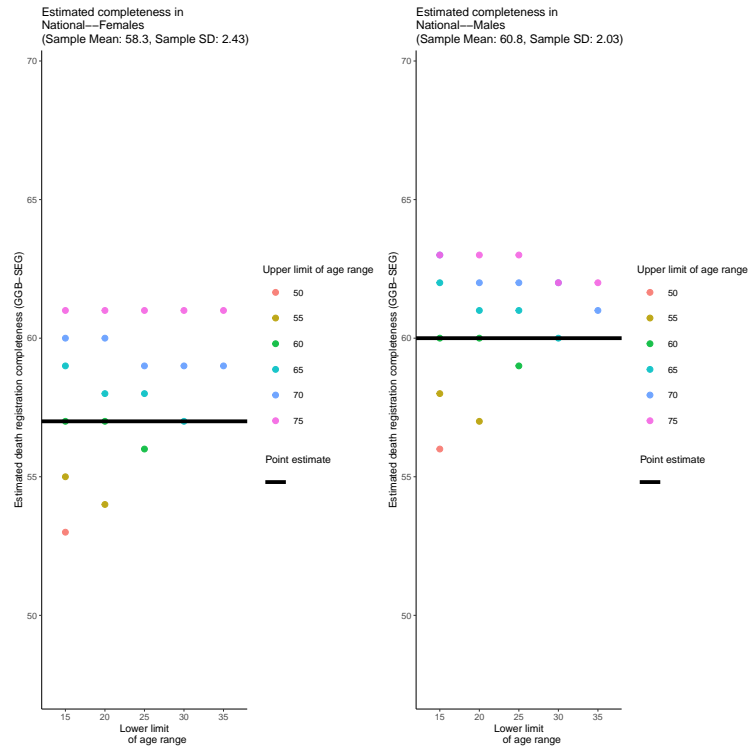


Figure 17: Sensitivity of point estimates of death registration completeness in Ecuador from 2001-2010 to choice of age-range parameter in the GGB-SEG method

```
knitr::include_graphics(path=paste0(my_plots_dir,
  "ggbseg_RMSE_National_",
  Sys.Date(),
  ".pdf"))
```

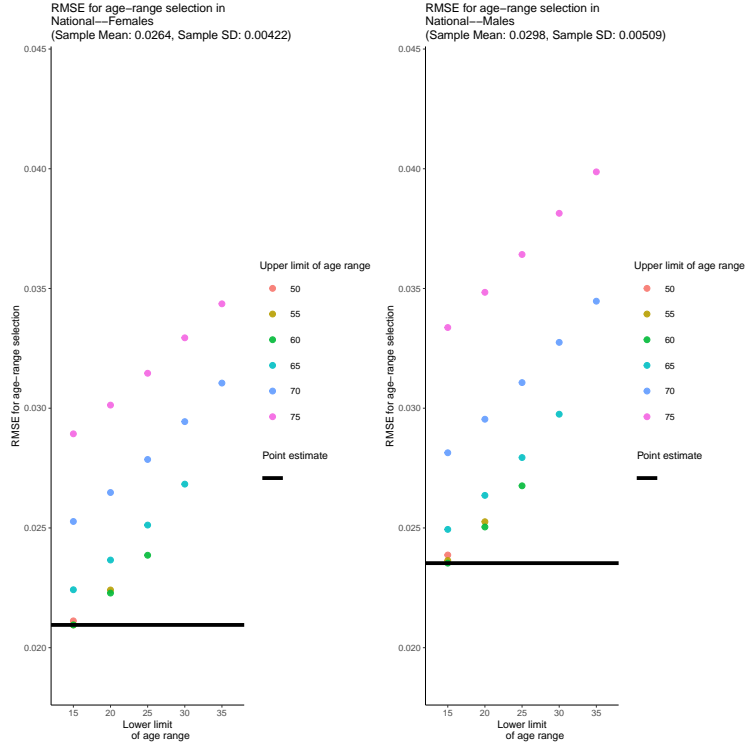


Figure 18: Sensitivity of point estimates of death registration completeness in Ecuador from 2001-2010 to choice of age-range parameter in the GGB-SEG method

## References

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- Riffe, T, JM Aburto, M Alexander, S Fennell, I Kashnitsky, M Pascariu, and P Gerland. 2019. “DemoTools: An R Package of Tools for Aggregate Demographic Analysis.”
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