

# SubnationalCRVS R Package: Demo

Jeremy Roth

## Contents

<b>1</b>	<b>Setup</b>	<b>1</b>
<b>2</b>	<b>View the Example Datasets Included with the SubnationalCRVS Package</b>	<b>2</b>
<b>3</b>	<b>Conduct Demographic Data Quality Assessment (DDQA)</b>	<b>3</b>
3.1	Sex ratio: PlotSexRatios() . . . . .	3
3.2	Age ratios: PlotAgeRatios() . . . . .	6
3.3	Potential age heaping: PlotPotentialAgeHeaping() . . . . .	8
3.4	Age heaping indices: PlotAgeHeapingScores() . . . . .	10
<b>4</b>	<b>DDM Estimation of Death Registration Completeness</b>	<b>12</b>
4.1	Compute DDM estimates: EstimateDDM . . . . .	13
4.2	Plot DDM estimates: PlotDDM() . . . . .	13

## 1 Setup

Since the SubnationalCRVS package is still under active development, it is hosted on my GitHub page ([www.github.com/jroth-unfpa/SubnationalCRVS](https://www.github.com/jroth-unfpa/SubnationalCRVS)) rather than on CRAN. As a result, SubnationalCRVS cannot be installed with the usual `install.packages()` function; instead, we install SubnationalCRVS with the `install.github()` function from the devtools package. The dependency DemoTools is also not hosted on CRAN and must be installed with `install.github()`.

```
library(devtools)
install_github("timriffe/DemoTools") # installing the DemoTools dependency (not on CRAN)
install_github("jroth-unfpa/SubnationalCRVS") # installing the SubnationalCRVS package itself
```

Now we can load SubnationalCRVS, specify the name of the local folder in which we will save the plots, 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
```

We will also load the dplyr package to customize the display of some of the tables returned by SubnationalCRVS.

```
library(dplyr)
```

## 2 View the Example Datasets Included with the Subnational-CRVS Package

The `SubnationalCRVS` package comes with two tabulations – disaggregated by age, sex, and province – created from publicly available datasets from Ecuador (LINKS), based on the country's 2001 Census, 2010 Census, and annual counts of registered deaths from 2001 through 2010.

```
head(ecuador_single_year_ages)
```

##	province_name	province_name_short	sex	age	pop1	pop2	date1	date2
## 1	Azuay	Azu	m	0	6086	6750	2001-11-25	2010-11-28
## 2	Azuay	Azu	m	1	6555	6984	2001-11-25	2010-11-28
## 3	Azuay	Azu	m	2	7232	7090	2001-11-25	2010-11-28
## 4	Azuay	Azu	m	3	7101	7095	2001-11-25	2010-11-28
## 5	Azuay	Azu	m	4	7083	6961	2001-11-25	2010-11-28
## 6	Azuay	Azu	m	5	6583	6895	2001-11-25	2010-11-28

The `ecuador_single_year_ages` 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 (full name in the `province` column and abbreviated name in the `province_name_short` column).

```
head(ecuador_five_year_ages)
```

##	province_name	province_name_short	sex	age	pop1	pop2	deaths	date1
## 1	Azuay	Azu	m	0	34101	34886	772	2001-11-25
## 2	Azuay	Azu	m	10	34946	38125	223	2001-11-25
## 3	Azuay	Azu	m	15	32387	37611	416	2001-11-25
## 4	Azuay	Azu	m	20	25634	33665	480	2001-11-25
## 5	Azuay	Azu	m	25	18606	28376	475	2001-11-25
## 6	Azuay	Azu	m	30	16193	22026	456	2001-11-25
##	date2							
## 1	2010-11-28							
## 2	2010-11-28							
## 3	2010-11-28							
## 4	2010-11-28							
## 5	2010-11-28							
## 6	2010-11-28							

The `ecuador_five_year_ages` dataset has the same variables as `ecuador_single_year_ages` 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.

## 3 Conduct Demographic Data Quality Assessment (DDQA)

### 3.1 Sex ratio: PlotSexRatios()

One step in the demographic data quality assessment (DDQA) process is to use the `PlotSexRatios()` function in the `SubnationalCRVS` package 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`.

We must provide our tabulated data frame in the `data` argument of `PlotSexRatios()` and, additionally, we need to provide a few additional arguments to `PlotSexRatios()` to describe the variable names and values in the dataset:

- `name.disaggregations` is the name of variable representing the subnational disaggregation (apart from sex) 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.date1` is the name of variable that provides the data of the earlier of the two time periods. Here we specify `name.date1="date1"` (the value of this variable, "2001-11-15" is the date of Ecuador's 2001 Census)
- `name.date2` is the name of variable that provides the data of the later of the two time periods. Here we specify `name.date2="date2"` (the value of this variable, "2010-11-28" is the date 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.year1="pop1"`
- `name.population.year2` is the name of variable representing the population in the later 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
- `name.female` is the name of value of the `name.sex` variable that represents females

In addition, we specify the option argument `plots.dir=my_plots_dir` so that the plots will save in a local folder `Plots`; the default behavior would be to save the plots in the working directory of the R script.

```
c_s <- ComputeSexRatios(data=ecuador_single_year_ages,
                        name.disaggregations="province_name",
                        name.males="m",
                        name.females="f",
                        name.age="age",
                        name.sex="sex",
                        name.date1="date1",
                        name.date2="date2",
                        name.population.year1="pop1",
                        name.population.year2="pop2")

s <- PlotSexRatios(data=ecuador_single_year_ages,
                  name.disaggregations="province_name",
                  name.males="m",
                  name.females="f",
                  name.age="age",
```

```

name.sex="sex",
name.date1="date1",
name.date2="date2",
name.population.year1="pop1",
name.population.year2="pop2",
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`.

### 3.1.1 View sex ratios in combined plot

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"))

```

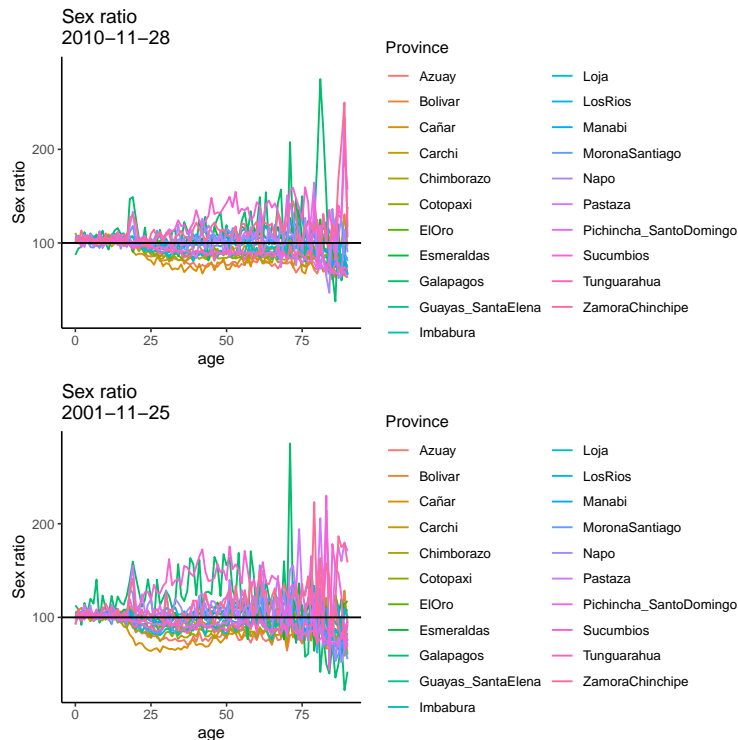


Figure 1: Sex ratios in Ecuador by province, combined plot

### 3.1.2 View sex ratios in disaggregated plots

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"))
```

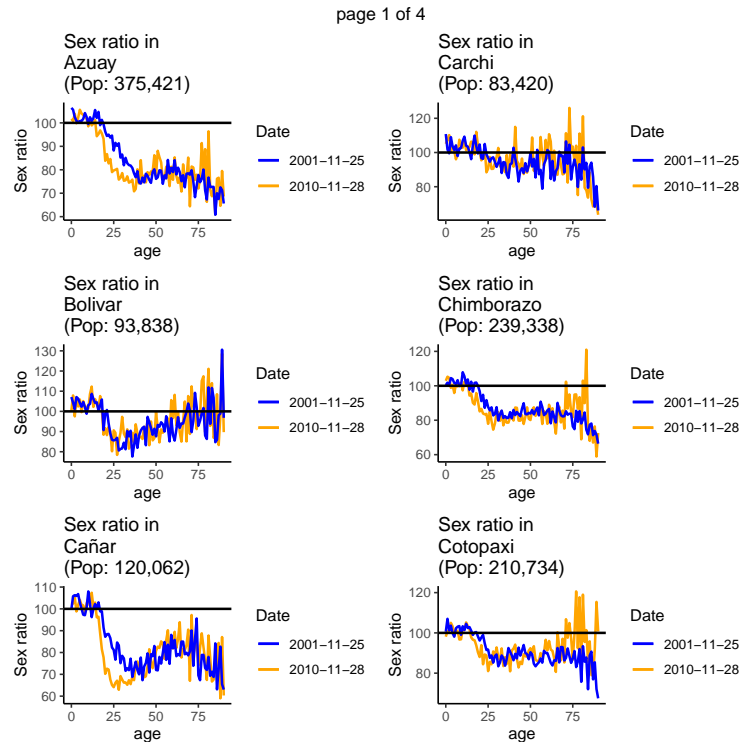


Figure 2: Sex ratios in Ecuador by province, disaggregated plots

### 3.1.3 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	5987	6342	101.7	106.4
## 2	Azuay	1	6505	6612	100.8	105.6
## 3	Azuay	2	6963	6924	103.9	102.4
## 4	Azuay	3	7116	7029	99.8	100.9
## 5	Azuay	4	6929	6950	102.2	100.2
## 6	Azuay	5	6232	6841	105.6	100.8

## 3.2 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`.

```
a <- PlotAgeRatios(data=ecuador_five_year_ages,
  name.disaggregations="province_name",
  name.males="m",
  name.females="f",
  name.age="age",
  name.sex="sex",
  name.date1="date1",
  name.date2="date2",
  name.population.year1="pop1",
  name.population.year2="pop2",
  label.subnational.level="Province",
  plots.dir="Plots/")
```

The arguments we provided 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`) instead of the tabulation with single-year ages.

### 3.2.1 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"))
```

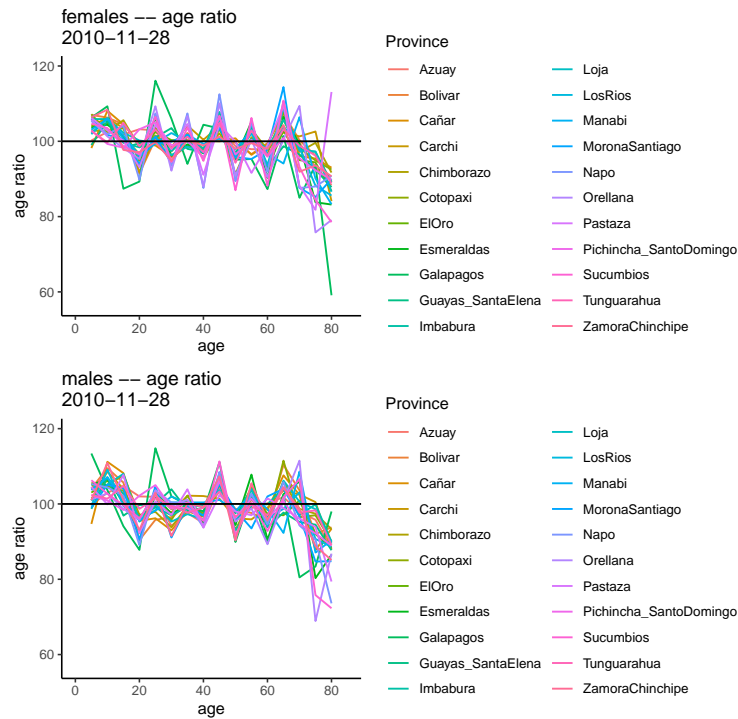


Figure 3: Age ratios in Ecuador by province, combined plot

### 3.2.2 View age ratios in disaggregated plots

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"))
```

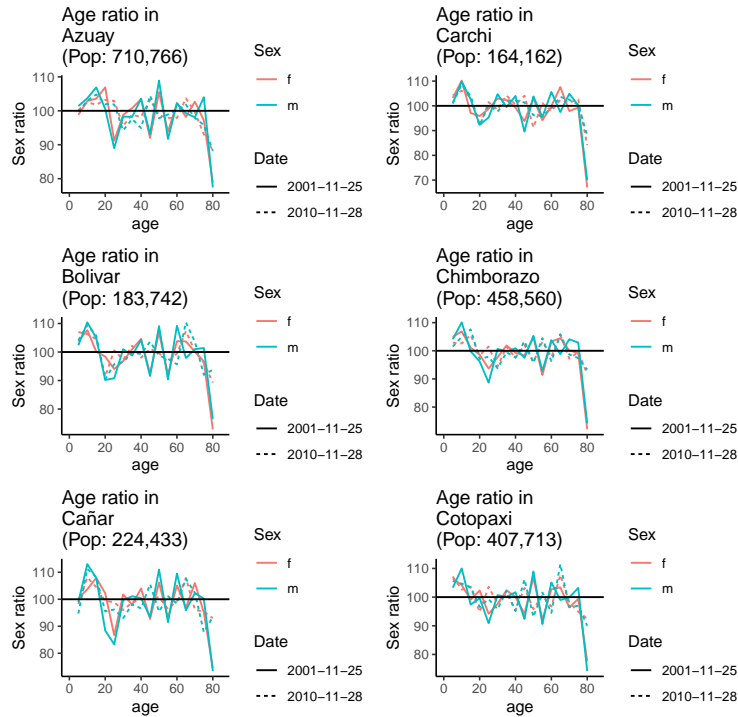


Figure 4: Age ratios in Ecuador by province, disaggregated plots

### 3.2.3 View age ratios in table

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

### 3.3 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` 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,
  name.disaggregations="province_name",
  name.males="m",
```



```

name.females="f",
name.age="age",
name.sex="sex",
name.date1="date1",
name.date2="date2",
name.population.year1="pop1",
name.population.year2="pop2",
label.subnational.level="Province",
plots.dir="Plots/")

```

### 3.3.1 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"))

```

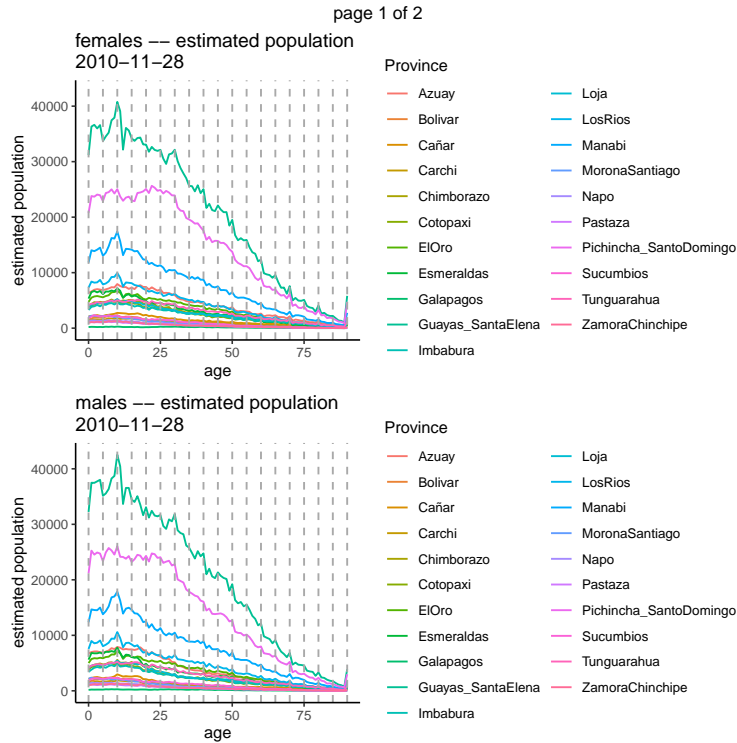


Figure 5: Population counts in Ecuador by single-year age, combined plot

### 3.3.2 View potential age heaping in disaggregated plots

`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"))
```

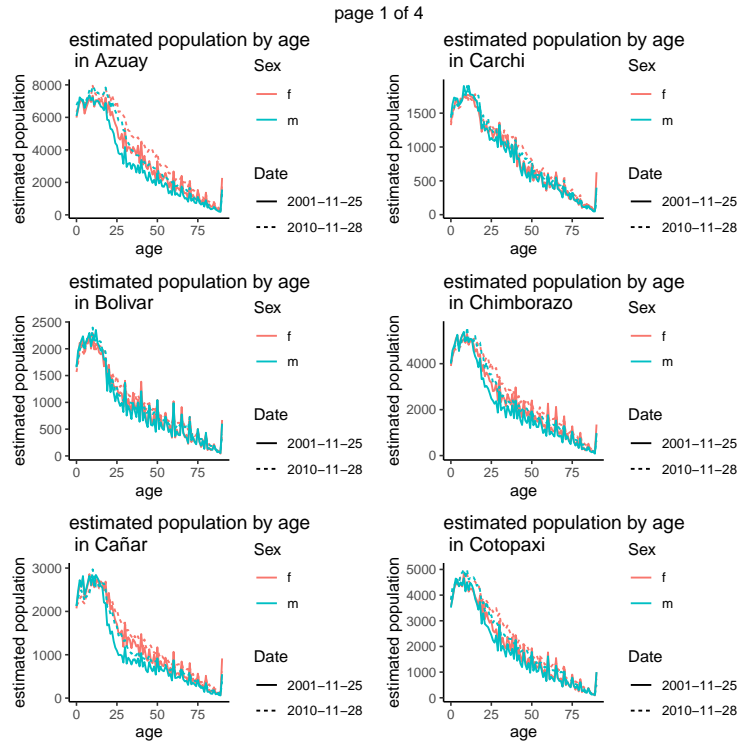


Figure 6: Population counts in Ecuador by single-year age, disaggregated plots

### 3.4 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,
                                   name.disaggregations="province_name",
                                   name.males="m",
                                   name.females="f",
                                   name.age="age",
                                   name.sex="sex",
                                   name.date1="date1",
                                   name.date2="date2",
                                   name.population.year1="pop1",
                                   name.population.year2="pop2",
                                   label.subnational.levels="Province",
                                   plots.dir="Plots/")
```

### 3.4.1 View age heaping indices in plots

`PlotAgeHeapingScores` plots the values of three age-heaping indices within each combination of province, sex, and data year: Roughness, Whipple, and Myers. The indices are computed with the `DemoTools::check_heaping_roughness`, `DemoTools::check_heaping_Whipple`, and `DemoTools::check_heaping_myers` functions, respectively, from the `DemoTools` package.

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

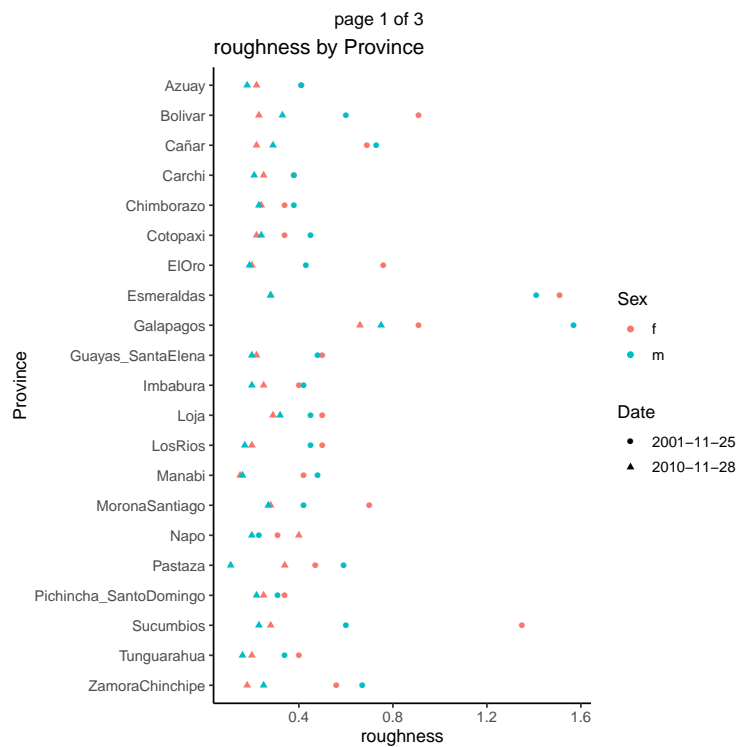


Figure 7: Age heaping indices in Ecuador by province

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

### 3.4.2 View age heaping indices in table

```
head(ageheaping)
```

##	province_name	date	sex	roughness	Whipple	Myers
## 1	Azuay	2001-11-25	f	0.41	1.18	4.21
## 2	Bolivar	2001-11-25	f	0.91	1.37	7.39
## 3	Cañar	2001-11-25	f	0.69	1.22	4.89
## 4	Carchi	2001-11-25	f	0.38	1.18	3.75
## 5	Chimborazo	2001-11-25	f	0.34	1.25	5.44
## 6	Cotopaxi	2001-11-25	f	0.34	1.27	5.99

### 3.4.3 View Noubissi age-heaping indices: ComputeAgeHeapingScores()

We can also use the `ComputeAgeHeapingScores` function – which is called within `PlotAgeHeapingScores` – with the optional argument `Noubissi.display=TRUE` to view the Noubissi index computed for single-year ages ending with 0, 1, 2, ..., 9.

```
ageheaping_with_Noubissi <- ComputeAgeHeapingScores(data=ecuador_single_year_ages,
  name.disaggregations="province_name",
  name.males="m",
  name.females="f",
  name.age="age",
  name.sex="sex",
  name.date1="date1",
  name.date2="date2",
  name.population.year1="pop1",
  name.population.year2="pop2",
  Noubissi.display=TRUE)

head(ageheaping_with_Noubissi)
```

```
##      date sex province_name roughness Whipple Myers Noubissi_0 Noubissi_1
## 1 2001-11-25 f      Azuay      0.41    1.18  4.21      1.18      0.91
## 2 2001-11-25 f    Bolivar      0.91    1.37  7.39      1.35      0.84
## 3 2001-11-25 f    Cañar      0.69    1.22  4.89      1.22      0.90
## 4 2001-11-25 f    Carchi      0.38    1.18  3.75      1.15      0.91
## 5 2001-11-25 f Chimborazo      0.34    1.25  5.44      1.23      0.88
## 6 2001-11-25 f    Cotopaxi      0.34    1.27  5.99      1.25      0.88
##  Noubissi_2 Noubissi_3 Noubissi_4 Noubissi_5 Noubissi_6 Noubissi_7
## 1      1.02      0.98      0.95      1.08      0.98      0.94
## 2      1.03      0.91      0.86      1.22      0.97      0.92
## 3      1.02      0.98      0.91      1.11      0.98      0.95
## 4      1.05      0.96      0.92      1.11      0.98      0.96
## 5      1.01      0.97      0.89      1.12      0.98      0.94
## 6      1.02      0.95      0.90      1.18      0.98      0.91
##  Noubissi_8 Noubissi_9
## 1      1.04      0.86
## 2      1.03      0.75
## 3      1.03      0.80
## 4      1.04      0.86
## 5      1.05      0.82
## 6      1.06      0.80
```

## 4 DDM Estimation of Death Registration Completeness

The structure of the `ecuador_five_year_ages` dataset is inspired by the requirements for the `DDM::ddm` function from the `DDM` package, which uses established Death Distribution Methods (DDM) to estimate death registration completeness between two consecutive Censuses. As a result, the `SubnationalCRVS` package offers the `EstimateDDM` function as a simple wrapper to `DDM::ddm` to perform DDM estimation of death registration completeness.

## 4.1 Compute DDM estimates: EstimateDDM

```
ddm_results <- EstimateDDM(data=ecuador_five_year_ages,  
  name.disaggregations="province_name",  
  name.age="age",  
  name.sex="sex",  
  name.males="m",  
  name.females="f",  
  name.date1="date1",  
  name.date2="date2",  
  name.population.year1="pop1",  
  name.population.year2="pop2",  
  name.deaths="deaths",  
  deaths.summed=TRUE)
```

```
## [1] "performing DDM estimation within each of 21 possible age ranges..."
```

We called `EstimateDDM` above using the same arguments as we used with `PlotAgeRatios`, with two additions:

- `name.deaths` provides the name of the variable represented 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.

## 4.2 Plot DDM estimates: PlotDDM()

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

```
PlotDDM(ddm_results=ddm_results,  
  label.completeness="Estimated Completeness (%)",  
  label.subnational.levels="Province",  
  plots.dir="Plots/")
```

### 4.2.1 View estimates of death registration completeness in plot

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

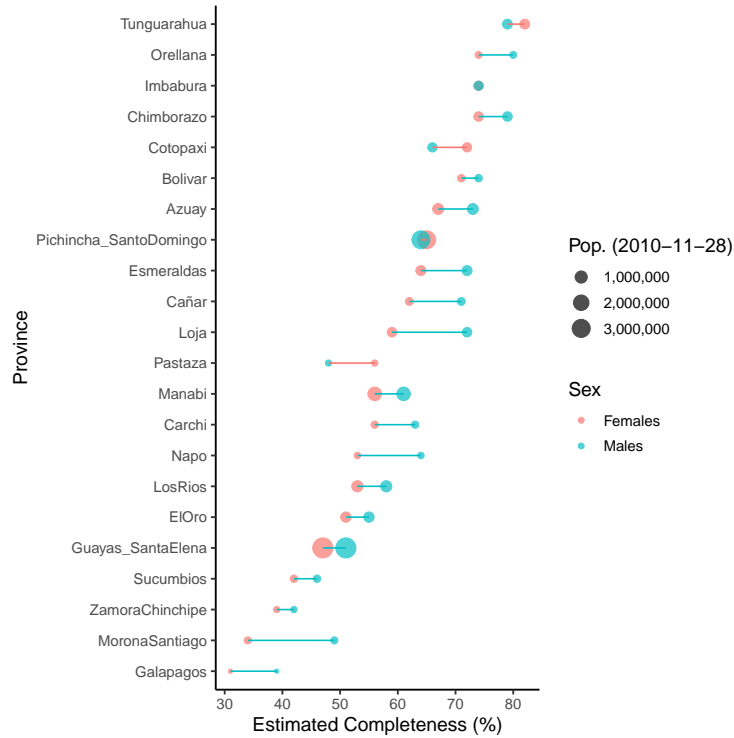


Figure 8: Point estimates of death registration completeness in Ecuador from 2001-2010, using the GGB-SEG method

#### 4.2.2 View DDM point estimates in table

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$ddm_estimates)
```

##	cod	sex	ggbseg	ggb	seg	lower_age_range	upper_age_range	total_pop1
## 1	Azuay	Females	0.67	0.99	0.81	15	50	599313
## 2	Azuay	Males	0.73	1.07	0.92	15	50	599313
## 3	Bolivar	Females	0.71	0.99	0.72	20	60	170696
## 4	Bolivar	Males	0.74	0.96	0.80	25	60	170696
## 5	Cañar	Females	0.62	1.00	0.58	20	55	206346
## 6	Cañar	Males	0.71	0.95	0.79	15	50	206346
##	total_pop2							
## 1	710766							
## 2	710766							
## 3	183742							
## 4	183742							
## 5	224433							
## 6	224433							

#### 4.2.3 View sensitivity of DDM point estimates in plot

Additionally, `PlotDDM` presents a visualization of the GGB-SEG estimates of death-registration completeness to the all permitted values of the age range (i.e. sensitivity) that is selected `DDM::ddm()` as part of the fitting

procedure underlying its estimation.

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

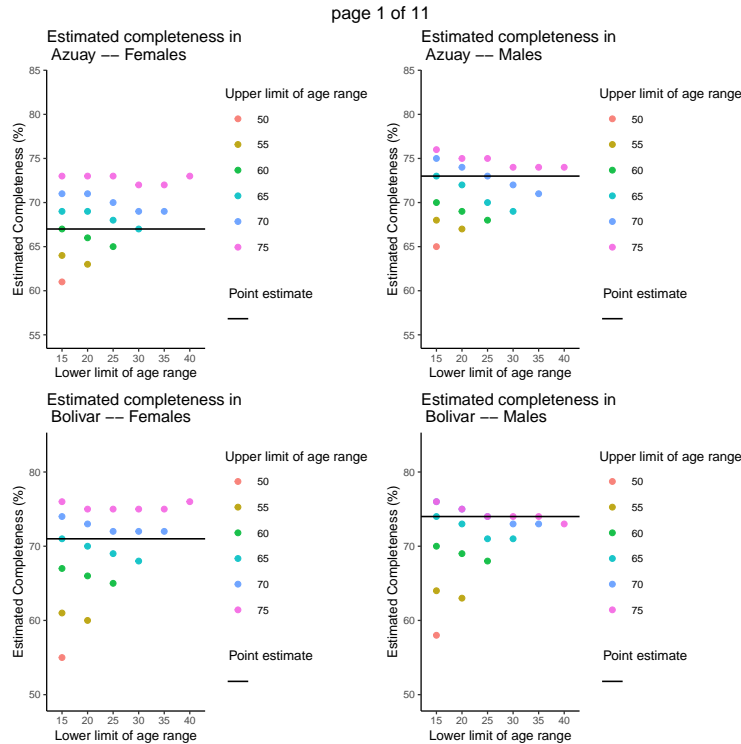


Figure 9: Sensitivity of point estimates of death registration completeness in Ecuador from 2001-2010 to choice of age-range parameter in the GGB-SEG method

#### 4.2.4 View age-range sensitivity of DDM point estimates in table

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

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

```
##      cod      sex ggbseg  ggb  seg lower_age_range upper_age_range total_pop1
## 1 Azuay Females  0.61 0.99 0.81          15          50      599313
## 2 Azuay Females  0.64 0.87 0.81          15          55      599313
## 3 Azuay Females  0.63 0.86 0.81          20          55      599313
## 4 Azuay Females  0.67 0.83 0.81          15          60      599313
## 5 Azuay Females  0.66 0.82 0.80          20          60      599313
## total_pop2
## 1      710766
## 2      710766
## 3      710766
```

## 4	710766
## 5	710766