

R Notebook for SAPE 2019

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Data set created by
Author: Bernhard Scheliga

```
## [1] "Dataset version: 0.1"
```

```
## [1] "Date: 2020-10-29"
```

```
## [1] "R version 4.0.3 (2020-10-10)"
```

1. Summary:

Initially, the plan was to include this data directly in the Scottish Vulnerability Resource (SVR). However, the Small Area Population Estimate (SAPE) 2019 data set for Scotland is a bit too large in itself to be included directly. The idea now is, that it is an additional resource to the SVR and we will provide a script for user to combine the SVR and the SVR_SAPE2019 dataset.

The SVR_SAPE2019 resource is postcode(PC) searchable and enables the user to swiftly retrieve data for datazones based on a postcode search. *!Data is not broken down to postcode level!* The spatial resolution of the resource are datazone. For more details see our in depth ***LINK TO THE RIGHT FILE*** documentation on github.com/AbdnCHDS/Scotland_Vulnerability_Resource

Here we describe how the SVR_SAPE2019 was created in *R*, which code and information was used. This is to enable the interested user to follow our process and get a better understanding of the decisions we during the creation process of the SVR_SAPE2019.

The SVR_SAPE2019 is build around the Small Area Population Estimate (SAPE) 2019 data set for Scotland. The current version of the SVR_SAPE 2019 include Female, Male and Both population estimates, the postcodes (PC) of the respective SIMD2020v2 data zones, their data zone names and the NHS Health board regions.

Keywords: *R script, SAPE 2019, SIMD2020v2, postcode searchable, datazone names, reproducible, open access*

2. Creating the data set

```
setwd("~/Scotland_Vulnerability_Resource/SVR-data/")  
dir()
```

2.1 Loading source data

```
## [1] "2019_Small_Area_Population_Estimates_FeMale_SVR.csv"
## [2] "Scotland-Vulnerability-Resource_v0.1.csv"
## [3] "Scotland-Vulnerability-Resource_v0.2.csv"

df_SVR <- read.csv("Scotland-Vulnerability-Resource_v0.2.csv") # we only want the first few columns

## The SAPE 2019 Data
setwd("~/Scotland_Vulnerability_Resource/Input-data/")
dir()

## [1] "Datazone_areas_sizes.csv"
## [2] "Input-data_documentation"
## [3] "NHS_Health_Board_regions.csv"
## [4] "sape-2019-females_Table 1c Females (2019).csv"
## [5] "sape-2019-males_Table 1b Males (2019).csv"
## [6] "SIMD2020v2datazones.csv"
## [7] "SIMD2020v2indicators.csv"
## [8] "SIMD2020v2indicators_desc.csv"
## [9] "SIMD2020v2postcodes.csv"

df_SAPE2019.Female <- read.csv("sape-2019-females_Table 1c Females (2019).csv", skip = 3)
df_SAPE2019.Male <- read.csv("sape-2019-males_Table 1b Males (2019).csv", skip = 3)
#df_SAPE2019.Person <- read.csv("sape-2019-persons_Table 1a Persons (2019).csv", skip = 3)

library("tidyverse")

## -- Attaching packages ----- tidyverse 1.3.0 --

## v ggplot2 3.3.2      v purrr 0.3.4
## v tibble 3.0.4       v dplyr 1.0.2
## v tidyr 1.1.2        v stringr 1.4.0
## v readr 1.4.0        v forcats 0.5.0

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

2.2 Cleaning source data

```
df_SAPE2019.Female %>% head()
```

2.2.1 Removing excess data from source data

```
##           X           X.1           X.2           X.3 X.4  AGE0
## 1                SCOTLAND                2,800,297  NA 24,634
## 2 DataZone2011Code DataZone2011Name CouncilArea209Name      NA
## 3      S01006506      Culter - 01      Aberdeen City      456  NA      1
```

## 4	S01006507	Culter - 02	Aberdeen City	417	NA	1					
## 5	S01006508	Culter - 03	Aberdeen City	292	NA	5					
## 6	S01006509	Culter - 04	Aberdeen City	281	NA	4					
##	AGE1	AGE2	AGE3	AGE4	AGE5	AGE6	AGE7	AGE8	AGE9	AGE10	AGE11
## 1	25,675	26,221	27,272	27,931	28,087	28,829	29,392	30,495	29,216	30,000	29,891
## 2											
## 3	2	5	0	2	7	5	5	12	11	6	6
## 4	4	4	4	4	3	3	6	5	3	4	11
## 5	3	3	0	2	1	0	3	1	0	2	1
## 6	1	2	2	3	3	2	7	2	2	6	2
##	AGE12	AGE13	AGE14	AGE15	AGE16	AGE17	AGE18	AGE19	AGE20	AGE21	AGE22
## 1	28,588	28,135	27,573	27,014	26,428	26,256	27,931	30,157	32,426	33,432	35,056
## 2											
## 3	1	7	8	5	5	5	1	4	0	2	1
## 4	4	8	0	1	1	2	3	0	2	2	6
## 5	0	3	0	3	1	1	1	1	1	3	4
## 6	5	6	5	5	3	3	2	1	0	0	1
##	AGE23	AGE24	AGE25	AGE26	AGE27	AGE28	AGE29	AGE30	AGE31	AGE32	AGE33
## 1	35,144	35,332	36,292	37,629	39,778	39,398	38,013	37,992	38,188	37,272	36,601
## 2											
## 3	6	2	5	6	11	6	12	1	6	6	1
## 4	5	6	4	8	9	7	6	3	3	3	3
## 5	8	5	6	10	7	4	10	7	11	2	3
## 6	1	0	4	1	0	5	5	8	4	2	1
##	AGE34	AGE35	AGE36	AGE37	AGE38	AGE39	AGE40	AGE41	AGE42	AGE43	AGE44
## 1	36,775	35,630	36,112	36,673	36,109	35,374	33,774	31,920	30,822	32,847	33,279
## 2											
## 3	15	7	3	9	7	9	9	5	3	4	5
## 4	5	1	4	5	4	11	6	4	10	6	5
## 5	7	5	5	2	3	5	2	4	5	2	6
## 6	2	5	5	3	1	5	9	2	6	4	3
##	AGE45	AGE46	AGE47	AGE48	AGE49	AGE50	AGE51	AGE52	AGE53	AGE54	AGE55
## 1	33,508	35,671	38,220	39,969	39,278	40,785	41,473	41,221	41,241	42,430	42,245
## 2											
## 3	7	10	10	4	7	9	5	3	2	8	8
## 4	2	5	1	6	2	3	7	5	5	9	5
## 5	2	5	10	2	5	1	7	4	7	6	2
## 6	5	4	6	3	2	4	2	1	2	2	4
##	AGE56	AGE57	AGE58	AGE59	AGE60	AGE61	AGE62	AGE63	AGE64	AGE65	AGE66
## 1	41,909	40,514	39,564	38,116	37,971	36,854	35,598	34,439	32,979	32,241	31,538
## 2											
## 3	8	5	9	5	7	7	6	3	3	5	3
## 4	6	11	4	4	14	4	4	2	7	6	4
## 5	8	7	5	5	6	1	3	1	5	3	5
## 6	5	5	0	3	4	2	4	5	4	0	6
##	AGE67	AGE68	AGE69	AGE70	AGE71	AGE72	AGE73	AGE74	AGE75	AGE76	AGE77
## 1	30,335	30,482	30,388	30,897	31,711	34,215	25,481	24,213	24,270	23,327	21,360
## 2											
## 3	5	5	0	8	7	9	2	4	9	4	3
## 4	6	10	9	5	10	9	8	1	6	0	2
## 5	5	1	3	2	2	2	3	1	0	0	3
## 6	8	2	9	3	4	5	0	1	1	2	3
##	AGE78	AGE79	AGE80	AGE81	AGE82	AGE83	AGE84	AGE85	AGE86	AGE87	AGE88
## 1	19,566	19,542	18,728	17,788	16,665	15,555	14,223	12,865	11,405	10,617	9,213

```
## 2
## 3      0      3      2      1      3      2      0      1      5      2      1
## 4      4      6      4      3      6      0      3      3      2      1      3
## 5      0      0      1      1      0      0      0      1      4      0      0
## 6      3      2      3      3      1      2      1      3      1      2      1
##   AGE89 AGE90. X.5 X.6
## 1 7,847 30,247 NA NA
## 2              NA NA
## 3      2      5 NA NA
## 4      0      1 NA NA
## 5      1      4 NA NA
## 6      1      4 NA NA
```

This is what all SAPE 2019 data set look like. The first two row and last two rows we do not need. Row 1 has the total popultation in Scotland for the respective gender and age group. Row 2 has only the header for for the first 3 columns. We will remove column X.1, X.2, X.3, X.4, X.5 & X.6

```
df_SAPE2019.Female <- df_SAPE2019.Female %>% select(-c(X.1, X.2, X.3, X.4, X.5, X.6)) %>% slice(-c(1,2,
df_SAPE2019.Male <- df_SAPE2019.Male %>% select(-c(X.1, X.2, X.3, X.4, X.5, X.6)) %>% slice(-c(1,2,6979
#df_SAPE2019.Person <- df_SAPE2019.Person %>% select(-c(X.1, X.2, X.3, X.4, X.5, X.6)) %>% slice(-c(1,2
```

We need to change two column names, see below

```
colnames(df_SAPE2019.Female)[c(1,92)]
```

```
## [1] "X"      "AGE90."
```

```
colnames(df_SAPE2019.Female)[c(1,92)] <- c("Data_Zone","AGE90PLUS")
colnames(df_SAPE2019.Male)[c(1,92)] <- c("Data_Zone","AGE90PLUS")
#colnames(df_SAPE2019.Person)[c(1,92)] <- c("Data_Zone","AGE90PLUS")

colnames(df_SAPE2019.Female)[c(1,92)]
```

```
## [1] "Data_Zone" "AGE90PLUS"
```

Better.

Adding a gender column This is done to distinglish the data later once it is merged

```
df_SAPE2019.Female$Gender <- "Female"
df_SAPE2019.Male$Gender <- "Male"
#df_SAPE2019.Person$Gender <- "Person"
```

3. Joining the Datasets We actually just need to use `bind_rows()` the datasets. **Note!**: We only joined the female & male data since the all three datasets together would be to big for GitHub.

```
df_SAPE2019 <- bind_rows(df_SAPE2019.Female,df_SAPE2019.Male)
```

Now, we have all three SAPE 2019 data set in one object. The next step is to take the *Postcode* -, *Data_zone* -, *Intermediate_Zone* -, *Council_area* - & *NHS_Health_Board_Region* - columns from the SVR dataset and join them to the SAPE 2019 (*df_SAPE2019*).

```
df_SAPE2019.SVRfront <- df_SVR %>% select(Postcode, Data_Zone, Intermediate_Zone,Council_area, NHS_Heal
# quick reordering of the columns
df_SAPE2019.SVRfront <- df_SAPE2019.SVRfront[,c(1:5,97,6:96)]

## Check for NA
sapply(df_SAPE2019.SVRfront, function(x) sum(is.na(x)))
```

##	Postcode	Data_Zone	Intermediate_Zone
##	4	0	0
##	Council_area NHS_Health_Board_Region		Gender
##	0	0	0
##	AGE0	AGE1	AGE2
##	0	0	0
##	AGE3	AGE4	AGE5
##	0	0	0
##	AGE6	AGE7	AGE8
##	0	0	0
##	AGE9	AGE10	AGE11
##	0	0	0
##	AGE12	AGE13	AGE14
##	0	0	0
##	AGE15	AGE16	AGE17
##	0	0	0
##	AGE18	AGE19	AGE20
##	0	0	0
##	AGE21	AGE22	AGE23
##	0	0	0
##	AGE24	AGE25	AGE26
##	0	0	0
##	AGE27	AGE28	AGE29
##	0	0	0
##	AGE30	AGE31	AGE32
##	0	0	0
##	AGE33	AGE34	AGE35
##	0	0	0
##	AGE36	AGE37	AGE38
##	0	0	0
##	AGE39	AGE40	AGE41
##	0	0	0
##	AGE42	AGE43	AGE44
##	0	0	0
##	AGE45	AGE46	AGE47
##	0	0	0
##	AGE48	AGE49	AGE50
##	0	0	0
##	AGE51	AGE52	AGE53
##	0	0	0
##	AGE54	AGE55	AGE56
##	0	0	0
##	AGE57	AGE58	AGE59
##	0	0	0
##	AGE60	AGE61	AGE62
##	0	0	0

##	AGE63	AGE64	AGE65
##	0	0	0
##	AGE66	AGE67	AGE68
##	0	0	0
##	AGE69	AGE70	AGE71
##	0	0	0
##	AGE72	AGE73	AGE74
##	0	0	0
##	AGE75	AGE76	AGE77
##	0	0	0
##	AGE78	AGE79	AGE80
##	0	0	0
##	AGE81	AGE82	AGE83
##	0	0	0
##	AGE84	AGE85	AGE86
##	0	0	0
##	AGE87	AGE88	AGE89
##	0	0	0
##	AGE90PLUS		
##	0		

6 NA in Postcode. That will be the same NA postcodes as in the SVR data (Petershill & Sighthill) just time three

```
df_SAPE2019.SVRfront[is.na(df_SAPE2019.SVRfront$Postcode),]
```

##	Postcode	Data_Zone	Intermediate_Zone	Council_area									
##	169411	<NA>	S01010206	Petershill	Glasgow City								
##	169412	<NA>	S01010206	Petershill	Glasgow City								
##	170199	<NA>	S01010226	Sighthill	Glasgow City								
##	170200	<NA>	S01010226	Sighthill	Glasgow City								
##		NHS_Health_Board_Region	Gender	AGE0	AGE1	AGE2	AGE3	AGE4	AGE5	AGE6	AGE7		
##	169411	Greater Glasgow and Clyde	Female	0	0	0	0	0	0	0	0		
##	169412	Greater Glasgow and Clyde	Male	0	0	0	0	0	0	0	0		
##	170199	Greater Glasgow and Clyde	Female	0	0	0	0	0	0	0	0		
##	170200	Greater Glasgow and Clyde	Male	0	0	0	0	0	0	0	0		
##		AGE8	AGE9	AGE10	AGE11	AGE12	AGE13	AGE14	AGE15	AGE16	AGE17	AGE18	AGE19
##	169411	0	0	0	0	0	0	0	0	0	0	0	0
##	169412	0	0	0	0	0	0	0	0	0	0	0	0
##	170199	0	0	0	0	0	0	0	0	0	0	0	0
##	170200	0	0	0	0	0	0	0	0	0	0	0	0
##		AGE20	AGE21	AGE22	AGE23	AGE24	AGE25	AGE26	AGE27	AGE28	AGE29	AGE30	AGE31
##	169411	0	0	0	0	0	0	0	0	0	0	0	0
##	169412	0	0	0	0	0	0	0	0	0	0	0	0
##	170199	0	0	0	0	0	0	0	0	0	0	0	0
##	170200	0	0	0	0	0	0	0	0	0	0	0	0
##		AGE32	AGE33	AGE34	AGE35	AGE36	AGE37	AGE38	AGE39	AGE40	AGE41	AGE42	AGE43
##	169411	0	0	0	0	0	0	0	0	0	0	0	0
##	169412	0	0	0	0	0	0	0	0	0	0	0	0
##	170199	0	0	0	0	0	0	0	0	0	0	0	0
##	170200	0	0	0	0	0	0	0	0	0	0	0	0
##		AGE44	AGE45	AGE46	AGE47	AGE48	AGE49	AGE50	AGE51	AGE52	AGE53	AGE54	AGE55
##	169411	0	0	0	0	0	0	0	0	0	0	0	0

```
## 169412    0    0    0    0    0    0    0    0    0    0    0    0    0
## 170199    0    0    0    0    0    0    0    0    0    0    0    0    0
## 170200    0    0    0    0    0    0    0    0    0    0    0    0    0
##          AGE56 AGE57 AGE58 AGE59 AGE60 AGE61 AGE62 AGE63 AGE64 AGE65 AGE66 AGE67
## 169411    0    0    0    0    0    0    0    0    0    0    0    0    0
## 169412    0    0    0    0    0    0    0    0    0    0    0    0    0
## 170199    0    0    0    0    0    0    0    0    0    0    0    0    0
## 170200    0    0    0    0    0    0    0    0    0    0    0    0    0
##          AGE68 AGE69 AGE70 AGE71 AGE72 AGE73 AGE74 AGE75 AGE76 AGE77 AGE78 AGE79
## 169411    0    0    0    0    0    0    0    0    0    0    0    0    0
## 169412    0    0    0    0    0    0    0    0    0    0    0    0    0
## 170199    0    0    0    0    0    0    0    0    0    0    0    0    0
## 170200    0    0    0    0    0    0    0    0    0    0    0    0    0
##          AGE80 AGE81 AGE82 AGE83 AGE84 AGE85 AGE86 AGE87 AGE88 AGE89 AGE90PLUS
## 169411    0    0    0    0    0    0    0    0    0    0    0    0
## 169412    0    0    0    0    0    0    0    0    0    0    0    0
## 170199    0    0    0    0    0    0    0    0    0    0    0    0
## 170200    0    0    0    0    0    0    0    0    0    0    0    0
```

We want to check now, if there are only *numeric* value in the *AGE*-columns. By changing the value type from *character* to *numeric* in those columns non-numeric values should turn into NA-values.

```
df_SAPE2019.SVRfront[,c(7:97)] <- sapply(df_SAPE2019.SVRfront[,c(7:97)], as.numeric) # Couldn't find a
df_SAPE2019.SVRfront%>%select(AGE0:AGE90PLUS)%>%sapply(function(x) sum(is.na(x)))
```

```
##          AGE0          AGE1          AGE2          AGE3          AGE4          AGE5          AGE6          AGE7
##          0            0            0            0            0            0            0            0
##          AGE8          AGE9          AGE10         AGE11         AGE12         AGE13         AGE14         AGE15
##          0            0            0            0            0            0            0            0
##          AGE16         AGE17         AGE18         AGE19         AGE20         AGE21         AGE22         AGE23
##          0            0            0            0            0            0            0            0
##          AGE24         AGE25         AGE26         AGE27         AGE28         AGE29         AGE30         AGE31
##          0            0            0            0            0            0            0            0
##          AGE32         AGE33         AGE34         AGE35         AGE36         AGE37         AGE38         AGE39
##          0            0            0            0            0            0            0            0
##          AGE40         AGE41         AGE42         AGE43         AGE44         AGE45         AGE46         AGE47
##          0            0            0            0            0            0            0            0
##          AGE48         AGE49         AGE50         AGE51         AGE52         AGE53         AGE54         AGE55
##          0            0            0            0            0            0            0            0
##          AGE56         AGE57         AGE58         AGE59         AGE60         AGE61         AGE62         AGE63
##          0            0            0            0            0            0            0            0
##          AGE64         AGE65         AGE66         AGE67         AGE68         AGE69         AGE70         AGE71
##          0            0            0            0            0            0            0            0
##          AGE72         AGE73         AGE74         AGE75         AGE76         AGE77         AGE78         AGE79
##          0            0            0            0            0            0            0            0
##          AGE80         AGE81         AGE82         AGE83         AGE84         AGE85         AGE86         AGE87
##          0            0            0            0            0            0            0            0
##          AGE88         AGE89 AGE90PLUS
##          0            0            0
```

No NA-values, perfect!

```
setwd("~/Scotland_Vulnerability_Resource/SVR-data/")  
write.csv(df_SAPE2019.SVRfront, paste("2019_Small_Area_Population_Estimates_FeMale_SVR", ".csv", sep = "
```

2.4 Saving the data set