

Calibration adjustments to address bias in mortality analyses due to informative sampling - a census-linked survey analysis in Switzerland

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To ensure reproducibility of our main results, we provide an anonymised dataset (comma-separated file "SwissCensus2010.csv") of the 2010 study population (STATPOP and SE). Because of privacy preservation of study participants we were not allowed to deliver exact age information - as in the original analysis -, but only rounded to one decimal digit. We highlight that rounding had no influence on the interpretation of our results (i.e., mortality rates changed at maximum at the first decimal place), compared to exact information. IP weights were constructed in Stata, as described in a separate code.

We provide data and analysis code for Tables 1 and 2, and Figure 1 and 2, in the main manuscript. Estimates for life expectancy (Figure 2) are shown as Stata output, using the same dataset (Swiss-Census2010.csv).

Variables used:

- female: Gender (0: Male; 1 Female)
- agestart_rounded: Age at December 31, 2010, rounded to one decimal place
- ageend_rounded: Age at December 31, 2011, or death, rounded to one decimal place
- statpop_agecat5: 15 age categories (0: [15,20); 1: [20,25); ...), at December 31, 2010
- civil: Civil status (1: Single; 2: Married; 3: Widowed; 4: Other)
- nat: Nationality (1: Swiss; 2: EEA; 3: Other Europe; 4: Other World)
- canton: 29 regions
- died2011: Individual died between December 31, 2010, and December 31, 2011
- inse2010: Binary indicator for "being sampled" in structural enquiry (SE) 2010 (0: Not sampled; 1: Sampled)
- fupt_rounded: Follow-up time (in years), rounded to two decimal places
- weight2010: Calibrated survey weights (CSW)
- ipw2010: Inverse probability weights (IPW)

```
library(reporttools)
library(rms)
library(ggplot2)

data <- read.csv( paste(path, "SwissCensus2010.csv", sep=""))
# Ordering of columns
data <- data[, c("female", "agestart_rounded", "ageend_rounded",
                "statpop_agecat5", "civil", "nat", "canton",
                "died2011", "inse2010", "fupt_rounded",
                "weight2010", "ipw2010")]
```

The dataset contains 12 variables,

```
head(data[data$died2011==1 & data$inse2010==1, ], 3)
```

```
##      female agestart_rounded ageend_rounded statpop_agecat5 civil nat
## 1896      0          79.3          79.8           12      2      1
## 10821     1          96.4          97.1           14      3      1
## 12955     0          47.1          47.5            6      4      2
##      canton died2011 inse2010 fupt_rounded weight2010 ipw2010
## 1896      23         1         1           0.5  32.89608 41.50185
## 10821     12         1         1           0.7  28.83645 62.33053
## 12955     21         1         1           0.4  18.05437 23.09814
```

with a total of 6,729,363 individuals (STATPOP 2010),

```
nrow(data)
```

```
## [1] 6729363
```

and 317,079 individuals sampled in the structural enquiry (SE2010).

```
nrow(data[data$inse2010==1, ])
```

```
## [1] 317079
```

Table 1

```
tableNominal(vars=data[,c("female", "statpop_agecat5", "civil", "nat",
                           "canton")], group=data$died2011, cumsum=F,
              cap="STATPOP 2010: Categorical variables")
```

Variable	Levels	n ₀	% ₀	n ₁	% ₁	n _{all}	% _{all}
female	0	3270394	49.0	29804	48.4	3300198	49.0
	1	3397430	51.0	31735	51.6	3429165	51.0
	all	6667824	100.0	61539	100.0	6729363	100.0
statpop_agecat5	0	453418	6.8	125	0.2	453543	6.7
	1	497932	7.5	195	0.3	498127	7.4
	2	535327	8.0	179	0.3	535506	8.0
	3	542177	8.1	250	0.4	542427	8.1
	4	562531	8.4	330	0.5	562861	8.4
	5	636715	9.6	662	1.1	637377	9.5
	6	651880	9.8	1087	1.8	652967	9.7
	7	565305	8.5	1525	2.5	566830	8.4
	8	484779	7.3	2048	3.3	486827	7.2
	9	456872	6.8	3110	5.0	459982	6.8
	10	393251	5.9	4259	6.9	397510	5.9
	11	298540	4.5	5082	8.3	303622	4.5
	12	248981	3.7	7565	12.3	256546	3.8
	13	184460	2.8	10711	17.4	195171	2.9
	14	155656	2.3	24411	39.7	180067	2.7
all		6667824	100.0	61539	100.0	6729363	100.0
civil	1	2224776	33.4	7455	12.1	2232231	33.2
	2	3449816	51.7	24937	40.5	3474753	51.6

	3	387352	5.8	23109	37.5	410461	6.1
	4	605880	9.1	6038	9.8	611918	9.1
	all	6667824	100.0	61539	100.0	6729363	100.0
nat	1	5147117	77.2	56418	91.7	5203535	77.3
	2	989417	14.8	4309	7.0	993726	14.8
	3	317233	4.8	538	0.9	317771	4.7
	4	214057	3.2	274	0.4	214331	3.2
	all	6667824	100.0	61539	100.0	6729363	100.0
canton	1	837660	12.6	6681	10.9	844341	12.6
	2	689868	10.3	7072	11.5	696940	10.4
	3	318109	4.8	2750	4.5	320859	4.8
	4	29375	0.4	315	0.5	29690	0.4
	5	122585	1.8	1053	1.7	123638	1.8
	6	29675	0.4	249	0.4	29924	0.4
	7	34471	0.5	270	0.4	34741	0.5
	8	32785	0.5	364	0.6	33149	0.5
	9	94883	1.4	704	1.1	95587	1.4
	10	228838	3.4	1948	3.2	230786	3.4
	11	216854	3.2	2161	3.5	219015	3.2
	12	162021	2.4	2026	3.3	164047	2.4
	13	234029	3.5	2195	3.6	236224	3.5
	14	65688	1.0	713	1.2	66401	1.0
	15	43940	0.7	444	0.7	44384	0.7
	16	12738	0.2	135	0.2	12873	0.2
	17	405528	6.1	3657	5.9	409185	6.1
	18	174072	2.6	1615	2.6	175687	2.6
	19	513802	7.7	4350	7.1	518152	7.7
	20	209311	3.1	1881	3.1	211192	3.1
	21	284417	4.3	2856	4.6	287273	4.3
	22	603119	9.1	5165	8.4	608284	9.0
	23	265481	4.0	2492	4.0	267973	4.0
	24	143272	2.1	1501	2.4	144773	2.1
	25	365634	5.5	3013	4.9	368647	5.5
	26	57600	0.9	642	1.0	58242	0.9
	27	43374	0.6	518	0.8	43892	0.6
	28	113968	1.7	1331	2.2	115299	1.7
	29	334727	5.0	3438	5.6	338165	5.0
	all	6667824	100.0	61539	100.0	6729363	100.0

Table 1: STATPOP 2010: Categorical variables

```
tableNominal(vars=data[,c("female", "statpop_agecat5", "civil", "nat",
                          "canton", "inse2010")], group=data$died2011,
             data$inse2010==1, cumsum=F, cap="SE 2010: Categorical variables")
```

Variable	Levels	n ₀	% ₀	n ₁	% ₁	n _{all}	% _{all}
female	0	151684	48.1	1099	55.8	152783	48.2
	1	163424	51.9	872	44.2	164296	51.8
	all	315108	100.0	1971	100.0	317079	100.0
statpop_agecat5	0	19890	6.3	6	0.3	19896	6.3
	1	20467	6.5	10	0.5	20477	6.5
	2	23935	7.6	3	0.1	23938	7.5
	3	26445	8.4	8	0.4	26453	8.3
	4	27456	8.7	11	0.6	27467	8.7
	5	30282	9.6	21	1.1	30303	9.6
	6	31172	9.9	42	2.1	31214	9.8

	7	26896	8.5	52	2.6	26948	8.5
	8	23522	7.5	70	3.5	23592	7.4
	9	22204	7.0	122	6.2	22326	7.0
	10	19730	6.3	180	9.1	19910	6.3
	11	15000	4.8	163	8.3	15163	4.8
	12	12520	4.0	300	15.2	12820	4.0
	13	9040	2.9	323	16.4	9363	3.0
	14	6549	2.1	660	33.5	7209	2.3
	all	315108	100.0	1971	100.0	317079	100.0
civil	1	101868	32.3	189	9.6	102057	32.2
	2	166452	52.8	999	50.7	167451	52.8
	3	17942	5.7	592	30.0	18534	5.8
	4	28846	9.2	191	9.7	29037	9.2
	all	315108	100.0	1971	100.0	317079	100.0
nat	1	249184	79.1	1779	90.3	250963	79.2
	2	45200	14.3	172	8.7	45372	14.3
	3	12503	4.0	14	0.7	12517	4.0
	4	8221	2.6	6	0.3	8227	2.6
	all	315108	100.0	1971	100.0	317079	100.0
canton	1	24948	7.9	119	6.0	25067	7.9
	2	21957	7.0	130	6.6	22087	7.0
	3	19208	6.1	96	4.9	19304	6.1
	4	976	0.3	5	0.2	981	0.3
	5	3589	1.1	19	1.0	3608	1.1
	6	828	0.3	4	0.2	832	0.3
	7	1039	0.3	6	0.3	1045	0.3
	8	914	0.3	7	0.4	921	0.3
	9	5205	1.6	23	1.2	5228	1.6
	10	6570	2.1	39	2.0	6609	2.1
	11	6598	2.1	54	2.7	6652	2.1
	12	5113	1.6	26	1.3	5139	1.6
	13	7366	2.3	44	2.2	7410	2.3
	14	2001	0.6	18	0.9	2019	0.6
	15	1382	0.4	15	0.8	1397	0.4
	16	381	0.1	2	0.1	383	0.1
	17	12174	3.9	57	2.9	12231	3.9
	18	5070	1.6	32	1.6	5102	1.6
	19	30046	9.5	164	8.3	30210	9.5
	20	12137	3.8	73	3.7	12210	3.8
	21	17343	5.5	123	6.2	17466	5.5
	22	34465	10.9	232	11.8	34697	10.9
	23	7112	2.3	49	2.5	7161	2.3
	24	9295	3.0	73	3.7	9368	3.0
	25	20510	6.5	112	5.7	20622	6.5
	26	3723	1.2	30	1.5	3753	1.2
	27	2899	0.9	22	1.1	2921	0.9
	28	13139	4.2	111	5.6	13250	4.2
	29	39120	12.4	286	14.5	39406	12.4
	all	315108	100.0	1971	100.0	317079	100.0
inse2010	1	315108	100.0	1971	100.0	317079	100.0
	all	315108	100.0	1971	100.0	317079	100.0

Table 2: SE 2010: Categorical variables

```
tableContinuous(vars=data[,c("weight2010", "ipw2010")], group=data$died2011,
cap="STATPOP 2010: Continuous variables")
```

Variable	Levels	n	Min	q ₁	\tilde{x}	\bar{x}	q ₃	Max	s	IQR	#NA
weight2010	0	315108	3.3	15.0	16.8	20.6	30.3	75.0	9.4	15.3	6352716
	1	1971	6.4	14.6	16.1	19.4	30.1	43.4	9.3	15.5	59568
	all	317079	3.3	15.0	16.8	20.6	30.3	75.0	9.4	15.3	6412284
ipw2010	0	315108	7.3	15.2	17.6	21.2	30.2	61.8	9.6	15.0	6352716
	1	1971	5.8	17.0	26.0	30.2	39.3	113.4	17.3	22.3	59568
	all	317079	5.8	15.2	17.6	21.2	30.2	113.4	9.7	15.0	6412284

Table 3: STATPOP 2010: Continuous variables

```
tableContinuous(vars=data[,c("weight2010", "ipw2010")], group=data$died2011,
subset=data$inse2010==1, cap="SE 2010: Continuous variables")
```

Variable	Levels	n	Min	q ₁	\tilde{x}	\bar{x}	q ₃	Max	s	IQR	#NA
weight2010	0	315108	3.3	15.0	16.8	20.6	30.3	75.0	9.4	15.3	0
	1	1971	6.4	14.6	16.1	19.4	30.1	43.4	9.3	15.5	0
	all	317079	3.3	15.0	16.8	20.6	30.3	75.0	9.4	15.3	0
ipw2010	0	315108	7.3	15.2	17.6	21.2	30.2	61.8	9.6	15.0	0
	1	1971	5.8	17.0	26.0	30.2	39.3	113.4	17.3	22.3	0
	all	317079	5.8	15.2	17.6	21.2	30.2	113.4	9.7	15.0	0

Table 4: SE 2010: Continuous variables

Table 2

```
output <- c()

mod <- glm(inse2010 ~ factor(statpop_agecat5), data=data, family = binomial())

output <- rbind(output,
  data.frame(var=names(mod$coefficients[2:length(mod$coefficients)]),
    est=exp(mod$coefficients[2:length(mod$coefficients)]),
    lci=exp(confint.default(mod)[2:length(mod$coefficients),1]),
    uci=exp(confint.default(mod)[2:length(mod$coefficients),2]))
rm(mod)

mod <- glm(inse2010 ~ female, data=data, family = binomial())

output <- rbind(output,
  data.frame(var=names(mod$coefficients[2:length(mod$coefficients)]),
    est=exp(mod$coefficients[2:length(mod$coefficients)]),
    lci=exp(confint.default(mod)[2:length(mod$coefficients),1]),
    uci=exp(confint.default(mod)[2:length(mod$coefficients),2]))
rm(mod)

mod <- glm(inse2010 ~ factor(nat), data=data, family = binomial())

output <- rbind(output,
```

```

        data.frame(var=names(mod$coefficients[2:length(mod$coefficients)]),
                  est=exp(mod$coefficients[2:length(mod$coefficients)]),
                  lci=exp(confint.default(mod)[2:length(mod$coefficients),1]),
                  uci=exp(confint.default(mod)[2:length(mod$coefficients),2]))
rm(mod)

mod <- glm(inse2010 ~ factor(civil), data=data, family = binomial())

output <- rbind(output,
                data.frame(var=names(mod$coefficients[2:length(mod$coefficients)]),
                          est=exp(mod$coefficients[2:length(mod$coefficients)]),
                          lci=exp(confint.default(mod)[2:length(mod$coefficients),1]),
                          uci=exp(confint.default(mod)[2:length(mod$coefficients),2]))
rm(mod)

mod <- glm(inse2010 ~ factor(canton), data=data, family = binomial())

output <- rbind(output,
                data.frame(var=names(mod$coefficients[2:length(mod$coefficients)]),
                          est=exp(mod$coefficients[2:length(mod$coefficients)]),
                          lci=exp(confint.default(mod)[2:length(mod$coefficients),1]),
                          uci=exp(confint.default(mod)[2:length(mod$coefficients),2]))
rm(mod)

mod <- glm(inse2010 ~ died2011, data=data, family = binomial())

output <- rbind(output,
                data.frame(var=names(mod$coefficients[2:length(mod$coefficients)]),
                          est=exp(mod$coefficients[2:length(mod$coefficients)]),
                          lci=exp(confint.default(mod)[2:length(mod$coefficients),1]),
                          uci=exp(confint.default(mod)[2:length(mod$coefficients),2]))
rm(mod)

dimnames(output)[[1]] <- 1:length(dimnames(output)[[1]])

output

```

```

##          var      est      lci      uci
## 1 factor(statpop_agecat5)1 0.9343878 0.9159445 0.9532024
## 2 factor(statpop_agecat5)2 1.0198941 1.0004681 1.0396972
## 3 factor(statpop_agecat5)3 1.1174232 1.0965803 1.1386622
## 4 factor(statpop_agecat5)4 1.1181712 1.0974791 1.1392534
## 5 factor(statpop_agecat5)5 1.0879641 1.0682335 1.1080591
## 6 factor(statpop_agecat5)6 1.0942136 1.0744830 1.1143065
## 7 factor(statpop_agecat5)7 1.0879232 1.0677159 1.1085130
## 8 factor(statpop_agecat5)8 1.1100281 1.0887962 1.1316740
## 9 factor(statpop_agecat5)9 1.1118561 1.0903167 1.1338210

```

```

## 10 factor(statpop_agecat5)10 1.1492376 1.1263416 1.1725991
## 11 factor(statpop_agecat5)11 1.1457018 1.1211694 1.1707710
## 12 factor(statpop_agecat5)12 1.1464542 1.1206745 1.1728270
## 13 factor(statpop_agecat5)13 1.0983005 1.0710174 1.1262786
## 14 factor(statpop_agecat5)14 0.9089838 0.8843204 0.9343351
## 15           female 1.0366692 1.0292986 1.0440927
## 16           factor(nat)2 0.9441447 0.9345296 0.9538587
## 17           factor(nat)3 0.8092078 0.7945234 0.8241636
## 18           factor(nat)4 0.7877280 0.7702867 0.8055643
## 19           factor(civil)2 1.0567839 1.0483939 1.0652410
## 20           factor(civil)3 0.9870434 0.9713627 1.0029773
## 21           factor(civil)4 1.0397864 1.0259945 1.0537636
## 22           factor(canton)2 1.0696812 1.0502078 1.0895156
## 23           factor(canton)3 2.0922211 2.0523785 2.1328372
## 24           factor(canton)4 1.1168060 1.0466629 1.1916499
## 25           factor(canton)5 0.9824343 0.9482447 1.0178567
## 26           factor(canton)6 0.9347093 0.8714731 1.0025340
## 27           factor(canton)7 1.0135957 0.9518682 1.0793261
## 28           factor(canton)8 0.9340134 0.8737523 0.9984305
## 29           factor(canton)9 1.8909976 1.8340432 1.9497207
## 30           factor(canton)10 0.9635439 0.9374064 0.9904102
## 31           factor(canton)11 1.0237648 0.9960445 1.0522566
## 32           factor(canton)12 1.0569627 1.0252226 1.0896855
## 33           factor(canton)13 1.0584307 1.0309285 1.0866666
## 34           factor(canton)14 1.0249405 0.9788160 1.0732386
## 35           factor(canton)15 1.0621506 1.0055655 1.1219198
## 36           factor(canton)16 1.0022202 0.9046300 1.1103382
## 37           factor(canton)17 1.0070443 0.9851831 1.0293907
## 38           factor(canton)18 0.9775217 0.9481088 1.0078472
## 39           factor(canton)19 2.0235285 1.9891880 2.0584619
## 40           factor(canton)20 2.0055278 1.9615391 2.0505030
## 41           factor(canton)21 2.1157632 2.0742793 2.1580767
## 42           factor(canton)22 1.9770581 1.9445223 2.0101383
## 43           factor(canton)23 0.8973728 0.8737921 0.9215899
## 44           factor(canton)24 2.2612001 2.2066489 2.3170998
## 45           factor(canton)25 1.9366322 1.9004721 1.9734803
## 46           factor(canton)26 2.2511091 2.1728492 2.3321876
## 47           factor(canton)27 2.3301389 2.2397066 2.4242225
## 48           factor(canton)28 4.2435951 4.1511140 4.3381365
## 49           factor(canton)29 4.3109022 4.2408682 4.3820927
## 50           died2011 0.6670748 0.6377409 0.6977579

```

```
rm(mod)
```

Figure 1

```
output <- c()

### Mortality rates of STATPOP population (type=1), by gender

type <- 1

sex <- 1
mod <- glm(died2011 ~ factor(statpop_agecat5)-1, data=data,
           offset=log(fupt_rounded), subset=female==sex, family=poisson())
output <- rbind(output, data.frame(agecat=as.numeric(mod$xlevels[[1]]),
                                   female=sex, rate=exp(mod$coeff), type=type))

sex <- 0
mod <- glm(died2011 ~ factor(statpop_agecat5)-1, data=data,
           offset=log(fupt_rounded), subset=female==sex, family=poisson())
output <- rbind(output, data.frame(agecat=as.numeric(mod$xlevels[[1]]),
                                   female=sex, rate=exp(mod$coeff), type=type))

### Restrict to SE 2010 population
data <- data[data$inse2010==1, ]

### Mortality rates of SE 2010 population using IPW weights (type=2), by gender

type <- 2

sex <- 1
mod <- glm(died2011 ~ factor(statpop_agecat5)-1, data=data, offset=log(fupt_rounded),
           subset=female==sex, family=poisson(), weights=ipw2010)
output <- rbind(output, data.frame(agecat=as.numeric(mod$xlevels[[1]]),
                                   female=sex, rate=exp(mod$coeff), type=type))

sex <- 0
mod <- glm(died2011 ~ factor(statpop_agecat5)-1, data=data, offset=log(fupt_rounded),
           subset=female==sex, family=poisson(), weights=ipw2010)
output <- rbind(output, data.frame(agecat=as.numeric(mod$xlevels[[1]]),
                                   female=sex, rate=exp(mod$coeff), type=type))

### Mortality rates of SE 2010 population using CSW weights (type=3), by gender

type <- 3

sex <- 1
mod <- glm(died2011 ~ factor(statpop_agecat5)-1, data=data, offset=log(fupt_rounded),
           subset=female==sex, family=poisson(), weights=weight2010)
output <- rbind(output, data.frame(agecat=as.numeric(mod$xlevels[[1]]),
```



```

        female=sex, rate=exp(mod$coeff), type=type))

sex <- 0
mod <- glm(died2011 ~ factor(statpop_agecat5)-1, data=data, offset=log(fupt_rounded),
           subset=female==sex, family=poisson(), weights=weight2010)
output <- rbind(output, data.frame(agecat=as.numeric(mod$xlevels[[1]]),
                                   female=sex, rate=exp(mod$coeff), type=type))

### Mortality rates of SE 2010 population using no weights (type=4), by gender

type <- 4

sex <- 1
mod <- glm(died2011 ~ factor(statpop_agecat5)-1, data=data, offset=log(fupt_rounded),
           subset=female==sex, family=poisson())
output <- rbind(output, data.frame(agecat=as.numeric(mod$xlevels[[1]]),
                                   female=sex, rate=exp(mod$coeff), type=type))

sex <- 0
mod <- glm(died2011 ~ factor(statpop_agecat5)-1, data=data, offset=log(fupt_rounded),
           subset=female==sex, family=poisson())
output <- rbind(output, data.frame(agecat=as.numeric(mod$xlevels[[1]]),
                                   female=sex, rate=exp(mod$coeff), type=type))

output$agecat <- factor(output$agecat, levels=0:14,
                        labels=c("[15,20)", "[20,25)", "[25,30)", "[30,35)",
                                "[35,40)", "[40,45)", "[45,50)", "[50,55)",
                                "[55,60)", "[60,65)", "[65,70)", "[70,75)",
                                "[75,80)", "[80,85)", ">=85" ))
output$female <- factor(output$female, levels=0:1, labels=c("Men", "Women"))
output$type <- factor(output$type, levels=1:4, labels=c("STATPOP 2010",
               "SE 2010 (IPW)", "SE 2010 (CSW)", "SE 2010 (unweighted)"))

factor <- 100000
output$rate <- output$rate*factor
dimnames(output)[[1]] <- 1:length(dimnames(output)[[1]])

```

```
### Show results
```

```
output
```

##	agecat	female	rate	type
## 1	[15,20)	Women	15.377923	STATPOP 2010
## 2	[20,25)	Women	23.696305	STATPOP 2010
## 3	[25,30)	Women	21.996134	STATPOP 2010
## 4	[30,35)	Women	35.067326	STATPOP 2010

## 5	[35,40)	Women	41.586601	STATPOP	2010
## 6	[40,45)	Women	78.266849	STATPOP	2010
## 7	[45,50)	Women	130.144776	STATPOP	2010
## 8	[50,55)	Women	198.182110	STATPOP	2010
## 9	[55,60)	Women	310.278384	STATPOP	2010
## 10	[60,65)	Women	498.732925	STATPOP	2010
## 11	[65,70)	Women	784.287264	STATPOP	2010
## 12	[70,75)	Women	1251.000126	STATPOP	2010
## 13	[75,80)	Women	2286.724441	STATPOP	2010
## 14	[80,85)	Women	4644.773608	STATPOP	2010
## 15	>=85	Women	13657.925525	STATPOP	2010
## 16	[15,20)	Men	39.199816	STATPOP	2010
## 17	[20,25)	Men	54.180393	STATPOP	2010
## 18	[25,30)	Men	44.578863	STATPOP	2010
## 19	[30,35)	Men	56.975437	STATPOP	2010
## 20	[35,40)	Men	75.497232	STATPOP	2010
## 21	[40,45)	Men	129.308223	STATPOP	2010
## 22	[45,50)	Men	202.020019	STATPOP	2010
## 23	[50,55)	Men	338.941367	STATPOP	2010
## 24	[55,60)	Men	532.887068	STATPOP	2010
## 25	[60,65)	Men	864.189059	STATPOP	2010
## 26	[65,70)	Men	1392.530477	STATPOP	2010
## 27	[70,75)	Men	2209.014961	STATPOP	2010
## 28	[75,80)	Men	3934.881354	STATPOP	2010
## 29	[80,85)	Men	7271.082613	STATPOP	2010
## 30	>=85	Men	16579.584135	STATPOP	2010
## 31	[15,20)	Women	15.679440	SE	2010 (IPW)
## 32	[20,25)	Women	27.720144	SE	2010 (IPW)
## 33	[25,30)	Women	9.714367	SE	2010 (IPW)
## 34	[30,35)	Women	27.847254	SE	2010 (IPW)
## 35	[35,40)	Women	60.422426	SE	2010 (IPW)
## 36	[40,45)	Women	81.052945	SE	2010 (IPW)
## 37	[45,50)	Women	114.689100	SE	2010 (IPW)
## 38	[50,55)	Women	179.922461	SE	2010 (IPW)
## 39	[55,60)	Women	291.687019	SE	2010 (IPW)
## 40	[60,65)	Women	438.695440	SE	2010 (IPW)
## 41	[65,70)	Women	800.636296	SE	2010 (IPW)
## 42	[70,75)	Women	1393.834663	SE	2010 (IPW)
## 43	[75,80)	Women	2161.236569	SE	2010 (IPW)
## 44	[80,85)	Women	4665.285213	SE	2010 (IPW)
## 45	>=85	Women	12813.540168	SE	2010 (IPW)
## 46	[15,20)	Men	38.200556	SE	2010 (IPW)
## 47	[20,25)	Men	66.468834	SE	2010 (IPW)
## 48	[25,30)	Men	56.942586	SE	2010 (IPW)
## 49	[30,35)	Men	39.686788	SE	2010 (IPW)
## 50	[35,40)	Men	73.910474	SE	2010 (IPW)
## 51	[40,45)	Men	103.654033	SE	2010 (IPW)
## 52	[45,50)	Men	228.540248	SE	2010 (IPW)

## 53	[50,55)	Men	324.563608	SE 2010 (IPW)
## 54	[55,60)	Men	535.784586	SE 2010 (IPW)
## 55	[60,65)	Men	852.768369	SE 2010 (IPW)
## 56	[65,70)	Men	1405.105336	SE 2010 (IPW)
## 57	[70,75)	Men	2235.427910	SE 2010 (IPW)
## 58	[75,80)	Men	3811.316794	SE 2010 (IPW)
## 59	[80,85)	Men	7671.756630	SE 2010 (IPW)
## 60	>=85	Men	16225.323782	SE 2010 (IPW)
## 61	[15,20)	Women	21.449628	SE 2010 (CSW)
## 62	[20,25)	Women	45.284756	SE 2010 (CSW)
## 63	[25,30)	Women	4.242638	SE 2010 (CSW)
## 64	[30,35)	Women	36.903647	SE 2010 (CSW)
## 65	[35,40)	Women	64.170143	SE 2010 (CSW)
## 66	[40,45)	Women	47.662749	SE 2010 (CSW)
## 67	[45,50)	Women	75.698020	SE 2010 (CSW)
## 68	[50,55)	Women	121.975222	SE 2010 (CSW)
## 69	[55,60)	Women	213.247322	SE 2010 (CSW)
## 70	[60,65)	Women	329.160980	SE 2010 (CSW)
## 71	[65,70)	Women	533.659322	SE 2010 (CSW)
## 72	[70,75)	Women	759.558373	SE 2010 (CSW)
## 73	[75,80)	Women	1573.501336	SE 2010 (CSW)
## 74	[80,85)	Women	2619.861638	SE 2010 (CSW)
## 75	>=85	Women	7250.399009	SE 2010 (CSW)
## 76	[15,20)	Men	40.598662	SE 2010 (CSW)
## 77	[20,25)	Men	78.755315	SE 2010 (CSW)
## 78	[25,30)	Men	19.119003	SE 2010 (CSW)
## 79	[30,35)	Men	14.102527	SE 2010 (CSW)
## 80	[35,40)	Men	30.456613	SE 2010 (CSW)
## 81	[40,45)	Men	76.866091	SE 2010 (CSW)
## 82	[45,50)	Men	199.396834	SE 2010 (CSW)
## 83	[50,55)	Men	246.019683	SE 2010 (CSW)
## 84	[55,60)	Men	344.718246	SE 2010 (CSW)
## 85	[60,65)	Men	696.754791	SE 2010 (CSW)
## 86	[65,70)	Men	1299.106508	SE 2010 (CSW)
## 87	[70,75)	Men	1538.654401	SE 2010 (CSW)
## 88	[75,80)	Men	3302.448952	SE 2010 (CSW)
## 89	[80,85)	Men	5181.721530	SE 2010 (CSW)
## 90	>=85	Men	13284.599004	SE 2010 (CSW)
## 91	[15,20)	Women	20.639409	SE 2010 (unweighted)
## 92	[20,25)	Women	38.143917	SE 2010 (unweighted)
## 93	[25,30)	Women	8.166107	SE 2010 (unweighted)
## 94	[30,35)	Women	44.391832	SE 2010 (unweighted)
## 95	[35,40)	Women	50.256307	SE 2010 (unweighted)
## 96	[40,45)	Women	45.867352	SE 2010 (unweighted)
## 97	[45,50)	Women	96.508950	SE 2010 (unweighted)
## 98	[50,55)	Women	125.417752	SE 2010 (unweighted)
## 99	[55,60)	Women	233.759945	SE 2010 (unweighted)
## 100	[60,65)	Women	369.350910	SE 2010 (unweighted)

```

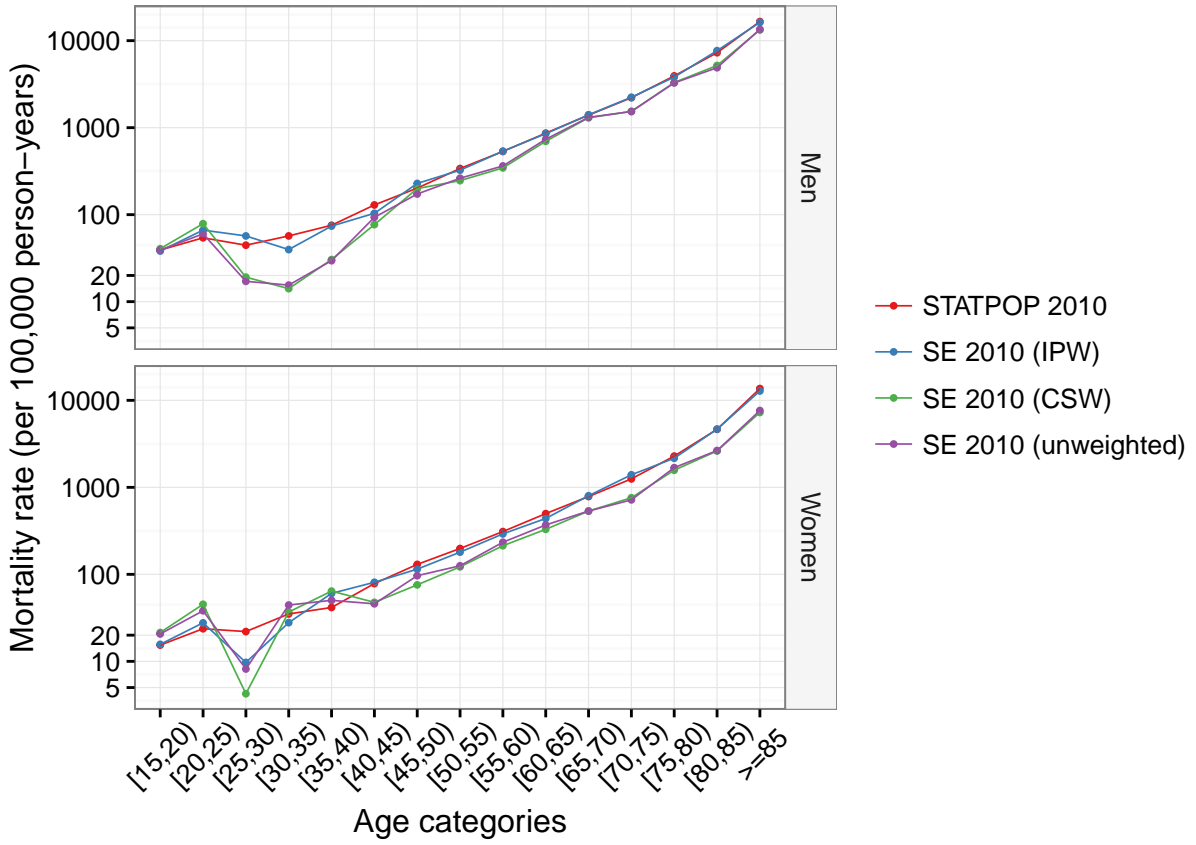
## 101 [65,70) Women 534.364495 SE 2010 (unweighted)
## 102 [70,75) Women 718.236012 SE 2010 (unweighted)
## 103 [75,80) Women 1683.780501 SE 2010 (unweighted)
## 104 [80,85) Women 2647.884340 SE 2010 (unweighted)
## 105 >=85 Women 7640.140785 SE 2010 (unweighted)
## 106 [15,20) Men 39.237998 SE 2010 (unweighted)
## 107 [20,25) Men 60.179334 SE 2010 (unweighted)
## 108 [25,30) Men 17.125634 SE 2010 (unweighted)
## 109 [30,35) Men 15.489708 SE 2010 (unweighted)
## 110 [35,40) Men 29.587331 SE 2010 (unweighted)
## 111 [40,45) Men 93.277988 SE 2010 (unweighted)
## 112 [45,50) Men 172.683956 SE 2010 (unweighted)
## 113 [50,55) Men 262.191924 SE 2010 (unweighted)
## 114 [55,60) Men 362.957586 SE 2010 (unweighted)
## 115 [60,65) Men 735.294118 SE 2010 (unweighted)
## 116 [65,70) Men 1312.997626 SE 2010 (unweighted)
## 117 [70,75) Men 1532.783714 SE 2010 (unweighted)
## 118 [75,80) Men 3270.615905 SE 2010 (unweighted)
## 119 [80,85) Men 4885.625530 SE 2010 (unweighted)
## 120 >=85 Men 13489.527867 SE 2010 (unweighted)

```

```

p <- ggplot(data=output, aes(x=agecat, y=rate, color=type))
p1 <- p + geom_point(size=0.7) + geom_line(aes(group=type), size=0.3)
p1 <- p1 + scale_y_log10(breaks=c(5,10,20,100,1000,10000)) + facet_grid(female~.)
p1 <- p1 + ylab("Mortality rate (per 100,000 person-years)") + xlab("Age categories")
p1 <- p1 + theme_bw()
p1 <- p1 + theme(strip.background = element_rect(fill="#F5F5F5"),
                 legend.key = element_blank(),
                 axis.text.x = element_text(angle=45, vjust=0.6))
p1 <- p1 + scale_color_brewer("", palette = "Set1")
p1

```



Life expectancy (Stata code)

Own written Stata command based on:

Moser A., Clough-Gorr K., Zwahlen M. 2015. Modeling absolute differences in life expectancy with a censored skew-normal regression approach. PeerJ 3:e1162. DOI: 10.7717/peerj.1162.

Stata code and command is available at <https://github.com/MoserGitHub/censn>.

```

-----
name: <unnamed>
log: F:\SNC\SurveyMortality\LE_Nov2017.txt
log type: text
opened on: 4 Nov 2017, 14: 50: 14

. import delimited using "F:\SNC\SurveyMortality\data\SwissCensus2010.csv", clear
(12 vars, 6,729,363 obs)

. adopath + "F:\SNC\SurveyMortality\ado\"
[1] (BASE) "C:\Program Files (x86)\Stata14\ado\base/"
[2] (SITE) "C:\Program Files (x86)\Stata14\ado\site/"
[3] " ."
[4] (PERSONAL) "c:\ado\personal/"
[5] (PLUS) "c:\ado\plus/"
[6] (OLDPLACE) "c:\ado/"
[7] "F:\SNC\SurveyMortality\ado\"

. replace agestart_rounded=agestart_rounded-30
(6,729,363 real changes made)

. replace ageend_rounded=ageend_rounded-30
(6,729,363 real changes made)

. * Restrict to individuals aged 30 years or older
. drop if agestart_rounded<0
(1,481,413 observations deleted)

. * STATPOP 2010
. censn ageend_rounded if female==0, lefttrun(agestart_rounded)
failure(died2011)

```

Fitting constant-only model:

```

Iteration 0: log likelihood = -204064.16 (not concave)
Iteration 1: log likelihood = -127946.07 (not concave)
Iteration 2: log likelihood = -120713.33
Iteration 3: log likelihood = -119590.76
Iteration 4: log likelihood = -119490.28
Iteration 5: log likelihood = -119483.2
Iteration 6: log likelihood = -119483.17
Iteration 7: log likelihood = -119483.17

```

```

Number of obs = 2,546,113
Wald chi2(0) = .
Prob > chi2 = .

```

Log likelihood = -119483.17

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	50.91508	.0568905	894.97	0.000	50.80358 51.02659
sigma					
_cons	11.81383	.0443428	266.42	0.000	11.72692 11.90074
shape					
_cons	-4.936167	.1105058	-44.67	0.000	-5.152754 -4.719579

Skewness parameter in CP

Coef.: -.84768077

Std. Err.: .00029225

[95% CI]: -.84825357 , -.84710797

. censn ageend_rounded if female==1, lefttrun(agestart_rounded)
failure(died2011)

Fitting constant-only model:

Iteration 0: log likelihood = -204947.75 (not concave)
Iteration 1: log likelihood = -197645.83 (not concave)
Iteration 2: log likelihood = -124982.25
Iteration 3: log likelihood = -121747.83
Iteration 4: log likelihood = -120931.22
Iteration 5: log likelihood = -120908.33
Iteration 6: log likelihood = -120908.13
Iteration 7: log likelihood = -120908.13

Number of obs = 2,701,837
Wald chi2(0) = .
Prob > chi2 = .

Log likelihood = -120908.13

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	54.78439	.0525072	1043.37	0.000	54.68148 54.8873
sigma					
_cons	10.93951	.0401325	272.58	0.000	10.86085 11.01817
shape					
_cons	-5.189691	.1015395	-51.11	0.000	-5.388704 -4.990677

Skewness parameter in CP

Coef.: -.86013039

Std. Err.: .00022696

[95% CI]: -.86057523 , -.85968556

. * SE 2010

. * Unweighted

. censn ageend_rounded if female==0 & inse2010==1, lefttrun(agestart_rounded)
failure(
> died2011) difficult

Fitting constant-only model:

Iteration 0: log likelihood = -9159.9159 (not concave)
Iteration 1: log likelihood = -5627.0971 (not concave)
Iteration 2: log likelihood = -5606.496 (not concave)
Iteration 3: log likelihood = -5321.1561 (not concave)
Iteration 4: log likelihood = -5248.5118 (not concave)
Iteration 5: log likelihood = -4964.3642 (not concave)
Iteration 6: log likelihood = -4760.7278 (not concave)
Iteration 7: log likelihood = -4711.1978
Iteration 8: log likelihood = -4700.0464
Iteration 9: log likelihood = -4699.5688
Iteration 10: log likelihood = -4699.5669
Iteration 11: log likelihood = -4699.5669

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Log likelihood = -4699.5669 Number of obs = 121,021
 Wald chi2(0) = .
 Prob > chi2 = .

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	53.32098	.2829822	188.43	0.000	52.76634 53.87561
sigma					
_cons	11.60652	.2229127	52.07	0.000	11.16962 12.04343
shape					
_cons	-5.181367	.7839979	-6.61	0.000	-6.717974 -3.644759

Skewness parameter in CP

Coef.: -.85974592

Std. Err.: .07764274

[95% CI]: -1.0119229 , -.70756894

. censn ageend_rounded if female==1 & inse2010==1, lefttrun(agestart_rounded)
 failure(
 > died2011) difficult

Fitting constant-only model:

Iteration 0: log likelihood = -8933.3732 (not concave)
 Iteration 1: log likelihood = -4910.6544 (not concave)
 Iteration 2: log likelihood = -4737.8554 (not concave)
 Iteration 3: log likelihood = -4515.3801 (not concave)
 Iteration 4: log likelihood = -4455.4714 (not concave)
 Iteration 5: log likelihood = -4288.4417 (not concave)
 Iteration 6: log likelihood = -4246.4338 (not concave)
 Iteration 7: log likelihood = -4101.2046
 Iteration 8: log likelihood = -4054.2681
 Iteration 9: log likelihood = -4044.9889
 Iteration 10: log likelihood = -4044.8558
 Iteration 11: log likelihood = -4023.6899
 Iteration 12: log likelihood = -3983.7399
 Iteration 13: log likelihood = -3975.368
 Iteration 14: log likelihood = -3973.2685
 Iteration 15: log likelihood = -3972.3757
 Iteration 16: log likelihood = -3972.2047
 Iteration 17: log likelihood = -3972.1824
 Iteration 18: log likelihood = -3972.1818
 Iteration 19: log likelihood = -3972.1818

Log likelihood = -3972.1818 Number of obs = 131,990
 Wald chi2(0) = .
 Prob > chi2 = .

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	58.86656	.320969	183.40	0.000	58.23747 59.49565
sigma					
_cons	11.50494	.2489221	46.22	0.000	11.01706 11.99282
shape					
_cons	-10.89448	6.909487	-1.58	0.115	-24.43682 2.647869

Skewness parameter in CP

Coef.: -.96163395

Std. Err.: .91502505

[95% CI]: -2.7550501 , .83178219

```
. * CS weighted
. censn ageend_rouned [pw=weight2010] if female==0 & inse2010==1,
lefttrun(agestart_r
> ounded) failure(died2011) difficult
```

Fitting constant-only model:

```
Iteration 0: log pseudolikelihood = -185108.1 (not concave)
Iteration 1: log pseudolikelihood = -117415.53 (not concave)
Iteration 2: log pseudolikelihood = -113323.03 (not concave)
Iteration 3: log pseudolikelihood = -109658.08 (not concave)
Iteration 4: log pseudolikelihood = -105581.28 (not concave)
Iteration 5: log pseudolikelihood = -103178.11 (not concave)
Iteration 6: log pseudolikelihood = -100420.17 (not concave)
Iteration 7: log pseudolikelihood = -96223.153
Iteration 8: log pseudolikelihood = -96040.641 (backed up)
Iteration 9: log pseudolikelihood = -94241.685
Iteration 10: log pseudolikelihood = -94222.974
Iteration 11: log pseudolikelihood = -94210.911
Iteration 12: log pseudolikelihood = -94210.89
Iteration 13: log pseudolikelihood = -94210.89
```

```
Log pseudolikelihood = -94210.89
Number of obs = 121,021
Wald chi2(0) = .
Prob > chi2 = .
```

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
mu						
_cons	53.35384	.3199026	166.78	0.000	52.72684	53.98084
sigma						
_cons	11.58996	.2629378	44.08	0.000	11.07461	12.10531
shape						
_cons	-4.435974	.4977457	-8.91	0.000	-5.411538	-3.46041

Skewness parameter in CP

Coef.: -.81784394

Std. Err.: .02362381

[95% CI]: -.86414576 , -.77154213

```
. censn ageend_rouned [pw=weight2010] if female==1 & inse2010==1,
lefttrun(agestart_r
> ounded) failure(died2011) difficult
```

Fitting constant-only model:

```
Iteration 0: log pseudolikelihood = -173967.84 (not concave)
Iteration 1: log pseudolikelihood = -92477.105 (not concave)
Iteration 2: log pseudolikelihood = -90265.987 (not concave)
```

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```
Iteration 3: log pseudolikelihood = -84392.726 (not concave)
Iteration 4: log pseudolikelihood = -84378.809 (not concave)
Iteration 5: log pseudolikelihood = -81224.765 (not concave)
Iteration 6: log pseudolikelihood = -79049.605 (not concave)
Iteration 7: log pseudolikelihood = -76734.701
Iteration 8: log pseudolikelihood = -76305.367
Iteration 9: log pseudolikelihood = -76038.98
Iteration 10: log pseudolikelihood = -76037.754
Iteration 11: log pseudolikelihood = -76037.691
Iteration 12: log pseudolikelihood = -76037.688
Iteration 13: log pseudolikelihood = -75585.041
Iteration 14: log pseudolikelihood = -74925.628
Iteration 15: log pseudolikelihood = -74804.441
Iteration 16: log pseudolikelihood = -74765.511
Iteration 17: log pseudolikelihood = -74757.048
Iteration 18: log pseudolikelihood = -74754.789
Iteration 19: log pseudolikelihood = -74754.544
Iteration 20: log pseudolikelihood = -74754.537
Iteration 21: log pseudolikelihood = -74754.537
```

```
Number of obs = 131,990
Wald chi2(0) = .
Prob > chi2 = .
```

Log pseudolikelihood = -74754.537

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	59.19851	.3831018	154.52	0.000	58.44764 59.94937
sigma					
_cons	11.49247	.3472371	33.10	0.000	10.8119 12.17305
shape					
_cons	-8.634522	3.368181	-2.56	0.010	-15.23604 -2.033009

Skewness parameter in CP

Coef.: -.94260365

Std. Err.: .71855036

[95% CI]: -2.3509365 , .46572918

* IP weighted

```
. cnsn ageend_rouned [pw=i pw2010] if female==0 & inse2010==1,
lefttrun(agestart_roun
> ded) failure(died2011) diffcult
```

Fitting constant-only model:

```
Iteration 0: log pseudolikelihood = -202293.98 (not concave)
Iteration 1: log pseudolikelihood = -135357.74 (not concave)
Iteration 2: log pseudolikelihood = -133832.29 (not concave)
Iteration 3: log pseudolikelihood = -126260.43 (not concave)
Iteration 4: log pseudolikelihood = -122469
Iteration 5: log pseudolikelihood = -119868.08
Iteration 6: log pseudolikelihood = -118332.94
Iteration 7: log pseudolikelihood = -118293.09
Iteration 8: log pseudolikelihood = -118292.42
Iteration 9: log pseudolikelihood = -118292.42
```

```
Number of obs = 121,021
Wald chi2(0) = .
Prob > chi2 = .
```

Log pseudolikelihood = -118292.42

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		Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
mu	_cons	50.92614	.3478712	146.39	0.000	50.24433	51.60796
sigma	_cons	11.71864	.3027535	38.71	0.000	11.12525	12.31202
shape	_cons	-4.373761	.445647	-9.81	0.000	-5.247213	-3.500309

Skewness parameter in CP

Coef.: -.81355052

Std. Err.: .01738051

[95% CI]: -.84761569 , -.77948535

```
. censn ageend_rounded [pw=i pw2010] if female==1 & inuse2010==1,
lefttrun(agestart_roun
> ded) failure(died2011) difficult
```

Fitting constant-only model:

```
Iteration 0: log pseudolikelihood = -204835.01 (not concave)
Iteration 1: log pseudolikelihood = -142195.92 (not concave)
Iteration 2: log pseudolikelihood = -130641.31 (not concave)
Iteration 3: log pseudolikelihood = -120798.93 (not concave)
Iteration 4: log pseudolikelihood = -117473.54 (not concave)
Iteration 5: log pseudolikelihood = -116953.74
Iteration 6: log pseudolikelihood = -116720.59 (not concave)
Iteration 7: log pseudolikelihood = -116161.54
Iteration 8: log pseudolikelihood = -116139.6
Iteration 9: log pseudolikelihood = -116131.58
Iteration 10: log pseudolikelihood = -116131.56
Iteration 11: log pseudolikelihood = -116131.56
```

```
Log pseudolikelihood = -116131.56
Number of obs = 131,990
Wald chi2(0) = .
Prob > chi2 = .
```

		Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
mu	_cons	54.80675	.3479427	157.52	0.000	54.1248	55.48871
sigma	_cons	10.70856	.2874598	37.25	0.000	10.14515	11.27197
shape	_cons	-6.421754	.7594985	-8.46	0.000	-7.910343	-4.933164

Skewness parameter in CP

Coef.: -.90335602

Std. Err.: .07179521

[95% CI]: -1.044072 , -.76264

. replace agestart_rounded=agestart_rounded-35
(5,247,950 real changes made)

. replace ageend_rounded=ageend_rounded-35
(5,247,950 real changes made)

. * Restrict to individuals aged 65 years or older
. drop if agestart_rounded<0
(3,910,542 observations deleted)

. * STATPOP 2010

. censn ageend_rounded if female==0, lefttrun(agestart_rounded)
failure(died2011)

Fitting constant-only model:

Iteration 0: log likelihood = -102837.2 (not concave)
Iteration 1: log likelihood = -98421.569 (not concave)
Iteration 2: log likelihood = -94910.871 (not concave)
Iteration 3: log likelihood = -93907.237 (not concave)
Iteration 4: log likelihood = -91186.706 (not concave)
Iteration 5: log likelihood = -84202.104 (not concave)
Iteration 6: log likelihood = -83998.888
Iteration 7: log likelihood = -83731.752
Iteration 8: log likelihood = -83718.549
Iteration 9: log likelihood = -83718.462
Iteration 10: log likelihood = -83718.462

Number of obs = 575,143
Wald chi2(0) = .
Prob > chi2 = .

Log likelihood = -83718.462

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	17.62199	.0851398	206.98	0.000	17.45512 17.78886
sigma					
_cons	9.904003	.0824305	120.15	0.000	9.742442 10.06556
shape					
_cons	-3.33546	.0985455	-33.85	0.000	-3.528606 -3.142315

Skewness parameter in CP

Coef.: -.71442172

Std. Err.: .00020754

[95% CI]: -.71482849 , -.71401495

. censn ageend_rounded if female==1, lefttrun(agestart_rounded)
failure(died2011)

Fitting constant-only model:

Iteration 0: log likelihood = -122946.3 (not concave)
Iteration 1: log likelihood = -100337.4 (not concave)
Iteration 2: log likelihood = -98131.809 (not concave)
Iteration 3: log likelihood = -97824.239
Iteration 4: log likelihood = -97492.706
Iteration 5: log likelihood = -97486.484

Iteration 6: log likelihood = -97486.482

Number of obs = 762,265
 Wald chi2(0) = .
 Prob > chi2 = .

Log likelihood = -97486.482

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	21.6647	.0562523	385.13	0.000	21.55445 21.77496
sigma					
_cons	8.82385	.0511099	172.64	0.000	8.723677 8.924024
shape					
_cons	-3.321786	.0746158	-44.52	0.000	-3.46803 -3.175542

Skewness parameter in CP

Coef.: -.71267205

Std. Err.: .00009029

[95% CI]: -.71284902 , -.71249508

. * SE 2010

. * Unweighted

. censn ageend_rounded if female==0 & inse2010==1, lefttrun(agestart_rounded)
 failure(> died2011) difficult

Fitting constant-only model:

Iteration 0: log likelihood = -4710.5291 (not concave)
 Iteration 1: log likelihood = -3605.2863 (not concave)
 Iteration 2: log likelihood = -3568.683 (not concave)
 Iteration 3: log likelihood = -3511.469 (not concave)
 Iteration 4: log likelihood = -3409.5631
 Iteration 5: log likelihood = -3385.0388
 Iteration 6: log likelihood = -3384.4301
 Iteration 7: log likelihood = -3384.3049
 Iteration 8: log likelihood = -3384.3048

Number of obs = 27,987
 Wald chi2(0) = .
 Prob > chi2 = .

Log likelihood = -3384.3048

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	19.31354	.394144	49.00	0.000	18.54103 20.08605
sigma					
_cons	10.35302	.3984034	25.99	0.000	9.57216 11.13387
shape					
_cons	-3.995615	.6438635	-6.21	0.000	-5.257564 -2.733665

Skewness parameter in CP

Coef.: -.78404767

Std. Err.: .04714702

[95% CI]: -.87645412 , -.69164121

```
. censn ageend_rouned if female==1 & inse2010==1, lefttrun(agestart_rouned)
failure(
> died2011) diffcult
```

Fitting constant-only model:

```
Iteration 0: log likelihood = -5247.3335 (not concave)
Iteration 1: log likelihood = -3242.8946 (not concave)
Iteration 2: log likelihood = -3165.9234 (not concave)
Iteration 3: log likelihood = -3108.3634
Iteration 4: log likelihood = -3089.9284
Iteration 5: log likelihood = -3084.47
Iteration 6: log likelihood = -3084.4142
Iteration 7: log likelihood = -3084.4118
Iteration 8: log likelihood = -3084.4117
Iteration 9: log likelihood = -3074.3715
Iteration 10: log likelihood = -3068.4501
Iteration 11: log likelihood = -3059.6613
Iteration 12: log likelihood = -3059.5725
Iteration 13: log likelihood = -3059.572
Iteration 14: log likelihood = -3059.572
```

```
Number of obs = 36,669
Wald chi2(0) = .
Prob > chi2 = .
```

Log likelihood = -3059.572

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	25.17702	.2873424	87.62	0.000	24.61384 25.7402
sigma					
_cons	9.040391	.2794137	32.35	0.000	8.49275 9.588031
shape					
_cons	-4.688725	1.234264	-3.80	0.000	-7.107838 -2.269612

Skewness parameter in CP

Coef.: -.83389166

Std. Err.: .21169414

[95% CI]: -1.2488045 , -.41897878

```
. * CS weighted
. censn ageend_rouned [pw=weight2010] if female==0 & inse2010==1,
lefttrun(agestart_rouned) failure(died2011) diffcult
```

Fitting constant-only model:

```
Iteration 0: log pseudolikelihood = -93569.748 (not concave)
Iteration 1: log pseudolikelihood = -69745.151 (not concave)
Iteration 2: log pseudolikelihood = -68316.215 (not concave)
Iteration 3: log pseudolikelihood = -67773.325
Iteration 4: log pseudolikelihood = -67684.126
Iteration 5: log pseudolikelihood = -67675.956
Iteration 6: log pseudolikelihood = -67675.945
Iteration 7: log pseudolikelihood = -67675.945
```

LE_Nov2017. txt

Log pseudolikelihood = -67675.945
 Number of obs = 27,987
 Wald chi2(0) = .
 Prob > chi2 = .

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
mu						
_cons	19.37432	.4233577	45.76	0.000	18.54456	20.20409
sigma						
_cons	10.27155	.4361618	23.55	0.000	9.41669	11.12641
shape						
_cons	-3.334177	.4582202	-7.28	0.000	-4.232273	-2.436082

Skewness parameter in CP

Coef.: -.71425818

Std. Err.: .01878486

[95% CI]: -.75107582 , -.67744054

. censn ageend_rounded [pw=weight2010] if female==1 & inuse2010==1,
 lefttrun(agestart_rou
 > ounded) failure(died2011) difficult

Fitting constant-only model:

Iteration 0: log pseudolikelihood = -100139.91 (not concave)
 Iteration 1: log pseudolikelihood = -60542.492 (not concave)
 Iteration 2: log pseudolikelihood = -59160.972 (not concave)
 Iteration 3: log pseudolikelihood = -58171.492
 Iteration 4: log pseudolikelihood = -58069.301 (not concave)
 Iteration 5: log pseudolikelihood = -57827.926
 Iteration 6: log pseudolikelihood = -57580.745
 Iteration 7: log pseudolikelihood = -57566.531
 Iteration 8: log pseudolikelihood = -57566.386
 Iteration 9: log pseudolikelihood = -57566.385

Log pseudolikelihood = -57566.385
 Number of obs = 36,669
 Wald chi2(0) = .
 Prob > chi2 = .

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
mu						
_cons	25.30224	.3263267	77.54	0.000	24.66265	25.94183
sigma						
_cons	9.074467	.3335205	27.21	0.000	8.420779	9.728155
shape						
_cons	-4.605421	.8123849	-5.67	0.000	-6.197667	-3.013176

Skewness parameter in CP

Coef.: -.82883757

Std. Err.: .08467034

[95% CI]: -.99478839 , -.66288674


```
. * IP weighted
. cnsn ageend_rounded [pw=i pw2010] if femal e==0 & i nse2010==1,
leftttrun(agestart_roun
> ded) fail ure(died2011) di ffi cul t
```

Fitting constant-only model :

```
Iteration 0: log pseudol i kel i hood = -103939.86 (not concave)
Iteration 1: log pseudol i kel i hood = -83973.915 (not concave)
Iteration 2: log pseudol i kel i hood = -83633.612
Iteration 3: log pseudol i kel i hood = -83208.641
Iteration 4: log pseudol i kel i hood = -83127.167
Iteration 5: log pseudol i kel i hood = -83125.703
Iteration 6: log pseudol i kel i hood = -83125.703
```

```
Log pseudol i kel i hood = -83125.703
Number of obs = 27,987
Wald chi 2(0) = .
Prob > chi 2 = .
```

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]
mu					
_cons	17.5884	.4590447	38.32	0.000	16.68869 18.48811
sigma					
_cons	9.857571	.4486002	21.97	0.000	8.978331 10.73681
shape					
_cons	-2.960692	.4322386	-6.85	0.000	-3.807864 -2.11352

Skewness parameter in CP

Coef.: -.66080965

Std. Err.: .0159538

[95% CI]: -.69207853 , -.62954078

```
. cnsn ageend_rounded [pw=i pw2010] if femal e==1 & i nse2010==1,
leftttrun(agestart_roun
> ded) fail ure(died2011) di ffi cul t
```

Fitting constant-only model :

```
Iteration 0: log pseudol i kel i hood = -126311.64 (not concave)
Iteration 1: log pseudol i kel i hood = -99049.304 (not concave)
Iteration 2: log pseudol i kel i hood = -95606.852
Iteration 3: log pseudol i kel i hood = -95441.279
Iteration 4: log pseudol i kel i hood = -95379.23
Iteration 5: log pseudol i kel i hood = -95379.186
Iteration 6: log pseudol i kel i hood = -94113.993
Iteration 7: log pseudol i kel i hood = -93961.868
Iteration 8: log pseudol i kel i hood = -93956.38
Iteration 9: log pseudol i kel i hood = -93956.375
```

```
Log pseudol i kel i hood = -93956.375
Number of obs = 36,669
Wald chi 2(0) = .
Prob > chi 2 = .
```

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]
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LE_Nov2017. txt

mu	_cons	21.56338	.3762807	57.31	0.000	20.82589	22.30088
sigma	_cons	8.766594	.3525496	24.87	0.000	8.075609	9.457578
shape	_cons	-4.341851	.610001	-7.12	0.000	-5.537431	-3.146271

Skewness parameter in CP

Coef.: -.81129162

Std. Err.: .04091274

[95% CI]: -.89147911 , -.73110412

. log close
name: <unnamed>
log: F:\SNC\SurveyMortality\LE_Nov2017. txt
log type: text
closed on: 4 Nov 2017, 18:27:00

Construction of inverse probability weights (Stata code)

Note that, because of privacy preserving issues, we were not allowed to provide the exact date of death (variable `dod` below). Thus, the Nelson-Aalen estimator for the anonymised dataset is slightly different, than from the one used in the analysis with exact date information.

Original code:

```
gen statpop_enddate2011=d(31.12.2011)
replace statpop_enddate2011=dod if dod<=d(31.12.2011)
stset statpop_enddate2011, origin(statpop_startdate)
failure(died2011) scale(365.25)
```

Reproducible code:

```
stset endage_rounded, origin(startageage_rounded)
failure(died2011)
```

Weights.txt

```
use "F:/SNC/SurveyMortality/data/SwissCensuses.csv", clear

### Note that, because of privacy preserving issues, we were not allowed to
### provide the exact date of death (variable dod below). Thus, the
### Nelson-Aalen estimator is not exactly the same as in
### the original analysis.

### Original code:

### gen statpop_enddate2011=d(31.12.2011)
### replace statpop_enddate2011=dod if dod<=d(31.12.2011)

### stset statpop_enddate2011, origin(statpop_startdate) failure(di ed2011)
###      scale(365.25)

### Reproducible code:

stset endage_rounded, origin(startageage_rounded), failure(di ed2011)

### Nelson-Aalen estimator
sts gen na2011=na

drop _st _d _t _t0 statpop_enddate2011

*** Use only STATPOP/SE 2010
keep if statyear==2010

logistic inse2010 female##statpop_agecat5##i.di ed2011 na2011 ///
      i.r10_canton_se i.r10_nat i.r10_civil
predict prob

gen ipw2010=1/prob

drop prob1 na2011
```