

# Solution assignment 1

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First we need some data. We can go to <http://www.mortality.org> to download the data manually or use the HMDHFDplus package to download the data from within R. Here I choose to use the package. Before we can load the package via `library(HMDHFDplus)` we need to install it by running `install.packages('HMDHFDplus')`.

We use the function `readHMDweb()` to download time series of death counts and exposures for Sweden. You need to specify your own HMD username and password to the function.

```
library(HMDHFDplus)
# http://www.mortality.org/hmd/SWE/STATS/Deaths_1x1.txt
deaths <- readHMDweb(CNTRY = 'SWE', item = 'Deaths_1x1',
                    username = usr, password = pwd)
str(deaths)

## 'data.frame':   29304 obs. of  6 variables:
##  $ Year      : int  1751 1751 1751 1751 1751 1751 1751 1751 1751 1751 ...
##  $ Age       : int   0  1  2  3  4  5  6  7  8  9 ...
##  $ Female    : num  5988 1286 835 622 471 ...
##  $ Male      : num  6902 1360 882 655 498 ...
##  $ Total     : num  12890 2646 1717 1277 969 ...
##  $ OpenInterval: logi  FALSE FALSE FALSE FALSE FALSE FALSE ...

# http://www.mortality.org/hmd/SWE/STATS/Exposures_1x1.txt
exposures <- readHMDweb(CNTRY = 'SWE', item = 'Exposures_1x1',
                      username = usr, password = pwd)
str(exposures)

## 'data.frame':   29304 obs. of  6 variables:
##  $ Year      : int  1751 1751 1751 1751 1751 1751 1751 1751 1751 1751 ...
##  $ Age       : int   0  1  2  3  4  5  6  7  8  9 ...
##  $ Female    : num  28214 26035 25880 23918 19876 ...
##  $ Male      : num  28627 25683 25504 23501 19373 ...
##  $ Total     : num  56841 51718 51385 47419 39250 ...
##  $ OpenInterval: logi  FALSE FALSE FALSE FALSE FALSE FALSE ...
```

Both data series have exactly the same format: 6 variables, 29,304 rows, ages 0 to 110+ for years 1751 to 2014. This consistency makes it really easy to transform the data frame to a matrix. But why would we want to do so? Matrices with ages as rows and periods as columns are a convenient structure to store time series of life-table data. Calculating statistics for each year simply requires to loop over the columns of such a matrix.

```
years <- unique(deaths$Year) # labels for the years
ages <- 0:110 # labels for the age groups
w <- length(ages) # number of age groups

# create age-period matrices of death counts and exposures
D <- matrix(deaths$Total, nrow = w, dimnames = list(ages, years))
E <- matrix(exposures$Total, nrow = w, dimnames = list(ages, years))
```

The highest age group in our data is 110+. This makes sense for today's life-tables but historically the maximum age at death in any given year was often considerably lower. In consequence we have exposures and death counts of 0 in the highest ages for some years. Here's an example for the year 1800.

```
D[90:111, '1800']
```

```
##      89      90      91      92      93      94      95      96      97      98
## 104.58  91.92  68.14  49.39  34.99  24.23  16.39  10.83   6.97   4.40
##      99      100     101     102     103     104     105     106     107     108
##   2.70   1.39   0.65   0.00   0.00   0.00   0.00   0.00   0.00   0.00
##      109     110
##   0.00   0.00
```

```
E[90:111, '1800']
```

```
##      89      90      91      92      93      94      95      96      97      98
## 301.05 213.68 146.97  98.54  64.46  41.12  25.52  14.79   8.20   4.47
##      99      100     101     102     103     104     105     106     107     108
##   2.03   0.63   0.15   0.00   0.00   0.00   0.00   0.00   0.00   0.00
##      109     110
##   0.00   0.00
```

Zero exposures will cause all kinds of headaches when constructing a life-table. One technique to deal with this issue is to aggregate deaths and exposures over the highest ages, i.e. instead of an age group 110+ we can have 100+ as the last age group.

```
# sum up deaths and exposures past age 100
D100p <- colSums(D[101:w,])
E100p <- colSums(E[101:w,])

# remove single ages 100 and beyond from exposure and death matrices
D <- D[-(101:w),]
E <- E[-(101:w),]

# add open age group 100+ to exposure and death matrices
D <- rbind(D, D100p)
E <- rbind(E, E100p)
```

What we have now is age-period matrices of death counts and exposures over ages 0 to 100+. We calculate the age and period specific mortality rates (mortality rate matrix M) by dividing deaths by exposures.

```
M <- D/E
```

Now it's time to write a function calculating life-expectancy and the coefficient of variation for the life-table death distribution. Concerning the life-table construction we follow Preston (2001), p. 49, Box 3.1. As the coefficient of variation for the life-table death distribution depends on life-table statistics, we calculate it inside of our life-table function.

```
LTFun <- function (x, nmx) {
  # clear names of input vectors
  names(nmx) <- NULL; names(x) <- NULL

  # number of age groups
  w <- length(x)
  # width of age groups
  nx <- c(diff(x), NA)
  # time spent in age group if dying in age group
  nax <- 0.5*nx
  nax[w] <- 1/nmx[w]
  # probability of dying in age group given
  # survival until start of age group
```

```

nqx = nx*nmx / (1 + (nx-nax)*nmx)
nqx[length(nqx)] <- 1
# probability of surviving age group given
# survival until start of age group
npx <- 1 - nqx
# probability of surviving until start
# of age group
lx <- cumprod(c(1, npx[-w]))
# unconditional probability of dying in
# age group
ndx <- c(-diff(lx), lx[w])
# total exposure time spend in age group
# by life-table population
nLx <- c(lx[-1], 0)*nx + ndx*nax
nLx[w] <- nax[w]*lx[w]
# total exposure time yet to live by life-table
# population
Tx <- rev(cumsum(rev(nLx)))
# remaining life-expectancy past age x
ex <- Tx/lx

# total life-expectancy
e0 <- ex[1]
# coefficient of variation of life-table
# distribution of deaths
cv <- sqrt(sum(ndx*(x+nax-e0)^2)) / e0

# function output
list(
  lt = data.frame(x, nx, nmx, nax, nqx, npx, lx, ndx, nLx, Tx, ex),
  lt_summary = c(e0 = e0, cv = cv)
)
}

```

We can test this function on the first year in our data.

```
LTFun(0:100, M[,1])
```

```
## $lt
##      x nx      nmx      nax      nqx      npx      lx
## 1    0  1 0.226774404 0.50000 0.203679730 0.7963203 1.0000000000
## 2    1  1 0.051168600 0.50000 0.049892145 0.9501079 0.7963202705
## 3    2  1 0.033408706 0.50000 0.032859804 0.9671402 0.7565901443
## 4    3  1 0.026935855 0.50000 0.026577906 0.9734221 0.7317287401
## 5    4  1 0.024681101 0.50000 0.024380235 0.9756198 0.7122809225
## 6    5  1 0.019766134 0.50000 0.019572695 0.9804273 0.6949153460
## 7    6  1 0.013787682 0.50000 0.013693282 0.9863067 0.6813139796
## 8    7  1 0.009457109 0.50000 0.009412601 0.9905874 0.6719845550
## 9    8  1 0.006660414 0.50000 0.006638307 0.9933617 0.6656594322
## 10   9  1 0.005420721 0.50000 0.005406068 0.9945939 0.6612405803
## 11  10  1 0.005609643 0.50000 0.005593953 0.9944060 0.6576658685
## 12  11  1 0.006121171 0.50000 0.006102494 0.9938975 0.6539869166
## 13  12  1 0.006511421 0.50000 0.006490290 0.9935097 0.6499959654
## 14  13  1 0.006738204 0.50000 0.006715578 0.9932844 0.6457773030
## 15  14  1 0.006752224 0.50000 0.006729504 0.9932705 0.6414405351

```

## 16	15	1	0.006428534	0.50000	0.006407937	0.9935921	0.6371239581
## 17	16	1	0.006047135	0.50000	0.006028906	0.9939711	0.6330413078
## 18	17	1	0.005906809	0.50000	0.005889415	0.9941106	0.6292247612
## 19	18	1	0.006010376	0.50000	0.005992367	0.9940076	0.6255189951
## 20	19	1	0.006298215	0.50000	0.006278444	0.9937216	0.6217706555
## 21	20	1	0.006771070	0.50000	0.006748224	0.9932518	0.6178669034
## 22	21	1	0.007296608	0.50000	0.007270085	0.9927299	0.6136973992
## 23	22	1	0.007736466	0.50000	0.007706655	0.9922933	0.6092357672
## 24	23	1	0.008029954	0.50000	0.007997843	0.9920022	0.6045405974
## 25	24	1	0.008271070	0.50000	0.008237005	0.9917630	0.5997055767
## 26	25	1	0.008446224	0.50000	0.008410705	0.9915893	0.5947657988
## 27	26	1	0.008592461	0.50000	0.008555704	0.9914443	0.5897633992
## 28	27	1	0.008860153	0.50000	0.008821075	0.9911789	0.5847175581
## 29	28	1	0.009238528	0.50000	0.009196049	0.9908040	0.5795597206
## 30	29	1	0.009728055	0.50000	0.009680966	0.9903190	0.5742300610
## 31	30	1	0.010559577	0.50000	0.010504118	0.9894959	0.5686709592
## 32	31	1	0.011550715	0.50000	0.011484388	0.9885156	0.5626975725
## 33	32	1	0.012362153	0.50000	0.012286211	0.9877138	0.5562353351
## 34	33	1	0.012876962	0.50000	0.012794584	0.9872054	0.5494013105
## 35	34	1	0.012971218	0.50000	0.012887634	0.9871124	0.5423719491
## 36	35	1	0.012226992	0.50000	0.012152696	0.9878473	0.5353820579
## 37	36	1	0.011230847	0.50000	0.011168133	0.9888319	0.5288757224
## 38	37	1	0.010680310	0.50000	0.010623579	0.9893764	0.5229691678
## 39	38	1	0.010606182	0.50000	0.010550233	0.9894498	0.5174133638
## 40	39	1	0.011050591	0.50000	0.010989869	0.9890101	0.5119545324
## 41	40	1	0.012473023	0.50000	0.012395717	0.9876043	0.5063282192
## 42	41	1	0.014404399	0.50000	0.014301397	0.9856986	0.5000519181
## 43	42	1	0.016012235	0.50000	0.015885058	0.9841149	0.4929004771
## 44	43	1	0.017259635	0.50000	0.017111962	0.9828880	0.4850707246
## 45	44	1	0.018164015	0.50000	0.018000534	0.9819995	0.4767702126
## 46	45	1	0.017631463	0.50000	0.017477387	0.9825226	0.4681880940
## 47	46	1	0.016192501	0.50000	0.016062456	0.9839375	0.4600053895
## 48	47	1	0.015218651	0.50000	0.015103722	0.9848963	0.4526165733
## 49	48	1	0.014699593	0.50000	0.014592342	0.9854077	0.4457803786
## 50	49	1	0.014690661	0.50000	0.014583540	0.9854165	0.4392753990
## 51	50	1	0.016263498	0.50000	0.016132314	0.9838677	0.4328692088
## 52	51	1	0.019128983	0.50000	0.018947758	0.9810522	0.4258860268
## 53	52	1	0.022064302	0.50000	0.021823541	0.9781765	0.4178164416
## 54	53	1	0.024842156	0.50000	0.024537375	0.9754626	0.4086982073
## 55	54	1	0.027160869	0.50000	0.026796955	0.9732030	0.3986698262
## 56	55	1	0.026946452	0.50000	0.026588223	0.9734118	0.3879866887
## 57	56	1	0.025450854	0.50000	0.025131050	0.9748689	0.3776708123
## 58	57	1	0.025132208	0.50000	0.024820314	0.9751797	0.3681795480
## 59	58	1	0.025734980	0.50000	0.025408042	0.9745920	0.3590412162
## 60	59	1	0.027155129	0.50000	0.026791368	0.9732086	0.3499186819
## 61	60	1	0.030655627	0.50000	0.030192837	0.9698072	0.3405438818
## 62	61	1	0.035037345	0.50000	0.034434105	0.9655659	0.3302618959
## 63	62	1	0.039024625	0.50000	0.038277738	0.9617223	0.3188896229
## 64	63	1	0.043018911	0.50000	0.042113081	0.9578869	0.3066832496
## 65	64	1	0.046620443	0.50000	0.045558465	0.9544415	0.2937678730
## 66	65	1	0.048371720	0.50000	0.047229436	0.9527706	0.2803842596
## 67	66	1	0.049087418	0.50000	0.047911492	0.9520885	0.2671418692
## 68	67	1	0.050508626	0.50000	0.049264485	0.9507355	0.2543427036
## 69	68	1	0.052994078	0.50000	0.051626138	0.9483739	0.2418126413

## 70	69	1	0.056927854	0.50000	0.055352310	0.9446477	0.2293287885
## 71	70	1	0.064687268	0.50000	0.062660597	0.9373394	0.2166349103
## 72	71	1	0.075756151	0.50000	0.072991378	0.9270086	0.2030604375
## 73	72	1	0.087422487	0.50000	0.083761182	0.9162388	0.1882387764
## 74	73	1	0.099921124	0.50000	0.095166549	0.9048335	0.1724716741
## 75	74	1	0.112695743	0.50000	0.106684310	0.8933157	0.1560581400
## 76	75	1	0.116971061	0.50000	0.110507945	0.8894921	0.1394091851
## 77	76	1	0.113354284	0.50000	0.107274284	0.8927257	0.1240033625
## 78	77	1	0.110908148	0.50000	0.105080980	0.8949190	0.1107009905
## 79	78	1	0.110130738	0.50000	0.104382857	0.8956171	0.0990684220
## 80	79	1	0.114108805	0.50000	0.107949794	0.8920502	0.0887273770
## 81	80	1	0.128260135	0.50000	0.120530506	0.8794695	0.0791492750
## 82	81	1	0.148429341	0.50000	0.138174748	0.8618253	0.0696093729
## 83	82	1	0.168770237	0.50000	0.155636806	0.8443632	0.0599911153
## 84	83	1	0.188088353	0.50000	0.171920254	0.8280797	0.0506542897
## 85	84	1	0.202787763	0.50000	0.184119202	0.8158808	0.0419457914
## 86	85	1	0.208951957	0.50000	0.189186511	0.8108135	0.0342227657
## 87	86	1	0.212423347	0.50000	0.192027758	0.8079722	0.0277482801
## 88	87	1	0.215387989	0.50000	0.194447194	0.8055528	0.0224198401
## 89	88	1	0.219297659	0.50000	0.197627982	0.8023720	0.0180603651
## 90	89	1	0.227782307	0.50000	0.204492429	0.7955076	0.0144911316
## 91	90	1	0.282923355	0.50000	0.247860581	0.7521394	0.0115278049
## 92	91	1	0.302984205	0.50000	0.263123129	0.7368769	0.0086705165
## 93	92	1	0.324537269	0.50000	0.279227417	0.7207726	0.0063891030
## 94	93	1	0.347443352	0.50000	0.296018519	0.7039815	0.0046050903
## 95	94	1	0.370550393	0.50000	0.312628151	0.6873718	0.0032418983
## 96	95	1	0.393939394	0.50000	0.329113924	0.6708861	0.0022283896
## 97	96	1	0.420423892	0.50000	0.347396911	0.6526031	0.0014949956
## 98	97	1	0.452575488	0.50000	0.369061414	0.6309386	0.0009756387
## 99	98	1	0.484510533	0.50000	0.390024938	0.6099751	0.0006155681
## 100	99	1	0.518063837	0.50000	0.411477922	0.5885221	0.0003754812
## 101	100	NA	0.705835649	1.41676	1.000000000	0.0000000	0.0002209790
##			ndx	nLx		Tx	ex
## 1			0.2036797295	0.8981601352	3.810573e+01	38.105729	
## 2			0.0397301262	0.7764552074	3.720757e+01	46.724378	
## 3			0.0248614042	0.7441594422	3.643111e+01	48.151716	
## 4			0.0194478175	0.7220048313	3.568695e+01	48.770743	
## 5			0.0173655765	0.7035981343	3.496495e+01	49.088707	
## 6			0.0136013664	0.6881146628	3.426135e+01	49.302914	
## 7			0.0093294246	0.6766492673	3.357324e+01	49.277188	
## 8			0.0063251228	0.6688219936	3.289659e+01	48.954381	
## 9			0.0044188519	0.6634500063	3.222777e+01	48.414796	
## 10			0.0035747119	0.6594532244	3.156432e+01	47.734995	
## 11			0.0036789518	0.6558263926	3.090486e+01	46.991738	
## 12			0.0039909513	0.6519914410	3.024904e+01	46.253274	
## 13			0.0042186623	0.6478866342	2.959704e+01	45.534197	
## 14			0.0043367680	0.6436089190	2.894916e+01	44.828392	
## 15			0.0043165769	0.6392822466	2.830555e+01	44.128095	
## 16			0.0040826504	0.6350826330	2.766627e+01	43.423680	
## 17			0.0038165466	0.6311330345	2.703118e+01	42.700506	
## 18			0.0037057660	0.6273718781	2.640005e+01	41.956472	
## 19			0.0037483397	0.6236448253	2.577268e+01	41.202073	
## 20			0.0039037521	0.6198187794	2.514903e+01	40.447445	
## 21			0.0041695042	0.6157821513	2.452922e+01	39.699838	

```

## 22 0.0044616320 0.6114665832 2.391343e+01 38.966164
## 23 0.0046951698 0.6068881823 2.330197e+01 38.247864
## 24 0.0048350207 0.6021230870 2.269508e+01 37.541033
## 25 0.0049397779 0.5972356877 2.209296e+01 36.839670
## 26 0.0050023995 0.5922645990 2.149572e+01 36.141486
## 27 0.0050458411 0.5872404787 2.090346e+01 35.443799
## 28 0.0051578375 0.5821386394 2.031621e+01 34.745348
## 29 0.0053296596 0.5768948908 1.973408e+01 34.050117
## 30 0.0055591019 0.5714505101 1.915718e+01 33.361509
## 31 0.0059733867 0.5656842659 1.858573e+01 32.682750
## 32 0.0064622374 0.5594664538 1.802005e+01 32.024390
## 33 0.0068340246 0.5528183228 1.746058e+01 31.390635
## 34 0.0070293615 0.5458866298 1.690776e+01 30.774884
## 35 0.0069898912 0.5388770035 1.636188e+01 30.167259
## 36 0.0065063355 0.5321288901 1.582300e+01 29.554592
## 37 0.0059065546 0.5259224451 1.529087e+01 28.912027
## 38 0.0055558040 0.5201912658 1.476495e+01 28.232920
## 39 0.0054588314 0.5146839481 1.424476e+01 27.530707
## 40 0.0056263132 0.5091413758 1.373007e+01 26.818928
## 41 0.0062763011 0.5031900686 1.322093e+01 26.111383
## 42 0.0071514410 0.4964761976 1.271774e+01 25.432840
## 43 0.0078297525 0.4889856008 1.222126e+01 24.794587
## 44 0.0083005120 0.4809204686 1.173228e+01 24.186738
## 45 0.0085821186 0.4724791533 1.125136e+01 23.599121
## 46 0.0081827045 0.4640967417 1.077888e+01 23.022539
## 47 0.0073888162 0.4563109814 1.031478e+01 22.423177
## 48 0.0068361947 0.4491984760 9.858471e+00 21.781065
## 49 0.0065049796 0.4425278888 9.409273e+00 21.107418
## 50 0.0064061902 0.4360723039 8.966745e+00 20.412581
## 51 0.0069831819 0.4293776178 8.530672e+00 19.707275
## 52 0.0080695852 0.4218512342 8.101295e+00 19.022213
## 53 0.0091182342 0.4132573245 7.679444e+00 18.379946
## 54 0.0100283812 0.4036840168 7.266186e+00 17.778855
## 55 0.0106831374 0.3933282575 6.862502e+00 17.213498
## 56 0.0103158764 0.3828287505 6.469174e+00 16.673701
## 57 0.0094912642 0.3729251802 6.086345e+00 16.115477
## 58 0.0091383319 0.3636103821 5.713420e+00 15.518027
## 59 0.0091225343 0.3544799490 5.349810e+00 14.900266
## 60 0.0093748001 0.3452312818 4.995330e+00 14.275687
## 61 0.0102819859 0.3354028888 4.650098e+00 13.654917
## 62 0.0113722730 0.3245757594 4.314696e+00 13.064467
## 63 0.0122063733 0.3127864363 3.990120e+00 12.512542
## 64 0.0129153766 0.3002255613 3.677333e+00 11.990656
## 65 0.0133836134 0.2870760663 3.377108e+00 11.495838
## 66 0.0132423904 0.2737630644 3.090032e+00 11.020703
## 67 0.0127991656 0.2607422864 2.816269e+00 10.542221
## 68 0.0125300623 0.2480776724 2.555526e+00 10.047571
## 69 0.0124838528 0.2355707149 2.307449e+00 9.542300
## 70 0.0126938782 0.2229818494 2.071878e+00 9.034531
## 71 0.0135744728 0.2098476739 1.848896e+00 8.534618
## 72 0.0148216611 0.1956496069 1.639048e+00 8.071727
## 73 0.0157671023 0.1803552252 1.443399e+00 7.667914
## 74 0.0164135341 0.1642649070 1.263044e+00 7.323194
## 75 0.0166489549 0.1477336625 1.098779e+00 7.040829

```

```
## 76 0.0154058226 0.1317062738 9.510450e-01 6.821968
## 77 0.0133023720 0.1173521765 8.193387e-01 6.607391
## 78 0.0116325685 0.1048847063 7.019866e-01 6.341285
## 79 0.0103410450 0.0938978995 5.971019e-01 6.027166
## 80 0.0095781020 0.0839383260 5.032040e-01 5.671349
## 81 0.0095399022 0.0743793239 4.192656e-01 5.297151
## 82 0.0096182575 0.0648002441 3.448863e-01 4.954596
## 83 0.0093368256 0.0553227025 2.800861e-01 4.668793
## 84 0.0087084984 0.0463000405 2.247634e-01 4.437203
## 85 0.0077230257 0.0380842785 1.784633e-01 4.254618
## 86 0.0064744856 0.0309855229 1.403790e-01 4.101920
## 87 0.0053284400 0.0250840601 1.093935e-01 3.942353
## 88 0.0043594750 0.0202401026 8.430947e-02 3.760485
## 89 0.0035692335 0.0162757483 6.406936e-02 3.547512
## 90 0.0029633267 0.0130094682 4.779362e-02 3.298129
## 91 0.0028572884 0.0100991607 3.478415e-02 3.017413
## 92 0.0022814134 0.0075298097 2.468499e-02 2.847003
## 93 0.0017840127 0.0054970967 1.715518e-02 2.685068
## 94 0.0013631920 0.0039234943 1.165808e-02 2.531564
## 95 0.0010135087 0.0027351440 7.734586e-03 2.385820
## 96 0.0007333941 0.0018616926 4.999442e-03 2.243522
## 97 0.0005193568 0.0012353171 3.137750e-03 2.098835
## 98 0.0003600706 0.0007956034 1.902432e-03 1.949935
## 99 0.0002400869 0.0004955247 1.106829e-03 1.798061
## 100 0.0001545022 0.0002982301 6.113044e-04 1.628056
## 101 0.0002209790 0.0003130743 3.130743e-04 1.416760
##
## $lt_summary
##      e0      cv
## 38.10573 0.81675
```

Now we need to apply the `LTFun()` to every year in our data, i.e. to every column of our matrix `M`. We start by preparing an empty list with 264 entries – one entry for each year in our data set. This list is where we will store the output of the life-table function for each year.

```
# number of years
k <- length(years)

lts <- vector(mode = 'list', length = k)
```

Next, we use a for-loop to iterate over all the different columns in our matrix `M`. For each column we apply the `LTFun()` and store its output in the corresponding entry of `lts`.

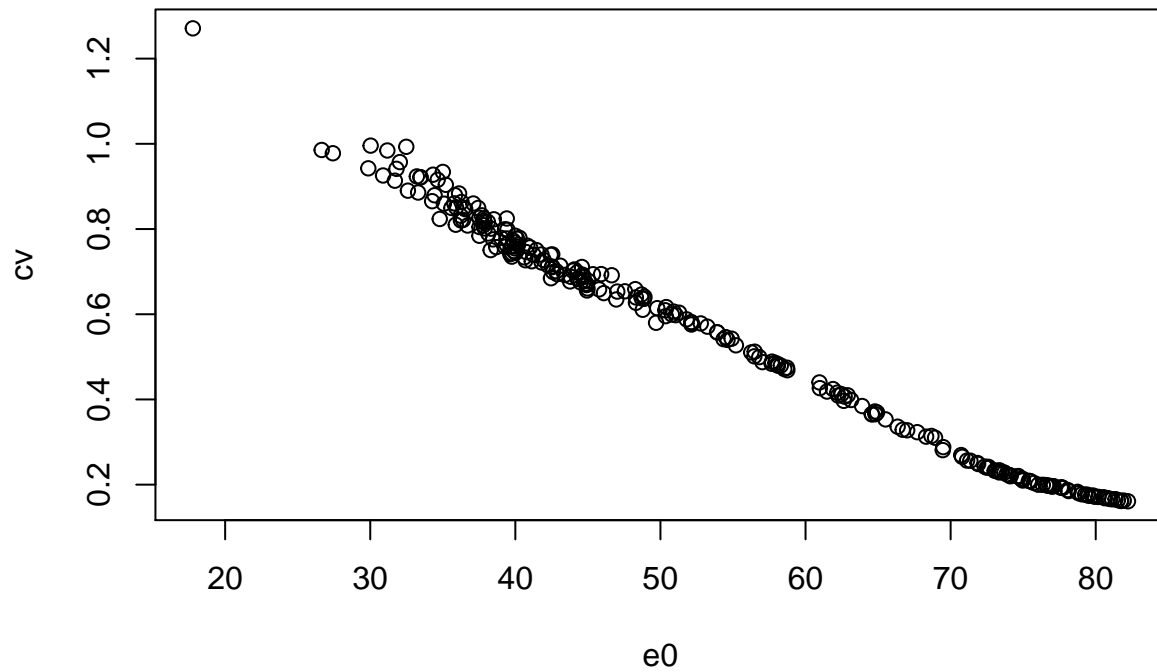
```
for (i in 1:k) {
  lts[[i]] <- LTFun(x = 0:100, nm = M[,i])
}
```

What we now have is a list of 264 entries, each entry featuring a life-table and some life-table summary statistics. We can use the function `sapply()` to go over each entry of the list and extract the total life-expectancy and the CV. You could achieve the same by writing another for loop.

```
e0 <- vector(mode = 'numeric', length = k)
for (i in 1:k) {
  e0[i] <- lts[[i]]$lt_summary['e0']
}
cv <- vector(mode = 'numeric', length = k)
```

```
for (i in 1:k) {  
  cv[i] <- lts[[i]]$lt_summary['cv']  
}
```

```
plot(e0, cv)
```



## Literature

Preston, S. H., Heuveline, P., & Guillot, M. (2001). Demography. Oxford, UK: Blackwell.