

MortalityLaws User Manual

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1 Package structure

MortalityLaws is an **R** package which exploits the available optimization methods to provide tools for fitting and analyzing a wide range of complex mortality models. The package can be used to construct full and abridge life tables given various input indices and to download data from Human Mortality Database as well. The main functions in the package are: **MortalityLaw**, **LifeTable** and **ReadHMD**. The package provides also support functions like **availableLaws**, **availableLF** and **availableHMD** that return information about the mortality laws implemented in the package and loss functions used in optimization process and HMD data availability. Generic functions like **predict**, **plot**, **summary**, **fitted.values** are created for **MortalityLaws** objects. Small data set for testing purposes **ahmd** is saved in the package.

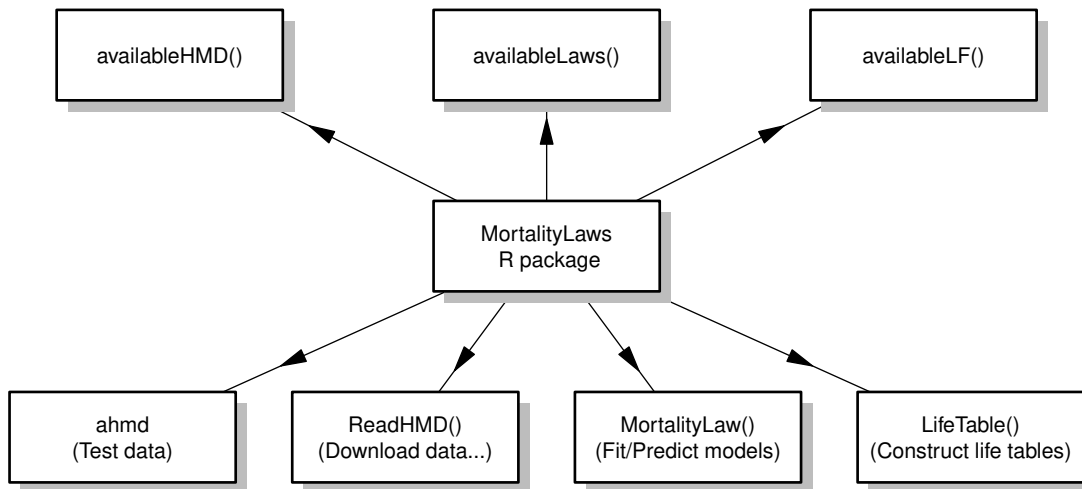


Figure 1: MortalityLaws R Package Structure

2 Downloading HMD Data

Download data form Human Mortality Database (2016) using the **ReadHMD** function:

```
library(MortalityLaws)
```

```
# Download HMD data - death counts
HMD_Dx <- ReadHMD(what = "Dx",
                  countries = "SWE",           # HMD country code for Sweden
                  interval = "1x1",           # specify data format
                  username = "user@email.com", # here add your HMD username
                  password = "password",       # here add your password account
                  save = FALSE)                # save data outside R
```

Here we download all the registered death counts **Dx** in Sweden from 1751 until 2014. In the same way one can download the following records: birth records **births**, exposure-to-risk **Ex**, deaths by Lexis triangles **lexis**, population size **population**, death-rates **mx**, life tables for females

LT_f, life tables for males LT_m, life tables both sexes combined LT_t, life expectancy at birth e0, cohort death-rates mxc and cohort exposures Exc for over 39 countries and regions in different formats.

3 Model fitting and diagnosis

Once we have data from HMD or other sources we can start analyzing it. For example, let's fit a (Heligman and Pollard 1980) model under a Poisson setting which is already implemented as one of the standard models in the package. We have to use the MortalityLaw function in this regard.

```
year    <- 1950
ages    <- 0:100
deaths  <- ahmd$Dx[paste(ages), paste(year)]
exposure <- ahmd$Ex[paste(ages), paste(year)]

fit <- MortalityLaw(x = ages,
                    Dx = deaths, # vector with death counts
                    Ex = exposure, # vector containing exposures
                    law = "HP",
                    opt.method = "LF2")

# inspect the output object
ls(fit)
## [1] "coefficients"      "fitted.values"      "goodness.of.fit"    "info"
## [5] "input"             "opt.diagnosis"      "residuals"          "stats"
```

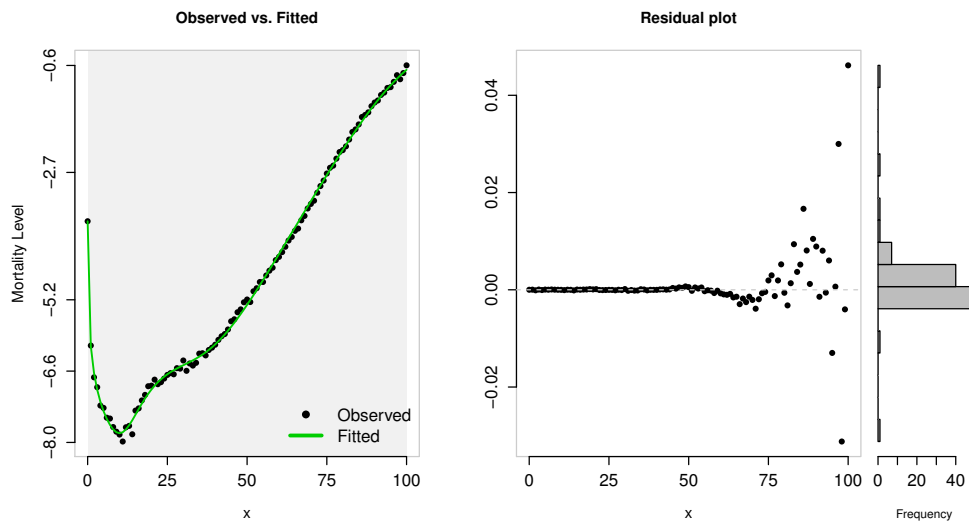
A summary can be obtained using the summary function:

```
summary(fit)
## Heligman-Pollard model:  $q[x]/p[x] = A^[(x + B)^C] + D \exp[-E \log(x/F)^2] + G H^x$ 
## Fitted values: mx
##
## Call: MortalityLaw(x = ages, Dx = deaths, Ex = exposure, law = "HP",
##   opt.method = "LF2")
##
## Parameters:
##      Estimate Std. Error t value Pr(>|t|)
## A  2.227e-03  7.391e-03   0.301   0.764
## B  1.461e-02  3.191e-01   0.046   0.964
## C  1.229e-01  5.363e-01   0.229   0.819
## D  9.122e-04  2.265e-03   0.403   0.688
## E  2.757e+00  1.762e+01   0.156   0.876
## F_ 2.901e+01  3.504e+01   0.828   0.410
## G  2.060e-05  2.703e-06  7.622 2.07e-11 ***
## H  1.114e+00  1.571e-03 709.358 < 2e-16 ***
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Deviance Residuals:
##    Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -0.0312 -0.0002  0.0000  0.0009  0.0003  0.0462
## Residual standard error: 0.007381 on 93 degrees of freedom
```

The standard plot helps us to investigate visually the goodness of fit.

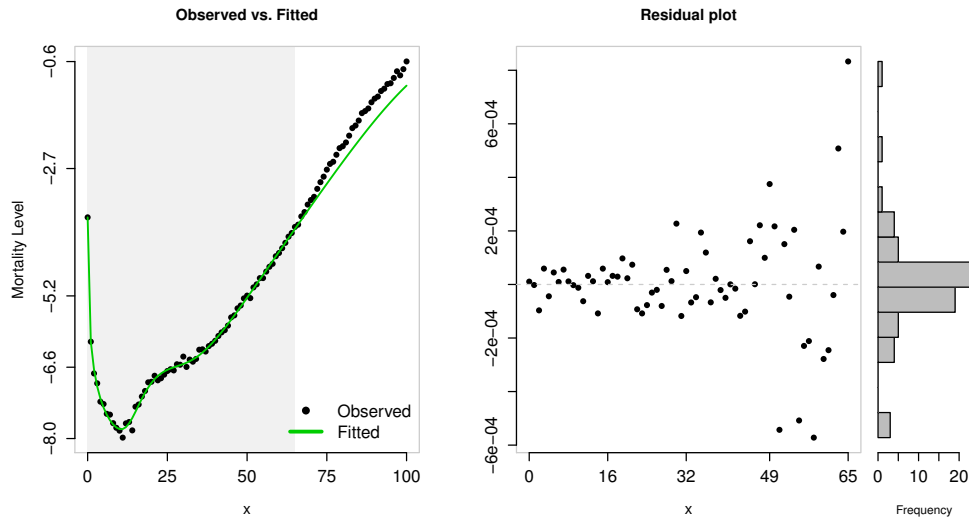
```
plot(fit)
```



A model can be fitted using a subset of the data only by specifying in `fit.this.x` age range to be covered:

```
fit.subset <- MortalityLaw(x    = ages,
                          Dx    = deaths,
                          Ex    = exposure,
                          law    = "HP",
                          opt.method = "LF2",
                          fit.this.x = 0:65)

plot(fit.subset)
```



The gray area on the plot showing the fitted value indicates the age range used in fitting the model.

4 Mortality laws in the package

In **R** one can check the availability of the implemented models using `availableLaws`:

```
availableLaws()
```

See table 1.

5 Loss function in the package

A parametric model is fitted by optimizing a loss function e.g. a likelihood function or a function that minimizes errors. In **MortalityLaws** 8 such functions are implemented and can be used to better capture different portions of a mortality curve. Check `availableLF` for more details.

```
availableLF()
##
## Loss functions available in the package:
##
## LOSS FUNCTION                                CODE
## L = -[Dx * log(mu) - mu*Ex]                  poissonL
## L = -[Dx * log(1 - exp(-mu)) - (Ex - Dx)*mu] binomialL
## L = [1 - mu/ov]^2                             LF1
## L = log[mu/ov]^2                             LF2
## L = [(ov - mu)^2]/ov                         LF3
## L = [ov - mu]^2                             LF4
## L = [ov - mu] * log[ov/mu]                   LF5
## L = abs(ov - mu)                             LF6
```

```
##
## LEGEND:
## Dx: Death counts
## Ex: Population exposed to risk
## mu: Estimated value
## ov: Observed value
```

6 Custom mortality laws

Now let's fit a mortality law that is not defined in the package, say a reparametrize version of Gompertz in terms of modal age at death (Missov et al. 2015),

$$\mu_x = \beta e^{\beta(x-M)}. \quad (1)$$

We have to define a function containing the desired hazard function and then using the `custom.law` argument it can be used in the `MortalityLaw` function.

```
# Here we define a function for our new model and provide start parameters
my_gompertz <- function(x, par = c(b = 0.13, M = 45)){
  hx <- with(as.list(par), b*exp(b*(x - M)) )
  # return everything inside this function
  return(as.list(environment()))
}
```

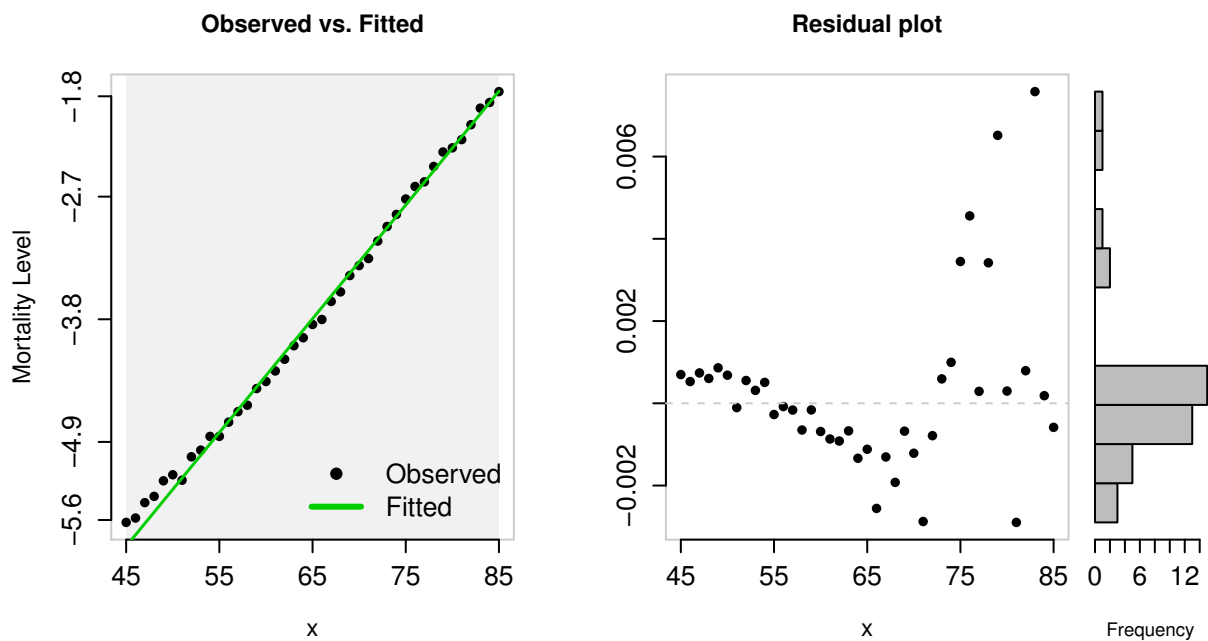
```
# Select data
year <- 1950
ages <- 45:85
deaths <- ahmd$Dx[paste(ages), paste(year)]
exposure <- ahmd$Ex[paste(ages), paste(year)]
```

```
# Use 'custom.law' argument to instruct the MortalityLaw function how to behave
my_model <- MortalityLaw(x = ages,
                        Dx = deaths,
                        Ex = exposure,
                        custom.law = my_gompertz)
```

```
summary(my_model)
## Custom Mortality Law
## Fitted values: mx
##
## Call: MortalityLaw(x = ages, Dx = deaths, Ex = exposure, custom.law = my_gompertz)
##
## Parameters:
##      Estimate Std. Error t value Pr(>|t|)
## b  0.102045   0.001123   90.87  <2e-16 ***
```

```
## M 79.828162    0.127930  624.00   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## logLik: 13.62    AIC: -23.24 BIC: -19.81
##
## Deviance Residuals:
##    Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -0.0029 -0.0008 -0.0001  0.0003  0.0007  0.0076
## Residual standard error: 0.002193 on 39 degrees of freedom
```

```
plot(my_model)
```



7 Full life tables

Using `LifeTable` function one can build full or abridge life table with various input choices like: death counts and mid-interval population estimates (Dx , Ex) or age-specific death rates (mx) or death probabilities (qx) or survivorship curve (lx) or a distribution of deaths (dx). If one of these options are specified, the other can be ignored.

```
# Life table for year of 1900
y <- 1900
x <- as.numeric(rownames(ahmd$mx))
Dx <- ahmd$Dx[, paste(y)]
Ex <- ahmd$Ex[, paste(y)]

LT1 <- LifeTable(x, Dx = Dx, Ex = Ex)
```

```

LT2 <- LifeTable(x, mx = LT1$lt$mx)
LT3 <- LifeTable(x, qx = LT1$lt$qx)
LT4 <- LifeTable(x, lx = LT1$lt$lx)
LT5 <- LifeTable(x, dx = LT1$lt$dx)

LT1
##
## Full Life Table
##
## Number of life tables: 1
## Dimension: 111 x 10
## Age intervals: [0,1) [1,2) [2,3) ... ... [108,109) [109,110) [110,+)
##
##      x.int  x      mx      qx      ax      lx      dx      Lx      Tx      ex
##      [0,1)  0 0.1513 0.1404 0.49 1e+05 14042 92802 4808384 51.62
##      [1,2)  1 0.0479 0.0468 0.5 85958 4023 83931 4715582 56.16
##      [2,3)  2 0.0191 0.0189 0.5 81935 1552 81157 4631651 57.07
##      [3,4)  3 0.013 0.0129 0.5 80383 1041 79861 4550494 56.98
##      [4,5)  4 0.0094 0.0093 0.5 79342 740 78971 4470633 56.61
##      [5,6)  5 0.0067 0.0067 0.5 78602 523 78340 4391661 56.06
##      <NA> ...      ...      ...      ...      ...      ...      ...      ...      ...
## [108,109) 108      0      1 0.15      0      0      0      0      0
## [109,110) 109      0      1 0.15      0      0      0      0      0
## [110,+) 110      0      1 0.15      0      0      0      0      0

ls(LT1)
## [1] "call"          "lt"             "process_date"

```

8 Abridge life tables

```

# Example
x <- c(0, 1, seq(5, 110, by = 5))
mx <- c(.053, .005, .001, .0012, .0018, .002, .003, .004,
        .004, .005, .006, .0093, .0129, .019, .031, .049,
        .084, .129, .180, .2354, .3085, .390, .478, .551)

lt <- LifeTable(x, mx = mx, sex = "female")

lt
##
## Abridge Life Table
##
## Number of life tables: 1
## Dimension: 24 x 10
## Age intervals: [0,1) [1,5) [5,10) ... ... [100,105) [105,110) [110,+)

```



```
##
##      x.int    x      mx      qx      ax      lx      dx      Lx      Tx      ex
##      [0,1)    0  0.053 0.0516  0.2 1e+05 5162  95878 6485467 65.54
##      [1,5)    1  0.005 0.0198  1.44 94838 1878 374547 6389590 67.86
##      [5,10)   5  0.001 0.005   2.5 92960 464 463640 6015042 64.87
##      [10,15) 10 0.0012 0.006   2.5 92496 553 461098 5551402 60.2
##      [15,20) 15 0.0018 0.009   2.5 91943 824 457653 5090304 55.61
##      [20,25) 20 0.002   0.01   2.5 91119 907 453326 4632651 51.1
##      <NA>    ...      ...      ...      ...      ...      ...      ...      ...      ...
##      [100,105) 100  0.39 0.8577  1.73  407 349   896   1014  3.55
##      [105,110) 105  0.478 0.9084  1.59   58  53   110    119  2.88
##      [110,+) 110  0.551      1  1.59   5   5    8     8  1.59
```

References

Heligman, Larry, and John H Pollard. 1980. “The Age Pattern of Mortality.” *Journal of the Institute of Actuaries* 107 (01). Cambridge Univ Press: 49–80.

Missov, Trifon I, Adam Lenart, Laszlo Nemeth, Vladimir Canudas-Romo, and James W Vaupel. 2015. “The Gompertz Force of Mortality in Terms of the Modal Age at Death.” *Demographic Research* 32: 1031–48.

University of California Berkeley, USA, and Max Planck Institute for Demographic Research, Germany. 2016. “Human Mortality Database.” <http://www.mortality.org/>.

Table 1: Parametric functions build in the MortalityLaws package

Mortality laws	Predictor	Code
Gompertz	$\mu(x) = Ae^{Bx}$	gompertz
Gompertz	$\mu(x) = \frac{1}{\sigma} \exp\left\{\frac{x-M}{\sigma}\right\}$	gompertz0
Inverse-Gompertz	$\mu(x) = \frac{1}{\sigma} \exp\left\{\frac{x-M}{\sigma}\right\} / \left(\exp\left\{e^{\frac{-(x-M)}{\sigma}}\right\} - 1\right)$	invgompertz
Makeham	$\mu(x) = Ae^{Bx} + C$	makeham
Makeham	$\mu(x) = \frac{1}{\sigma} \exp\left\{\frac{x-M}{\sigma}\right\} + C$	makeham0
Opperman	$\mu(x) = \frac{A}{\sqrt{x}} + B + C\sqrt[3]{x}$	opperman
Thiele	$\mu(x) = A_1 e^{-B_1 x} + A_2 e^{-\frac{1}{2} B_2 (x-C)^2} + A_3 e^{B_3 x}$	thiele
Wittstein & Bumstead	$q(x) = \frac{1}{B} A^{-(Bx)^N} + A^{-(M-x)^N}$	wittstein
Perks	$\mu(x) = (A + BC^x) / (BC^{-x} + 1 + DC^x)$	perks
Weibull	$\mu(x) = \frac{1}{\sigma} \left(\frac{x}{M}\right)^{\frac{M}{\sigma}-1}$	weibull
Inverse-Weibull	$\mu(x) = \frac{1}{\sigma} \left(\frac{x}{M}\right)^{-\frac{M}{\sigma}-1} / \left(\exp\left\{\left(\frac{x}{M}\right)^{-\frac{M}{\sigma}}\right\} - 1\right)$	invweibull
Van der Maen	$\mu(x) = A + Bx + Cx^2 + I/(N - x)$	vandermaen
Van der Maen	$\mu(x) = A + Bx + I/(N - x)$	vandermaen2
Quadratic	$\mu(x) = A + Bx + Cx^2$	quadratic
Beard	$\mu(x) = Ae^{Bx} / (1 + KAe^{Bx})$	beard
Makeham-Beard	$\mu(x) = Ae^{Bx} / (1 + KAe^{Bx}) + C$	makehambeard
Siler	$\mu(x) = A_1 e^{-B_1 x} + A_2 + A_3 e^{B_3 x}$	siler
Heligman - Pollard	$q(x)/p(x) = A^{(x+B)^C} + De^{-E(\ln x - \ln F)^2} + GH^x$	HP
Heligman - Pollard	$q(x) = A^{(x+B)^C} + De^{-E(\ln x - \ln F)^2} + \frac{GH^x}{1+GH^x}$	HP2
Heligman - Pollard	$q(x) = A^{(x+B)^C} + De^{-E(\ln x - \ln F)^2} + \frac{GH^x}{1+KGH^x}$	HP3
Heligman - Pollard	$q(x) = A^{(x+B)^C} + De^{-E(\ln x - \ln F)^2} + \frac{GH^{x^K}}{1+GH^{x^K}}$	HP3
Rogers-Planck	$q(x) = A_0 + A_1 e^{-Ax} + A_2 e^{\{B(x-U) - e^{-C(x-U)}\}} + A_3 e^{Dx}$	rogersplanck
Martinelle	$\mu(x) = (Ae^{Bx} + C) / (1 + De^{Bx}) + Ke^{Bx}$	martinelle
Carriere	$S(x) = \psi_1 S_1(x) + \psi_2 S_2(x) + \psi_3 S_3(x)$	carriere1
Carriere	$S(x) = \psi_1 S_1(x) + \psi_4 S_4(x) + \psi_3 S_3(x)$	carriere2
Kostaki	$q(x)/p(x) = A^{(x+B)^C} + De^{-E_i(\ln x - \ln F)^2} + GH^x$	kostaki
Kannisto	$\mu(x) = Ae^{Bx} / (1 + Ae^{Bx}) + C$	kannisto