

Validate performance of Lee-Carter method to forecast mortality

Afternoon lab on *Forecasting mortality*

As part of IMPRS-PHDS course on *Population Health*

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Source: <https://github.com/christina-bohk-ewald/2019-IMPRS-PHDS-forecasting-mortality>

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1. Some preparations in R

Open a new script in R and save it to a folder of your choice.

Create a filepath to a folder where you would like to save your outcome. For example,

```
the.plot.path <- c("C:/plots")
```

Set the working directory to this outcome path

```
setwd(the.plot.path)
```

2. Load and explore US female mortality

Install and load the fds package to access data of the Human Mortality Database

```
library(fds)
```

Load US female mortality by age, 0 through 110, and calendar year, 1933 through 2017, from the HMD

```
usa_wom <- read.hmd(country=params$country, sex="Female", file = "Mx_1x1.txt",  
username=your_username, password=your_password, yname="mortality rates")
```

Explore data object usa_wom. For example,

```
head(usa_wom[[2]][,as.character(1933:1940)])
```

##	1933	1934	1935	1936	1937	1938	1939	1940
## 0	0.054177	0.060211	0.053557	0.055160	0.053895	0.050748	0.046848	0.046847
## 1	0.008866	0.009894	0.008334	0.008082	0.007856	0.007401	0.005904	0.005263
## 2	0.004025	0.004540	0.003766	0.003923	0.003690	0.003322	0.002644	0.002471
## 3	0.002869	0.002979	0.002756	0.002675	0.002513	0.002233	0.001896	0.001679
## 4	0.002230	0.002236	0.002326	0.002154	0.001951	0.001785	0.001544	0.001359
## 5	0.001852	0.001858	0.001929	0.001758	0.001604	0.001458	0.001310	0.001158

3. Function to estimate and forecast mortality with the Lee-Carter method:

```
estimate_forecast_mortality_with_LC <- function(obs_mortality,
                                              base_start,
                                              base_end,
                                              number_sim,
                                              years_ahead){

  ## 1. fit mortality in base period:

  log_usa_wom <- log(obs_mortality[,as.character(base_start:base_end)],base=exp(1))
  mean_log_usa_wom <- rowMeans(log_usa_wom)
  centered_log_usa_wom <- log_usa_wom - mean_log_usa_wom
  svd_centered_log_usa_wom <- svd(t(centered_log_usa_wom))
  sum_v <- sum(svd_centered_log_usa_wom$v[,1])
  bx <- svd_centered_log_usa_wom$v[,1]/sum_v
  kt <- svd_centered_log_usa_wom$d[1] * svd_centered_log_usa_wom$u[,1] * sum_v
  names(kt) <- (base_start+1):base_end
  ax <- mean_log_usa_wom

  ## 2. forecast mortality:

  fit <- summary(lm(kt[2:length(kt)]-kt[1:(length(kt)-1)] ~ 1))
  kt_drift <- fit$coefficients[1,1]
  sec <- fit$coefficients[1,2]
  see <- fit$sigma
  fitted <- kt[1:(length(kt)-1)] + kt_drift
  residuals <- kt[2:length(kt)]-(kt[1:(length(kt)-1)]+kt_drift)
  kt_fitted_self <- fitted
  set.seed(1234)
  kt_trajectories <- matrix(NA,nr=number_sim,nc=years_ahead)
  for(run in 1:number_sim){
```

```

for(year in 1:1){

  current.epsilon <- rnorm(n=1,mean=0,sd=see)

  kt_trajectories[run,year] <- kt[length(kt)] + kt_drift + current.epsilon*year
}

for(year in 2:years_ahead){

  current.epsilon <- rnorm(n=1,mean=0,sd=see)

  kt_trajectories[run,year] <- kt_trajectories[run,(year-1)] + kt_drift
+ current.epsilon

}

}

kt_trajectories_median <- apply(kt_trajectories,2,median)

kt_trajectories_low <- apply(X=kt_trajectories,2,
FUN=function(X){quantile(X,probs=0.1)})

kt_trajectories_up <- apply(X=kt_trajectories,2,
FUN=function(X){quantile(X,probs=0.9)})

## 3. normalize kt:

kt_base_forecast <- c(kt_fitted_self , kt_trajectories_median)

names(kt_base_forecast) <- (base_start+1):(base_end+years_ahead)

kt_base_forecast_normalized <- kt_base_forecast -
kt_base_forecast[as.character(base_end)]

kt_base_forecast_low <- c(kt_fitted_self , kt_trajectories_low)

names(kt_base_forecast_low) <- (base_start+1):(base_end+years_ahead)

kt_base_forecast_low_normalized <- kt_base_forecast_low -
kt_base_forecast_low[as.character(base_end)]

kt_base_forecast_up <- c(kt_fitted_self , kt_trajectories_up)

names(kt_base_forecast_up) <- (base_start+1):(base_end+years_ahead)

kt_base_forecast_up_normalized <- kt_base_forecast_up -
kt_base_forecast_up[as.character(base_end)]

## 4. save LC estimated mortality:

mx_LC_estimated <- matrix(NA,nr=nrow(obs_mortality),

```

```

nc=length((base_start+1):base_end))

  colnames(mx_LC_estimated) <- (base_start+1):base_end

  rownames(mx_LC_estimated) <- names(obs_mortality[,1])

  for(year in (base_start+1):base_end){

    mx_LC_estimated[,as.character(year)] <- ax+bx*kt_base_forecast[as.character(year)]

  }

  ## 5. LC forecasted mortality:

  mx_LC_forecasted <- matrix(NA,nr=nrow(obs_mortality),
nc=length((base_end+1):(base_end+years_ahead)))

  colnames(mx_LC_forecasted) <- (base_end+1):(base_end+years_ahead)

  rownames(mx_LC_forecasted) <- names(obs_mortality[,1])

  mx_LC_forecasted_low <- matrix(NA,nr=nrow(obs_mortality),
nc=length((base_end+1):(base_end+years_ahead)))

  colnames(mx_LC_forecasted_low) <- (base_end+1):(base_end+years_ahead)

  rownames(mx_LC_forecasted_low) <- names(obs_mortality[,1])

  mx_LC_forecasted_up <- matrix(NA,nr=nrow(obs_mortality),
nc=length((base_end+1):(base_end+years_ahead)))

  colnames(mx_LC_forecasted_up) <- (base_end+1):(base_end+years_ahead)

  rownames(mx_LC_forecasted_up) <- names(obs_mortality[,1])

  for(year in (base_end+1):(base_end+years_ahead)){

    mx_LC_forecasted[,as.character(year)] <- ax+
bx*kt_base_forecast[as.character(year)]

    mx_LC_forecasted_low[,as.character(year)] <- ax+
bx*kt_base_forecast_low[as.character(year)]

    mx_LC_forecasted_up[,as.character(year)] <- ax+
bx*kt_base_forecast_up[as.character(year)]

  }

  ## 6. LC forecasted mortality, corrected for jump-off bias:

  mx_LC_forecasted_joyAdjusted <- matrix(NA,nr=nrow(obs_mortality),
nc=length((base_end+1):(base_end+years_ahead)))

```

```

colnames(mx_LC_forecasted_joyAdjusted) <- (base_end+1):(base_end+years_ahead)

rownames(mx_LC_forecasted_joyAdjusted) <- names(obs_mortality[,1])

mx_LC_forecasted_low_joyAdjusted <- matrix(NA,nr=nrow(obs_mortality),
nc=length((base_end+1):(base_end+years_ahead)))

colnames(mx_LC_forecasted_low_joyAdjusted) <- (base_end+1):(base_end+years_ahead)

rownames(mx_LC_forecasted_low_joyAdjusted) <- names(obs_mortality[,1])

mx_LC_forecasted_up_joyAdjusted <- matrix(NA,nr=nrow(obs_mortality),
nc=length((base_end+1):(base_end+years_ahead)))

colnames(mx_LC_forecasted_up_joyAdjusted) <- (base_end+1):(base_end+years_ahead)

rownames(mx_LC_forecasted_up_joyAdjusted) <- names(obs_mortality[,1])

for(year in (base_end+1):(base_end+years_ahead)){

  mx_LC_forecasted_joyAdjusted[,as.character(year)] <-
log_usa_wom[,as.character(base_end)]+ bx*kt_base_forecast_normalized[as.character(year)]

  mx_LC_forecasted_low_joyAdjusted[,as.character(year)] <-
log_usa_wom[,as.character(base_end)]+bx*kt_base_forecast_low_normalized[as.character(year)]

  mx_LC_forecasted_up_joyAdjusted[,as.character(year)] <-
log_usa_wom[,as.character(base_end)]+bx*kt_base_forecast_up_normalized[as.character(year)]

}

## 7. save and return selected results:

outcome_to_return <- list()

outcome_to_return$obs_mortality <- obs_mortality

outcome_to_return$base_start <- base_start

outcome_to_return$base_end <- base_end

outcome_to_return$number_sim <- number_sim

outcome_to_return$years_ahead <- years_ahead

outcome_to_return$alphax <- ax

outcome_to_return$betax <- bx

outcome_to_return$kappat <- kt

outcome_to_return$kappat_drift <- kt_drift

```

```

outcome_to_return$kt_fitted <- kt_fitted_self

outcome_to_return$kappat_median <- kt_trajectories_median

outcome_to_return$kappat_low80 <- kt_trajectories_low

outcome_to_return$kappat_up80 <- kt_trajectories_up

outcome_to_return$kappat_base_forecast <- kt_base_forecast

outcome_to_return$kappat_base_forecast_low80 <- kt_base_forecast_low

outcome_to_return$kappat_base_forecast_up80 <- kt_base_forecast_up

outcome_to_return$kappat_base_forecast_normalized <- kt_base_forecast_normalized

outcome_to_return$kappat_base_forecast_low80_normalized <-
kt_base_forecast_low_normalized

outcome_to_return$kappat_base_forecast_up80_normalized <-
kt_base_forecast_up_normalized

outcome_to_return$mx_LC_estimated <- mx_LC_estimated

outcome_to_return$mx_LC_forecasted <- mx_LC_forecasted

outcome_to_return$mx_LC_forecasted_low80 <- mx_LC_forecasted_low

outcome_to_return$mx_LC_forecasted_up80 <- mx_LC_forecasted_up

outcome_to_return$mx_LC_forecasted_joyAdjusted <- mx_LC_forecasted_joyAdjusted

outcome_to_return$mx_LC_forecasted_low80_joyAdjusted <-
mx_LC_forecasted_low_joyAdjusted

outcome_to_return$mx_LC_forecasted_up80_joyAdjusted <-
mx_LC_forecasted_up_joyAdjusted

return(outcome_to_return)
}

```


4. Function to calculate life expectancy at birth

```
lexp.age.specific <- function(mx,ax){  
  qx <- mx/(1+(1-ax)*mx)  
  px <- 1-qx  
  lx <- c(100000, (cumprod(px)*100000)[1:(length(qx)-1)])  
  dx <- c(-diff(lx), lx[length(lx)])  
  Lx1 <- lx[-1]+ax[1:(length(ax)-1)]*dx[1:(length(dx)-1)]  
  Lx2 <- if(mx[length(mx)] == 0){  
    0}else{  
      dx[length(dx)]/mx[length(mx)]  
    }  
  Lx <- c(Lx1, Lx2)  
  Tx <- rev(cumsum(rev(Lx)))  
  ex <- Tx/lx  
  ex[is.nan(ex)] <- 0  
  return(ex)  
}
```

5. Function to translate LC mortality into LC life expectancy at birth

Based on output of function *estimate_forecast_mortality_with_LC*.

```
estimate_forecast_e0_with_LC <- function(outcome_estimate_forecast_mortality_with_LC){  
  
  base_start <- outcome_estimate_forecast_mortality_with_LC$base_start  
  
  base_end <- outcome_estimate_forecast_mortality_with_LC$base_end  
  
  years_ahead <- outcome_estimate_forecast_mortality_with_LC$years_ahead  
  
  ## load function: lexp.age.specific  
  
  lt_ax <- c(0.06,  
rep(0.5,(length(outcome_estimate_forecast_mortality_with_LC$mx_LC_estimated[,1])-1)))  
  
  e0_observed <- matrix(NA,nr=1,nc=length((base_start+1):(base_end)))  
  
  colnames(e0_observed) <- ((base_start+1):(base_end))  
  
  e0_LC_estimated <- matrix(NA,nr=1,nc=length((base_start+1):(base_end)))  
  
  colnames(e0_LC_estimated) <- ((base_start+1):(base_end))  
  
  e0_LC_forecast <- matrix(NA,nr=1,nc=years_ahead)  
  
  colnames(e0_LC_forecast) <- (base_end+1):(base_end+years_ahead)  
  
  e0_LC_forecast_low80 <- matrix(NA,nr=1,nc=years_ahead)  
  
  colnames(e0_LC_forecast_low80) <- (base_end+1):(base_end+years_ahead)  
  
  e0_LC_forecast_up80 <- matrix(NA,nr=1,nc=years_ahead)  
  
  colnames(e0_LC_forecast_up80) <- (base_end+1):(base_end+years_ahead)  
  
  e0_LC_forecast_joyAdjusted <- matrix(NA,nr=1,nc=years_ahead)  
  
  colnames(e0_LC_forecast_joyAdjusted) <- (base_end+1):(base_end+years_ahead)  
  
  e0_LC_forecast_low80_joyAdjusted <- matrix(NA,nr=1,nc=years_ahead)  
  
  colnames(e0_LC_forecast_low80_joyAdjusted) <- (base_end+1):(base_end+years_ahead)  
  
  e0_LC_forecast_up80_joyAdjusted <- matrix(NA,nr=1,nc=years_ahead)  
  
  colnames(e0_LC_forecast_up80_joyAdjusted) <- (base_end+1):(base_end+years_ahead)  
  
  obs_mortality <- outcome_estimate_forecast_mortality_with_LC$obs_mortality  
  
  mx_LC_estimated <- outcome_estimate_forecast_mortality_with_LC$mx_LC_estimated  
  
  for(year in (base_start+1):base_end){
```

```

    e0_observed[1,as.character(year)] <-
lexp.age.specific(obs_mortality[,as.character(year)],lt_ax)[1]

    e0_LC_estimated[1,as.character(year)] <-
lexp.age.specific(exp(mx_LC_estimated[,as.character(year)]),lt_ax)[1]

  }

  mx_LC_forecasted <- outcome_estimate_forecast_mortality_with_LC$mx_LC_forecasted

  mx_LC_forecasted_low80 <-
outcome_estimate_forecast_mortality_with_LC$mx_LC_forecasted_low80

  mx_LC_forecasted_up80 <-
outcome_estimate_forecast_mortality_with_LC$mx_LC_forecasted_up80

  mx_LC_forecasted_joyAdjusted <-
outcome_estimate_forecast_mortality_with_LC$mx_LC_forecasted_joyAdjusted

  mx_LC_forecasted_low80_joyAdjusted <-
outcome_estimate_forecast_mortality_with_LC$mx_LC_forecasted_low80_joyAdjusted

  mx_LC_forecasted_up80_joyAdjusted <-
outcome_estimate_forecast_mortality_with_LC$mx_LC_forecasted_up80_joyAdjusted

  for(year in (base_end+1):(base_end+years_ahead)){

    e0_LC_forecast[1,as.character(year)] <-
lexp.age.specific(exp(mx_LC_forecasted[,as.character(year)]),lt_ax)[1]

    e0_LC_forecast_low80[1,as.character(year)] <-
lexp.age.specific(exp(mx_LC_forecasted_low80[,as.character(year)]),lt_ax)[1]

    e0_LC_forecast_up80[1,as.character(year)] <-
lexp.age.specific(exp(mx_LC_forecasted_up80[,as.character(year)]),lt_ax)[1]

    e0_LC_forecast_joyAdjusted[1,as.character(year)] <-
lexp.age.specific(exp(mx_LC_forecasted_joyAdjusted[,as.character(year)]),lt_ax)[1]

    e0_LC_forecast_low80_joyAdjusted[1,as.character(year)] <-
lexp.age.specific(exp(mx_LC_forecasted_low80_joyAdjusted[,as.character(year)]),lt_ax)[1]

    e0_LC_forecast_up80_joyAdjusted[1,as.character(year)] <-
lexp.age.specific(exp(mx_LC_forecasted_up80_joyAdjusted[,as.character(year)]),lt_ax)[1]

  }

  outcome_to_return <- list()

  outcome_to_return$e0_observed <- e0_observed

  outcome_to_return$e0_LC_estimated <- e0_LC_estimated

```

```

outcome_to_return$e0_LC_forecast <- e0_LC_forecast

outcome_to_return$e0_LC_forecast_low80 <- e0_LC_forecast_low80

outcome_to_return$e0_LC_forecast_up80 <- e0_LC_forecast_up80

outcome_to_return$e0_LC_forecast_joyAdjusted <- e0_LC_forecast_joyAdjusted

outcome_to_return$e0_LC_forecast_low80_joyAdjusted <- e0_LC_forecast_low80_joyAdjusted

outcome_to_return$e0_LC_forecast_up80_joyAdjusted <- e0_LC_forecast_up80_joyAdjusted

return(outcome_to_return)
}

```

6. Apply functions (steps 3 through 5) to estimate and forecast mortality and life expectancy at birth

We now apply the functions *estimate_forecast_mortality_with_LC* and *estimate_forecast_e0_with_LC* to forecast US female mortality from 2018 to 2067, based on data (1) 1933 to 2017 and (2) 1988 to 2017.

```

mx_lc_1933_2017_2067 <-
estimate_forecast_mortality_with_LC(obs_mortality=usa_wom[[2]][,as.character(1933:2017)],

                                   base_start=1933,

                                   base_end=2017,

                                   number_sim=1000,

                                   years_ahead=50)

e0_lc_1934_2017_2067 <-
estimate_forecast_e0_with_LC(outcome_estimate_forecast_mortality_with_LC=mx_lc_1933_2017_2067)

e0_lc_1934_2017_2067$e0_LC_forecast_joyAdjusted

```

	2018	2019	2020	2021	2022	2023	2024	2025
## [1,]	81.40823	81.55547	81.70143	81.84613	81.98958	82.1318	82.27279	82.41256
	2026	2027	2028	2029	2030	2031	2032	2033
## [1,]	82.55114	82.68852	82.82473	82.95977	83.09366	83.2264	83.35801	83.4885
	2034	2035	2036	2037	2038	2039	2040	2041
## [1,]	83.61787	83.74615	83.87333	83.99943	84.12445	84.24842	84.37133	84.4932
	2042	2043	2044	2045	2046	2047	2048	2049
## [1,]	84.61404	84.73385	84.85264	84.97043	85.08722	85.20302	85.31783	85.43168
	2050	2051	2052	2053	2054	2055	2056	2057
## [1,]	85.54456	85.65648	85.76746	85.87749	85.98659	86.09477	86.20203	86.30838
	2058	2059	2060	2061	2062	2063	2064	2065
## [1,]	86.41383	86.51838	86.62205	86.72483	86.82675	86.92779	87.02798	87.12732
	2066	2067						
## [1,]	87.22581	87.32346						

```

mx_lc_1988_2017_2067 <-
estimate_forecast_mortality_with_LC(obs_mortality=usa_wom[[2]][,as.character(1988:2017)],

                                base_start=1988,

                                base_end=2017,

                                number_sim=1000,

                                years_ahead=50)

e0_lc_1989_2017_2067 <-
estimate_forecast_e0_with_LC(outcome_estimate_forecast_mortality_with_LC=mx_lc_1988_2017_2067)

e0_lc_1989_2017_2067$e0_LC_forecast_joyAdjusted

##           2018      2019      2020      2021      2022      2023      2024      2025
## [1,] 81.43092 81.52036 81.60937 81.69795 81.78611 81.87385 81.96116 82.04806
##           2026      2027      2028      2029      2030      2031      2032      2033
## [1,] 82.13454 82.2206 82.30624 82.39147 82.47629 82.5607 82.6447 82.72829
##           2034      2035      2036      2037      2038      2039      2040      2041
## [1,] 82.81148 82.89426 82.97663 83.05861 83.14018 83.22136 83.30213 83.38251
##           2042      2043      2044      2045      2046      2047      2048      2049
## [1,] 83.46249 83.54208 83.62128 83.70008 83.7785 83.85653 83.93417 84.01142
##           2050      2051      2052      2053      2054      2055      2056      2057
## [1,] 84.08829 84.16478 84.24088 84.31661 84.39195 84.46692 84.54151 84.61573
##           2058      2059      2060      2061      2062      2063      2064      2065
## [1,] 84.68957 84.76304 84.83614 84.90887 84.98123 85.05322 85.12485 85.19612
##           2066      2067
## [1,] 85.26702 85.33756

```

7. Validate forecast performance of Lee-Carter method in the context of US female mortality.

We would like to make as many validation forecasts as possible, taking base period and forecast horizons that comprise 30 years each, so that we have jump-off years 1962 through 1987.

```

e0_1934_2017 <- e0_lc_1934_2017_2067$e0_observed

validate_e0_observed <- matrix(NA,nr=length(1933:1958),nc=30)

rownames(validate_e0_observed) <- 1933:1958

validate_e0_LC_forecast_joyAdjusted <- matrix(NA,nr=length(1933:1958),nc=30)

rownames(validate_e0_LC_forecast_joyAdjusted) <- 1933:1958

for(base_start in 1933:1958){

  validate_e0_observed[as.character(base_start),] <-
e0_1934_2017[1,(as.character((base_start+30):(base_start+30+29)))]

  current_mx_LC_estimate_forecast <-

```

```

estimate_forecast_mortality_with_LC(obs_mortality=
usa_wom[[2]][,as.character(base_start:(base_start+29))],

                                base_start=base_start,

                                base_end=(base_start+29),

                                number_sim=1000,

                                years_ahead=30)

current_LC_e0_forecast <-
estimate_forecast_e0_with_LC(outcome_estimate_forecast_mortality_with_LC=
current_mx_LC_estimate_forecast)

validate_e0_LC_forecast_joyAdjusted[as.character(base_start),] <-
current_LC_e0_forecast$e0_LC_forecast_joyAdjusted
}

error <- validate_e0_LC_forecast_joyAdjusted - validate_e0_observed

error[1:5,1:5]

```

```

##           [,1]      [,2]      [,3]      [,4]      [,5]
## 1933  0.133400683  0.086913290  0.22341276  0.44613568  0.3266086
## 1934 -0.555391640 -0.408271618 -0.17512998 -0.28443764  0.2218182
## 1935  0.003139586  0.239523880  0.13345724  0.64295061  0.5835606
## 1936  0.038919080 -0.051653315  0.47297610  0.42837326  0.4385689
## 1937 -0.525232513 -0.009997288 -0.06359627 -0.06201831 -0.1378467

```

We finally plot the forecast errors by jump-off year

```

colfunc <- colorRampPalette(c("black","blue","red"))

blackblue <- colfunc(length(1:30))

par(fig = c(0,1,0,1), las=1, mai=c(0.8,0.8,0.8,0.8))

plot(x=-100,y=-100,xlim=c(0,30),ylim=c(-2,3),xlab="Forecast horizon",ylab="",
main="Forecast error",axes=FALSE)

axis(side=1,at=seq(0,30,1),labels=FALSE,lwd=1,pos=-2)

axis(side=1,at=seq(0,30,5),labels=seq(0,30,5),lwd=3,pos=-2)

axis(side=2,at=seq(-2,3,0.5),labels=seq(-2,3,0.5),lwd=1,pos=0)

axis(side=2,at=seq(-2,3,1),labels=seq(-2,3,1),lwd=3,pos=0)

for(base_start in 1933:1958){

  lines(x=1:30,y=error[as.character(base_start),],col=blackblue[base_start-1932])

  points(x=1:30,y=error[as.character(base_start),],col=blackblue[base_start-1932],

```

```

lwd=2)
}

segments(x0=1,x1=30,y0=0,y1=0,lwd=2)

points(x=rep(2.5,30),y=seq(1.5,2.5,length=30),col=blackblue[1:30],lwd=2)

text(x=4.5,y=2.6,"Jump-off year:",pos=3)

text(x=4.5,y=1.5,"1962",pos=4)

text(x=4.5,y=2.5,"1987",pos=4)

```

