## Appendix - Matlab code

The perturbation analysis results here are obtained using matrix operations, and optimally computed using software designed for matrix calculations. The best such software is MATLAB, which is a nearly universal standard in physics, engineering, and mathematics. The widely used statistical package R can also perform the calculations, although it is not primarily a matrix language. In this section we provide MATLAB code, and in the next we provide the corresponding R code, for the calculations.

## Function needed in your working directory, titled vec.m

```
function v=vec(x)
v=x(:);
```

## Main code

```
% NEED TO HAVE FILE 'vec.m' in working directory
% INPUT DATA are "q" a vector of death probabilities
% (we used ages 0 to 110+ from HMD),
% and "x", a vector of the average age at death within each age interval,
% (i.e. \{a0, 1.5, 2.5, \ldots, 110.5, 110 + a(omega)\} - total length = q+1)
% Preliminaries
% defining survival probabilities
   p=1-q;
% # of transient states + 1, because in construction of U first row is zero
   s=length(p)+1;
% # transient states only
   s2=length(p);
% other things we will need later
   I = speye(s); % identity matrix
   e = ones(s,1); % column vector of ones for summations
   e1 = [1, linspace(0, 0, s2)]';
% C matrix is for calculating cumulative sums
   for i=1:s
       C(:,i) = [zeros(i,1); ones(s-i,1)];
   end
```

```
% Markov chain formulation of longevity
% U matrix describes transient states in the Markov chain
   U=sparse(diag(p,-1));
% N matrix, where (i,j) entry is the mean time spent in each age class i,
% conditional upon starting in age class j
   N=inv(I-U);
% M matrix has probability of death at each age on diagonal
   M=sparse(diag(1-p));
   M(s, s) = 1;
% The distribution of ages at death
   B=M*N; % the complete distribution of ages at death
   f = B*e1; % the distribution of ages at death from birth (or first age class)
% survivorship (alternatively ell=N*e1)
   ell = e - C * f;
% remaining life expectancy at each age
   mean_eta = sum(N)' - 0.5;
% life expectancy at birth (or first age class)
   eta = e'*N*e1 - 0.5;
% NB: in Markov chain formulations, the life expectancy at birth is always
% 0.5 years higher than that found by conventional life table methods,
% which is why we subtract 0.5 years
% Indices of lifespan variation
% variance in lifespan
   V=(sum(N)*(2*N-I) - mean_eta'.*mean_eta')';
% standard deviation in lifespan
   S=sqrt(V);
% e measure
   dagger_eta=mean_eta'*B;
% Theil's index
   T = f'*((x*eta^(-1)).*(log(x*eta^(-1))));
% Mean Log Deviation
   MLD = f' * (e*log(eta) - log(x));
```

```
% Gini coefficient
    G=1-(1/eta) *e'*(ell.*ell);
% IOR calculations
   F=[cumsum(f)]; % cumulative deaths
    age=[0:s-1]';
    xhat1=interp1(F(1:s-10),age(1:s-10),0.25);
% the 's-10' is in the code to handle zero deaths at oldest ages
    xhat2=interp1(F(1:s-10),age(1:s-10),0.75);
    IQR=xhat2-xhat1;
% SENSITIVITIES
% derivatives of U with respect to mortality change
   for i=1:s-1
       dU_dmu=sparse(zeros(s,s));
       dU_dmu(i+1,i) = -p(i);
       dvecU_dmu(:,i) = vec(dU_dmu);
    end
%derivatives of M with respect to mortality change
    for i=1:s-1
       dM_dmu=sparse(zeros(s,s));
       dM_dmu(i,i) = p(i);
       dvecM_dmu(:,i) = vec(dM_dmu);
    end
%derivative of f with respect to mortality change %should it be B or N
    df_dmu=kron(e1'*N',I)*dvecM_dmu + kron(e1'*N',B)*dvecU_dmu;
*sensitivity of expected longevity with respect to mortality change
    deta_dmu=kron(e1'*N',e'*N)*dvecU_dmu;
%sensitivity of variance in longevity with respect to mortality change
    dV_dmu = (2*kron((N')^2, mean_eta')...
           + 2*kron(N', mean_eta'*N)...
           - (I + 2*sparse(diag(mean_eta)))*kron(N', mean_eta'))...
           * dvecU_dmu;
*sensitivity of standard deviation with respect to mortality change
    dS_dmu=0.5*diag(1./S)*dV_dmu;
%sensitivity of eta-dagger with respect to mortality change
    deta_dagger_dmu = (kron(B(:,1)'*N', mean_eta') ...
                   + kron(N(:,1)',mean_eta'*B))*dvecU_dmu ...
                   + kron(N(:,1)',mean_eta')*dvecM_dmu;
```

```
%sensitivity of Theil's with respect to mortality change
    dT_dmu = (x'*eta^(-1)).*log(x'*eta^(-1))*df_dmu...
            - (T*eta^(-1)+f'*x*eta^(-2))*deta_dmu;
%sensitivity of MLD with respect to mortality change
    dMLD_dmu = (e'*log(eta)-log(x'))*df_dmu+eta^(-1)*deta_dmu;
%sensitivity of Gini with respect to mortality change
    dG_dmu = eta^(-2) *e' * (ell.*ell) *deta_dmu + 2/eta *e' *diag(ell) *C*df_dmu;
% sensitivity of IQR with respect to mortality change
    % sensitivity of cumulative deaths with respect to mortality change
        dF_dmu=[zeros(1,s2);cumsum(df_dmu)];
    % the death density at the quantiles by interpolation
        fxhat1=interp1(age, f, xhat1);
        fxhat2=interp1(age, f, xhat2);
        dFxhat1_dmuT=interp1([0:s]',dF_dmu(:,:),xhat1);
        dFxhat2_dmuT=interp1([0:s]',dF_dmu(:,:),xhat2);
    % finally taking sensitivities
        dxhat_dmu1 = -fxhat1^(-1)*dFxhat1_dmuT;
        dxhat_dmu2 = -fxhat2^(-1)*dFxhat2_dmuT;
        dIOR_dmu=dxhat_dmu2-dxhat_dmu1;
88888888888888888888888888
%ELASTICITIES
응응응응응응응응응응응응응응응응응응응응
% elasticity of standard deviation
   ES = (1./S(1))*dS_dmu(1,:)*diag(1-p);
% elasticity of variance
   EV = (1./V(1)) * dV_dmu(1,:) * diag(1-p);
% elasticity of edagger
    Eedag = (1./dagger_eta(1))'*deta_dagger_dmu*diag(1-p);
% elasticity of Gini
   EG = (1./G(1)) * dG_dmu * diag(1-p);
% elasticity of Theil's
   ET = (1./T(1))*dT_dmu*diag(1-p);
% elasticity of MLD
   EMLD = (1./MLD(1))*dMLD_dmu*diag(1-p);
% elasticity of IQR
   EIQR = (1./IQR(1))*dIQR_dmu*diag(1-p);
%SAVING THE RESULTS
$$$$$$$$$$$$$$$$$$$$$$$$$$
    res = [stV stedag stS stT stMLD stIQR stG EV Eedag ES ET EMLD EIQR EG];
```

res=full(res); % MATLAB can't save sparse arrays
save res.txt res -ascii;

## Appendix - R code

```
# R code for paper "Perturbation analysis of indices of lifespan variability"
# This code was designed for readability, not efficiency
rm(list = ls(all = TRUE))
options(scipen=6)
#-----
#---- Input data
#-----
#-- "q" a vector of death probabilities (we used ages 0 to 110+ from HMD),
#-- and "x", a vector of the average age at death within each age interval
#-- (i.e. a0, 1.5, 2.5 etc., length qi+1)
#-----
#---- Packages and preliminaries
#-----
 library(Matrix) # for sparse matrices
 library(signal) # for interp1 function
#-----
#--- special functions
#-----
#--- function for making subdiagonals (from Bill Venables)
 subdiag <- function (v, k) {</pre>
   n <- length(v) + abs(k)
   x \leftarrow matrix(0, n, n)
   if (k == 0)
      diag(x) \leftarrow v
   else if (k < 0)
     { ## sub-diagonal
      j <- 1:(n+k)
       i \leftarrow (1 - k):n
      x[cbind(i, j)] \leftarrow v
     }
```

```
#-----
# Preliminaries
#-----
# defining survival probabilities
 p <- 1-q
# number of transient states + 1, (in construction of U first row is zero)
 s \leftarrow length(p)+1
# number of transient states only
 s2 <- length(p)
# other things we will need later
 I <- diag(rep(1,s)) # identity matrix</pre>
 I <- as(I, "sparseMatrix")</pre>
 e <- rep(1,s) # vector of ones for summations
 e1 <- c(1,rep(0,s-1))
 age <- 0:s2
# C matrix is for calculating cumulative sums
 C <- Matrix(0,nrow=s,ncol=s)</pre>
   for (i in 1:s){
     C[,i] \leftarrow c(rep(0,i),rep(1,s-i))
 C <- as(C, "sparseMatrix")</pre>
#-----
# Markov chain formulation of longevity
#-----
# U matrix describes transient states in the Markov chain
 U \leftarrow subdiag(p,-1)
 U <- as(U, "sparseMatrix")</pre>
# N matrix, where (i,j) entry is the mean time spent in each age class i,
# conditional upon starting in age class j
 N <- solve(I-U)
```

x }

```
N <- as(N, "sparseMatrix")</pre>
# M matrix has probability of death at each age on diagonal
 M \leftarrow diag(c(q,1))
 M <- as(M, "sparseMatrix")</pre>
# The distribution of ages at death
  B <- M \%*\% N \, # the complete distribution of ages at death
  f <- B %*\% e1 # the distribution of ages at death from birth (1st age class)
  B <- as(B, "sparseMatrix")</pre>
# survivorship (alternatively ell <- e-C%*%f)</pre>
  ell <- N %*% e1
  ell <- as(ell, "sparseMatrix")</pre>
# remaining life expectancy at each age
  mean_eta <- colSums(N)-0.5</pre>
# life expectancy at birth (or first age class)
  eta <- mean_eta[1]</pre>
# NB: in Markov chain formulations, the life expectancy at birth is always
# 0.5 years higher than that found by conventional life table methods,
# which is why we subtract 0.5 years
# Indices of lifespan variation
#-----
# variance in lifespan
  V <- colSums(N) %*% (2*N-I) - mean_eta*mean_eta
# standard deviation in lifespan
  S <- sqrt(V)
# life disparity (e-dagger)
  eta_dagger <- mean_eta %*% B
# Theil's index
  T \leftarrow sum(f*x*eta^(-1)*log(x*eta^(-1)))
```

```
# Mean Log Deviation
  MLD \leftarrow sum(f*(e*log(eta)-log(x)))
# Gini coefficient
  G \leftarrow 1-(1/eta) * e %*% (ell*ell)
# Interquartile range
  F <- (cumsum(f))
  xhat1 \leftarrow interp1(F[1:(s-10)],age[1:(s-10)],0.25)
  xhat2 \leftarrow interp1(F[1:(s-10)], age[1:(s-10)], 0.75)
  IQR <- xhat2-xhat1</pre>
  # the "s-10" is in the code to handle zero deaths at oldest ages
#-----
# SENSITIVITIES
#-----
# derivatives of U with respect to mortality change
   dvecU_dmu <- Matrix(0,nrow=s*s,ncol=s-1)</pre>
   r <- seq(2,s*s,s+1) # rows that will contain the elements once stacked
     for (i in 1:(s-1)){
      dvecU_dmu[r[i],i] <- -p[i]</pre>
   dvecU_dmu <- as(dvecU_dmu, "sparseMatrix")</pre>
# derivatives of M with respect to mortality change
   dvecM_dmu <- Matrix(0,nrow=s*s,ncol=s-1)</pre>
   r2 <- seq(1,s*s,s+1) # rows that will contain the elements once stacked
    for (i in 1:(s-1)){
     dvecM_dmu[r2[i],i] <- p[i]</pre>
   dvecM_dmu <- as(dvecM_dmu, "sparseMatrix")</pre>
# derivative of f with respect to mortality change
   df_dmu <- t(kronecker(N[,1],I)) %*% dvecM_dmu +</pre>
     t(kronecker(N[,1],t(B))) %*% dvecU_dmu
# sensitivity of expected longevity with respect to mortality change
   deta_dmu <- t(kronecker(N[,1],colSums(N))) %*% dvecU_dmu</pre>
# sensitivity of variance in longevity with respect to mortality change
```

```
dV_dmu \leftarrow (2*kronecker(t(N) %*% t(N),t(mean_eta)) +
     2*kronecker(t(N),mean_eta %*% N) -
     (I + 2*(diag(mean_eta))) %*% kronecker(t(N),t(mean_eta))) %*% dvecU_dmu
# sensitivity of standard deviation with respect to mortality change
   dS_dmu <- 0.5*diag(as.vector(1/S)) %*% dV_dmu
# sensitivity of e-dagger with respect to mortality change
   deta_dagger_dmu <- (kronecker(t(B[,1]) %*% t(N),t(mean_eta)) +</pre>
      kronecker(t(N[,1]),(t(mean_eta)%*%B))) %*% dvecU_dmu +
      kronecker(t(N[,1]),t(mean_eta)) %*% dvecM_dmu
# sensitivity of Theil's with respect to mortality change
    dT_dmu \leftarrow (t(x) * eta^(-1) * log(t(x) * eta^(-1))) %*% df_dmu -
      (T*eta^{-1}) + t(f) %*% x * eta^{-2}) %*% deta_dmu
# sensitivity of MLD with respect to mortality change
    dMLD_dmu \leftarrow (t(e) * log(eta) - log(t(x))) %*% df_dmu + eta^(-1) %*% deta_dmu
# sensitivity of Gini with respect to mortality change
    dG_dmu <- eta^(-2) * t(e) %*% (ell*ell) %*% deta_dmu +
      2/eta * t(e) %*% diag(as.vector(ell)) %*% C %*% df_dmu
# sensitivity of IQR with respect to mortality change
    # sensitivity of cumulative deaths with respect to mortality change
        dF_dmu <- matrix(0,nrow=s,ncol=s2)</pre>
          for(i in 1:s2){
            dF_dmu[i:s2,i] <- cumsum(df_dmu[i:s2,i])</pre>
        dF_dmu <- rbind(rep(0,s2),dF_dmu)</pre>
    # the death density at the quantiles by interpolation
        fxhat1 <- interp1(age,f,xhat1)</pre>
        fxhat2 <- interp1(age,f,xhat2)</pre>
        dFxhat1_dmuT <- rep(0,s2)
        dFxhat2_dmuT <- rep(0,s2)
          for (i in 1:s2){
            dFxhat1_dmuT[i] <- interp1(0:s,dF_dmu[,i],xhat1)</pre>
            dFxhat2_dmuT[i] <- interp1(0:s,dF_dmu[,i],xhat2)</pre>
            }
```

```
# finally taking sensitivities
        dxhat_dmu1 <- -fxhat1^(-1)*dFxhat1_dmuT</pre>
        dxhat_dmu2 <- -fxhat2^(-1)*dFxhat2_dmuT</pre>
    dIQR_dmu <- dxhat_dmu2 - dxhat_dmu1</pre>
#-----
# ELASTICITIES
#-----
# elasticity of standard deviation
    ES <- 1/S[1] * dS_dmu[1,] %*% diag(1-p)
# elasticity of variance
    EV <- 1/V[1] * dV_dmu[1,] %*% diag(1-p)
# elasticity of edagger
    EEDAG <- 1/eta_dagger[1] * deta_dagger_dmu %*% diag(1-p)</pre>
# elasticity of Gini
    EG \leftarrow 1/G[1] * dG_dmu %*% diag(1-p)
# elasticity of Theil's
    ET <- 1/T[1] * dT_dmu %*% diag(1-p)
# elasticity of MLD
    EMLD <- 1/MLD[1] * dMLD_dmu %*% diag(1-p)</pre>
# elasticity of IQR
    EIQR <- 1/IQR[1] * dIQR_dmu %*% diag(1-p)</pre>
# PLOTTING THE ELASTICITIES
#-----
  ELAS <- cbind(EEDAG[1,],EG[1,],EIQR[1,],EMLD[1,],ES[1,],ET[1,],EV[1,])
  indices <- c("edag", "G", "IQR", "MLD", "S", "T", "V")</pre>
  names(ELAS) <- indices</pre>
  matplot(ELAS,xlab="Age",ylab="Elasticity of index",
```

```
#ylim=c(-0.05,0.1),
t="l",lty=1,lwd=2,col=1:7)
legend(legend=indices,"topright",col=1:7,lty=1,lwd=2)
abline(h=0,lty="dotted")
```