Demography of Diabetes in Denmark or: How to put real probabilities in your transition matrix and use them

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Demography of diabetes in DK

▶ How does diabetes spread in the population?

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- How does diabetes spread in the population?
- ▶ Life time risk of DM

Demography of diabetes in DK

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- Life time risk of DM
- ...and complications

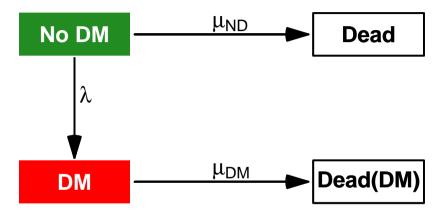
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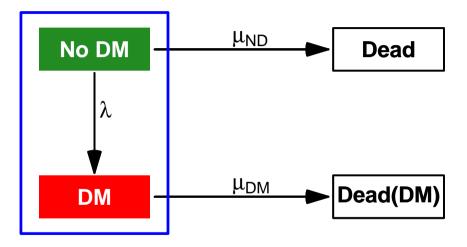
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- Prevalence of diabetes has been increasing, while
- ► Incidence rates have been **in**creasing (4% / year)
- Mortality rates have been decreasing (2% / year)
- ▶ What is the relative contribution of each?

Demographic scenario



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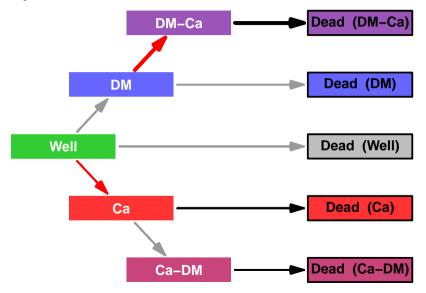
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 - ▶ Impact of the DM vs noDM cancer incidence RR

Demographic scenario



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- Age-specific transition rates
- ...as continuous functions of age
- ...and possibly other time scales

Transition rates between states as function of a and p:

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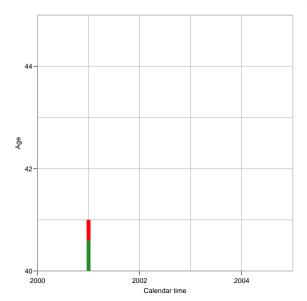
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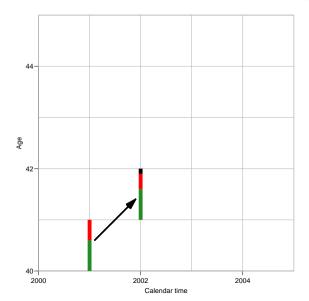
$$P_{\mathsf{ND},\mathsf{ND}}(\ell) = \exp(-(\lambda + \mu_{\mathsf{ND}})\ell)$$

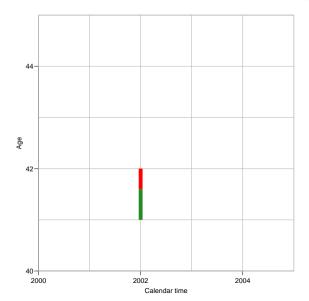
$$P_{\mathsf{ND},\mathsf{Dead}}(\ell) = \frac{\mu_{\mathsf{ND}}}{\lambda + \mu_{\mathsf{ND}}} \Big(1 - \exp(-(\lambda + \mu_{\mathsf{ND}})\ell) \Big)$$

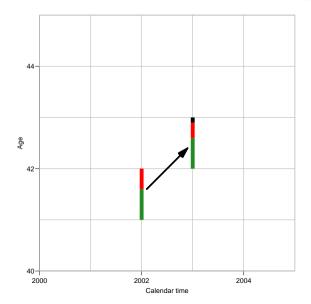
$$P_{\mathsf{ND},\mathsf{DM}}(\ell) = \frac{\lambda}{\lambda + \mu_{\mathsf{ND}}} \Big(1 - \exp(-(\lambda + \mu_{\mathsf{ND}})\ell) \Big)$$

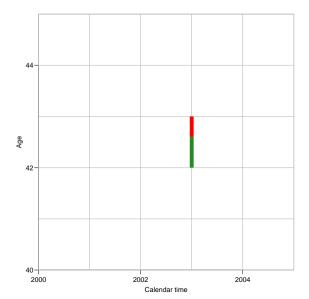
$$P_{\rm DM,Dead}(\ell) = 1 - \exp(-\mu_{\rm DM}\ell)$$

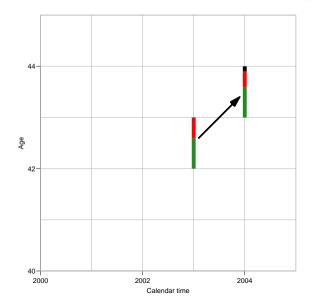


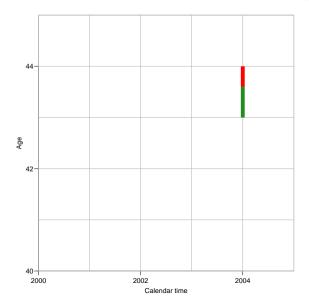




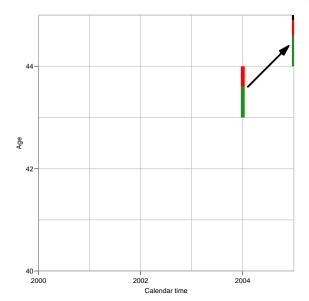




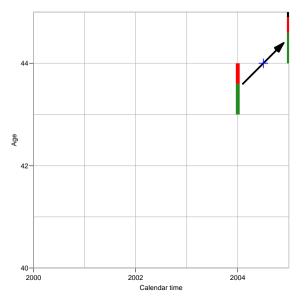




Prevalence of DM — updating



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But where do we get the rates from?

▶ National Diabetes Register, 1995–2011

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Example: state No DM

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► Classification of follow-up (time and events) by age (0-100), calendar time (1995-2011) and date of birth (1-year classes) (Lexis triangles)

- ▶ Time at risk:
 - from date of birth or start of study
 - ▶ to date of DM or Dead or Ca (or end of study)
- Events (transitions)
 - DM
 - Dead
 - ▶ Ca
- ► Classification of follow-up (time and events) by age (0-100), calendar time (1995-2011) and date of birth (1-year classes) (Lexis triangles)
- Similary for the study with cancer states



▶ Incident cases / deaths from each state

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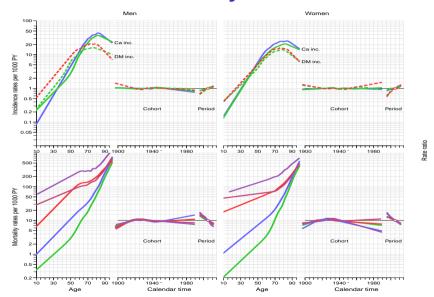
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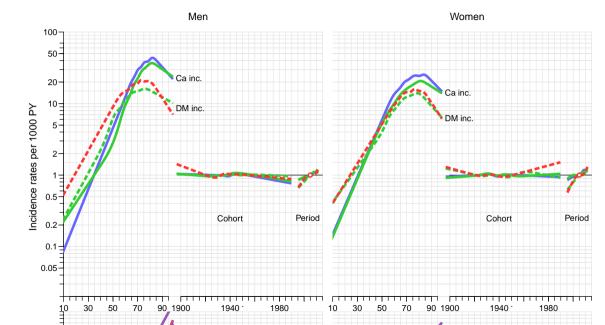
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- Note: Only use the predictions from the models

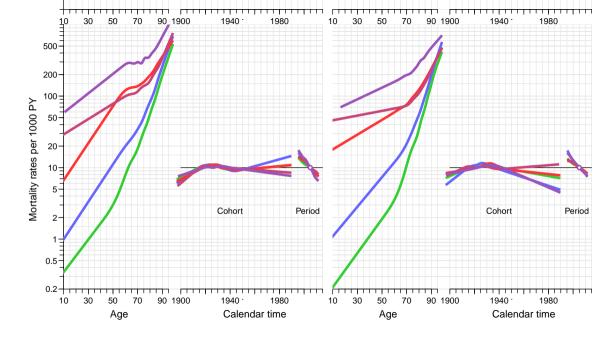
Events and risk time

```
> cbind(
+ xtabs(cbind(D.ca, D.dm, D.dd) ~ state, data=dcd), round(
+ xtabs( Y/1000 ~ state, data=dcd ), 1 ) )
        D.ca D.dm D.dd
      447419 345400 628705 87502.9
Well
DM
       35145
                   0 73480
                            2031.3
                            89.1
DM-Ca
                      24153
                                              DM-Ca
                                                           Dead (DM-Ca)
Ca
             23508 222966 1973.6
                      14703
                            117.0
Ca-DM
                                                            Dead (DM)
                                          DM
Dead
                          0
                              0.0
                                                            Dead (Well)
                                     Well
                                           Ca
                                                            Dead (Ca)
                                                          Dead (Ca-DM)
                                              Ca-DM
```









Transition rates

```
> int <- 1/12
> a.pt <- seq(int,102,int) - int/2
> system.time(
+ for( vv in dimnames(PR)[[4]])
+ {
+ nd <- data.frame( A=a.pt, P=as.numeric(yy), Y=int )
+ PR["Well" ,"DM" ,,yy,"M"] <- ci.pred( M.w2dm$model , newdata=nd )[,1]
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```

Transition matrices

Use the rates to generate the transition **probabilities**:

```
> print.table( round( addmargins( ci2pr( PR[,.800,1,1] )*10^4,
                                    margin=2)).
                zero.print="." )
       to
from
         Well
                  DM DM-Ca
                               Ca Ca-DM
                                           D-W
                                                D-DM
                                                             D-DC
                                                                           Siim
  Well
         9963
                   8
                               12
                                            17
                                                                         10000
  DM
               9943
                        16
                                                  40
                                                                         10000
                                                              422
                                                                         10000
  DM-Ca
                      9578
                             9815
                                                        175
                                                                         10000
  Ca
  Ca-DM
                                   9865
                                                                     135 10000
  D-W
                                       . 10000
                                                                         10000
                                               10000
  D-DM
                                                                         10000
  D-Ca
                                                     10000
                                                                         10000
  D-DC
                                                            10000
                                                                         10000
  D-CD
                                                                  10000 10000
```

State occupancy probabilites

```
> PV <- PR[1,,,,]*0
> for( sc in dimnames(PRp)[["per"]] )
+ for( sx in dimnames(PRp)[["sex"]] )
+ {
+ # Initialize to all well at age 0:
+ PV[,1,sc,sx] <- c(1,rep(0,9))
+ # Compute distribution at endpoint of each age-interval
+ for( ag in 1:dim(PRp)[3] ) PV[,ag,sc,sx] <- PV[ ,max(ag-1,1),sc,sx] %*%
+ PRp[,, ag ,sc,sx]
+ PRp[,, ag ,sc,sx]</pre>
```

▶ Start all in age 0 in state "Well"

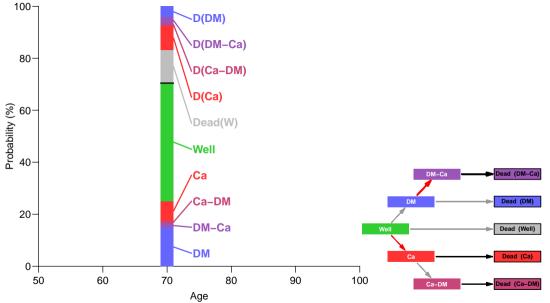
- ▶ Start all in age 0 in state "Well"
- ▶ Use rates to predict how many transfer to "DM", "Ca", "Dead" during a small interval

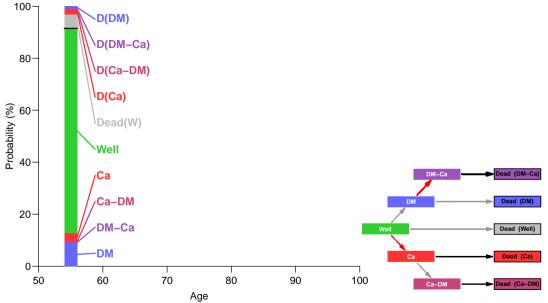
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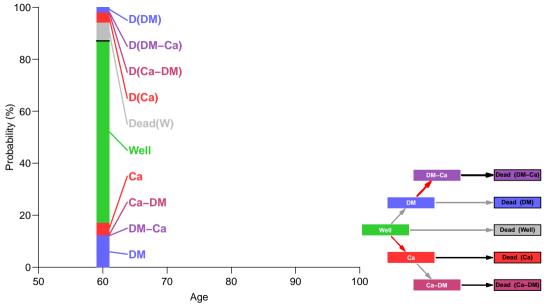
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- Interval length: 1 month

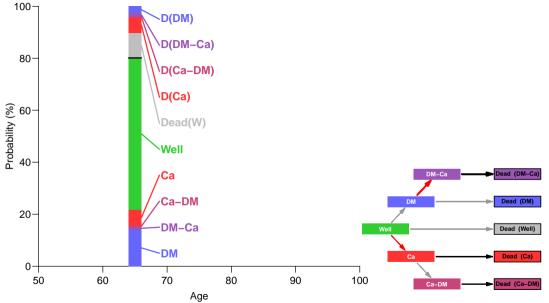
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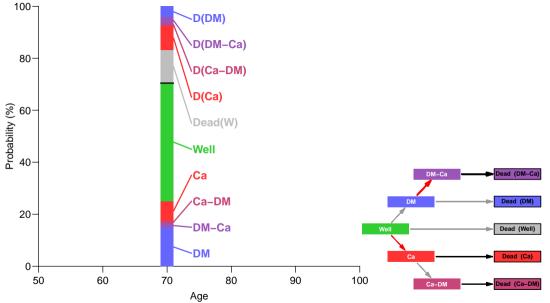
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- Different scenarios using estimated (cross-sectional) rates at 1 January 1995, 1996, ..., 2012

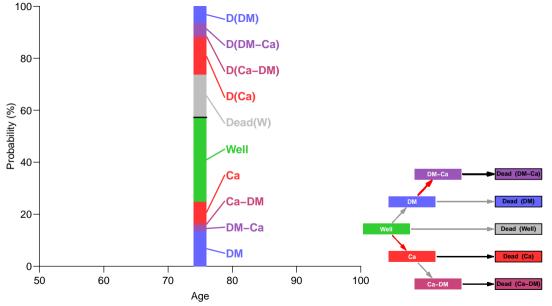


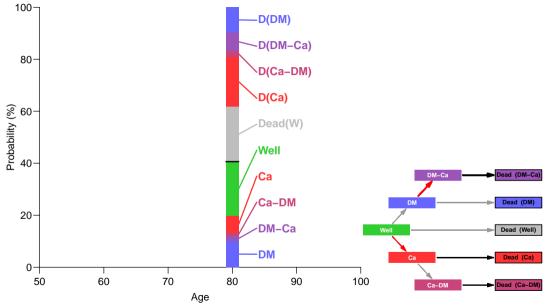


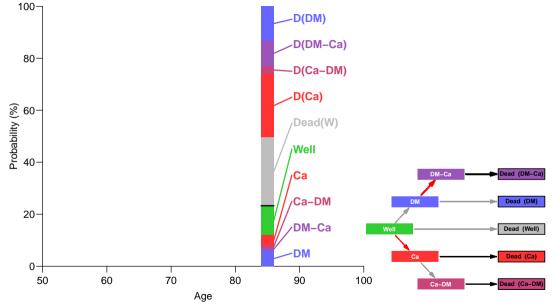


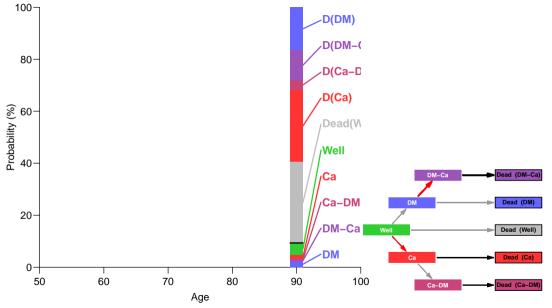


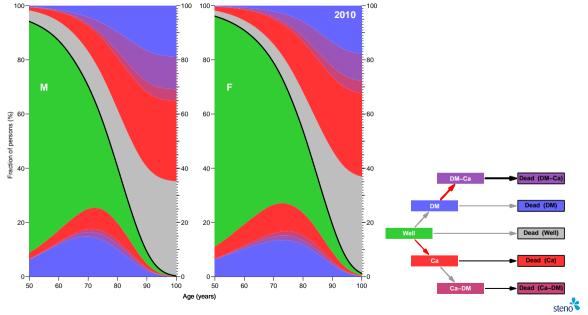


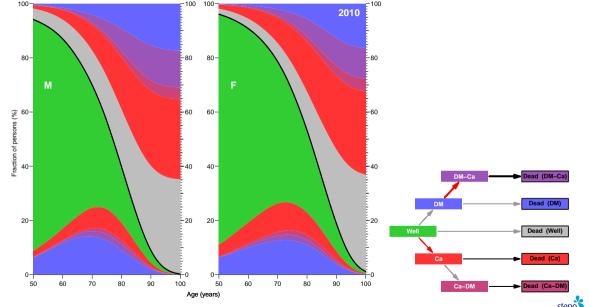




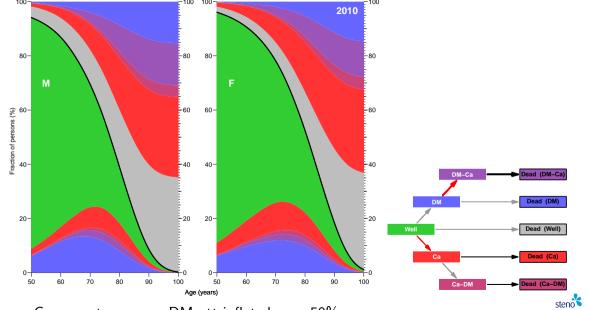








Cancer rates among DM-ptt inflated 20%



Cancer rates among DM-ptt inflated

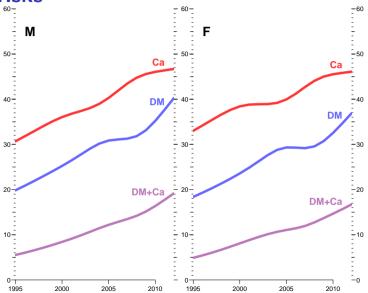
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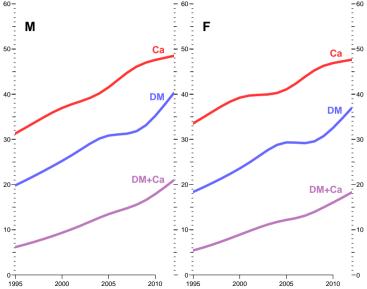
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```

Lifetime risks

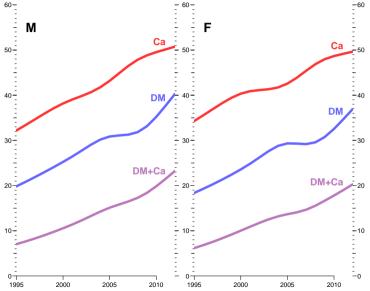




Lifetime risks - RR inflated 20%



Lifetime risks - RR inflated 50%





Demographic changes in DM & Cancer 1995–2012

► Changing **rates** in period 1995–2012:

Diabetes incidence	4% /year
Cancer incidence	2% /year
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Demographic changes in DM & Cancer 1995–2012

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► Changing **life-time risk** 1995–2012:

		+20% Ca DM	+50% Ca DM
Diabetes	19% to 38%	19% to 38%	19% to 38%
Cancer	32% to 46%	33% to 48%	34% to 50%
DM + Ca	6% to 18%	6% to 20%	7% to 22%

Conclusion — DM & Cancer

Increasing incidence rates of DM and Cancer is what matters for (changes in) lifetime risk...

Conclusion — DM & Cancer

- Increasing incidence rates of DM and Cancer is what matters for (changes in) lifetime risk...
- not the (slightly) elevated risk of Cancer among DM paitents.

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- ▶ Use fitted models for incidence and mortality as function of ge and calendar time — to predict prevalences 2012

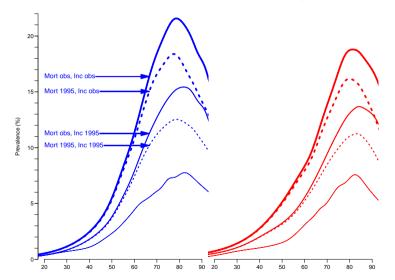
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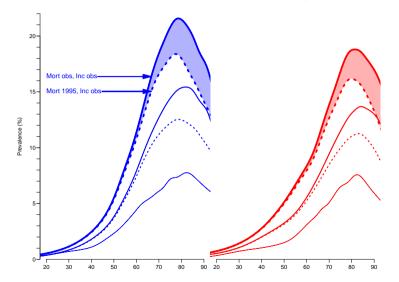
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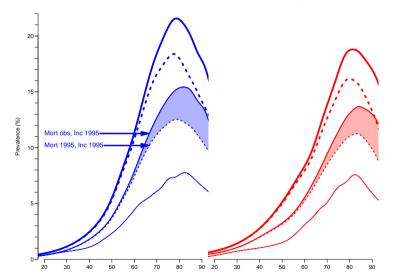
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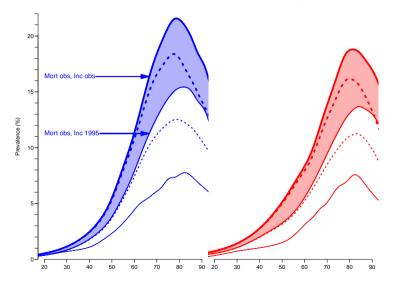
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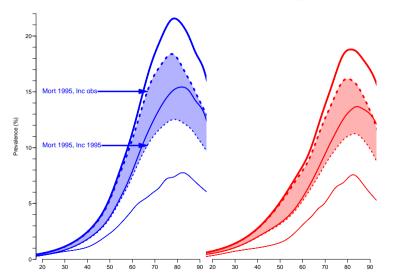
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- Assume:
 - Incidence rates had remained at 1995 level
 - Mortality rates had remained at 1995 level
 - Both had remained at 1995 level
- ▶ Differences between predicted prevalences gives the contribution from incidence rate changes, mortality rate changes and 1995 disequilibrium.



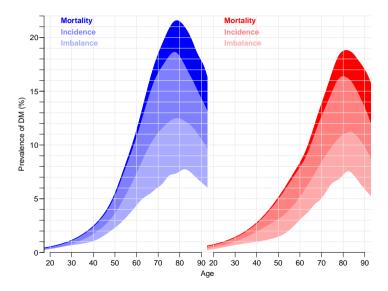




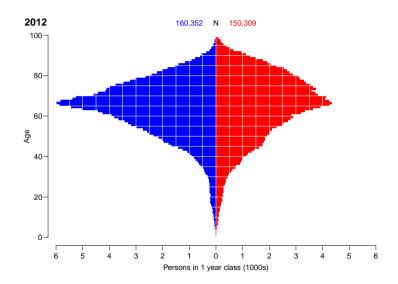




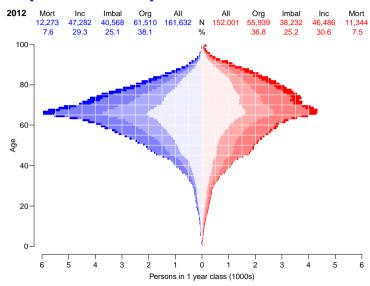
Componets of prevalent cases



Prevalent cases



Components of prevalent cases



Thanks for your attention

