

Appendix A: RIVM Chronic Disease Model

Introduction

The RIVM Chronic Disease Model (CDM) is a state transition Markov-type simulation model that describes how morbidity and mortality for several chronic diseases change over time in the Dutch population as a result of changes in epidemiological risk factors^{1,2}. In general, the state transition model is a suitable and accepted model to describe demographic / epidemiological processes³. Disease experts and modelers have cooperated in building and testing of the CDM and several studies with different applications of the model have already been published⁴⁻¹⁰. The CDM has been formulated as a set of time-continuous differential and is implemented in the software package Mathematica.

Basic structure

In the CDM different states are defined for the risk factor classes (e.g. never smokers, current smokers and former smokers) and states for the diseases included in the model (e.g. stroke: yes or no). In the starting year of the simulation period all persons are distributed over these states. Then, in time steps of 1-year, persons move from one state to another. These transitions are governed by so-called transition rates. E.g., class transition rates between the BMI states 'normal weight' and 'overweight' govern the change of the BMI distribution in the population, incidence rates between the states 'without diabetes' and 'with diabetes' govern the disease prevalence rates, and mortality rates from the state 'alive' to 'deceased' govern the surviving population numbers. For our calculations, we did not take into account transitions between risk factor classes over time. Thus, all cohorts are closed in the sense that no transitions occur between risk factor classes over the life-time. The transition rate is assumed independent from the preceding states and depends only on the present state defined by risk factor class, disease state, sex and age. The disease incidence rates depend on the risk factor class, using relative risk values. E.g., 'overweight' persons have higher diabetes risks than persons with 'normal weight'. For non diseased, the mortality rates depend on risk factor class, e.g., obese persons have higher mortality risks than persons with a normal weight. The mortality rates also depend on the disease states, but are conditional hereon independent on the risk factor values. E.g., the excess mortality risks of people with diabetes compared to people without diabetes are equal for all BMI states.

We assumed that all risk factors that are distinguished are independently distributed, e.g. we assumed that the distribution of smoking independent from BMI. All disease incidence risks

were made dependent on these risk factors by multiplying the baseline risk with relative risk values specified by risk factor class. Moreover, we assumed for some disease pairs independent effects of one disease on the other. E.g., people with diabetes have higher risks of myocardial infarction compared to people without diabetes, independently from overweight and the other risk factors.

The main model outcome variables are incidence, prevalence and mortality numbers, specified by disease, age and gender. For the calculation of lifetime health care costs of the different cohort we used the number of survivors and disease prevalence numbers of the different cohorts.

Input data

Smoking classes distinguished in the CDM are never smokers, current smokers and former smokers. Body weight is modeled in three classes using Body Mass Index (BMI) as indicator: $18.5 < \text{BMI} < 25$ (normal weight), $25 \leq \text{BMI} < 30$ (overweight), $\text{BMI} \geq 30$ (obesity). Table A1 displays the diseases modeled in the CDM that are related to BMI and/or smoking.

Table 1: diseases modeled in the CDM and their relation to smoking and obesity

	Related to smoking	Related to obesity
Cardiovascular disease		
<i>Acute myocardial infarct (AMI)</i>	+	+
<i>Angina pectoris</i>	+	+
<i>Chronic Heart Failure</i>	+	+
<i>Stroke (CVA)</i>	+	+
Cancer		
<i>Lung</i>	+	
<i>Stomach</i>	+	
<i>Oesophagus</i>	+	
<i>Pancreas</i>	+	
<i>Oral cavity</i>	+	
<i>Larynx</i>	+	
<i>Urinary bladder</i>	+	
<i>Kidney</i>	+	+
<i>Rectum</i>		+
<i>Colon</i>		+
<i>Breast</i>		+
<i>Prostate</i>		+
<i>Endometrium</i>		+
Other		
<i>COPD</i>	+	
<i>Diabetes</i>	+	+
<i>Atrhrosis of the hip</i>		+
<i>Arthrosis of the knee</i>		+
<i>Dorsopathies (low back pain)</i>		+

We included those diseases in the model, that are related to obesity according to current scientific consensus¹¹. For the selection of smoking related diseases we followed the report by the Surgeons' General¹². For all diseases related to smoking and/or obesity modeled in the CDM, age and sex specific incidence, prevalence and mortality rates were estimated using a three state transition model¹³⁻¹⁵. The higher mortality risks of patients compared to disease-free persons were calculated from published relative survival proportions for types of cancer, and from incidence and prevalence figures for other chronic diseases, using an Incidence, Prevalence, Mortality (IPM) model¹⁶. Data on risk factor class prevalence were obtained from representative national or regional surveys Risk factor prevalence rates for smoking are based on data of STIVORO¹⁷. For obesity, data from the annual POLS survey from Statistics Netherlands are used¹⁸. Relative risks on morbidity and mortality for smoking and obesity are based on several observational studies¹⁹⁻⁶⁵. Relative risks of the three BMI classes were calculated in three steps. First, a quadratic function was estimated to describe the non-linear relation between BMI and all cause mortality relative risks for different studies. The parameters of these functions were then plotted against age to estimate an age gradient. In a third step, average relative risks for the three different BMI classes were computed using the BMI distribution within these classes in the Netherlands. For the current and former smoking classes distinguished in the CDM, data were used from studies that reported relative risks for all current and/or all former smokers specified by gender and age. To estimate health care costs for the different cohorts data of the Costs of Illness in the Netherlands study were used⁶⁶. In that study the total direct health care costs in the Netherlands of 2003 are uniquely attributed to disease categories specified by gender and age classes. All input data were specified by gender and age (see Appendix B).

Mathematical model equations

The model consists of three parts. In the model initialization part the parameter values and the initial distribution of the population over all model states are calculated. In the model simulation part the 1-year changes of the prevalence numbers are calculated. These changes are the result of transitions between risk factor classes and disease states^a. The transition numbers are computed as the transition rates times the prevalence numbers at the start of the 1-year time-interval. Finally, in the model post-processing part the values of the model output variables such as health care costs and life expectancy, are calculated from the results of the simulation part. The Runge-Kutta method is used to find initial values and numerical approximations for 1 year time steps used in the CDM.

^a Since, in this application, we did not take into account transitions between risk factor classes we will focus exclusively on the transitions between disease states.

Model initialization part

The parameters calculated here are the baseline disease incidence rates, i.e. the incidence rate values for normal weight never smokers, the mortality rates for other causes of death.

Mortality rates from Statistics Netherlands for the year 2004 are attributed to risk factor classes to derive mortality rates specified by risk factor class. Assuming independence between risk factor class prevalence rates and multiplicative relative risks (i.e. no interaction on log-linear scale) we can write mortality rates for the different cohorts as (for notational simplicity, age and sex indices have been omitted in the notation throughout the paper):

$$m(\text{tot} \mid s_j, b_k) = m(\text{tot})_0 * RR(\text{tot} \mid s_j) * RR(\text{tot} \mid b_k) \quad (1)$$

$m(\text{tot} \mid s_j, b_k)$ all cause mortality rate for cohort for smoking class j BMI class k

$m(\text{tot})_0$ baseline all cause mortality rate for 'healthy living' cohort

$RR(\text{tot} \mid s_j)$ relative risk all cause mortality smoking class j

$RR(\text{tot} \mid b_k)$ relative risk all cause mortality BMI class k

Using (1) we can write the baseline mortality rate for the 'healthy living' cohort as:

$$m(\text{tot})_0 = \frac{m(\text{tot})}{\sum_{j,k} RR(\text{tot} \mid s_j) * RR(\text{tot} \mid b_k) * s_j * b_k} \quad (2)$$

$m(\text{tot})$ all cause mortality rate (Statistics Netherlands)

s_j prevalence rate smoking class j

b_k prevalence rate BMI class k

Baseline disease incidence rates and risk factor class specific disease incidence rates are calculated in the same fashion as mortality rates:

$$i(d \mid s_j, b_k) = i(d)_0 * RR(d \mid s_j) * RR(d \mid b_k) \quad (3)$$

$i(d | s_j, b_k)$ incidence rate disease d for cohort smoking class j BMI class k

$i(d)_0$ baseline incidence rate for 'healthy living' cohort

$RR(d | s_j)$ relative risk for disease d smoking class j

$RR(d | b_k)$ relative risk for disease d BMI class k

$$i(d)_0 = \frac{i(d)}{\sum_{j,k} RR(d | s_j) * RR(d | b_k) * s_j * b_k} \quad (4)$$

$i(d)$ population incidence rate disease d

The CDM describes disease prevalence numbers for each disease separately and it is assumed that the disease-specific attributed mortality rates are additive. The all cause mortality rates are the sum of the disease specific attributed mortality rates and the mortality rates from other causes of death:

$$m(oc) = m(tot) - \sum_d am(d) p(d) \quad (5)$$

$m(oc)$ mortality rate for other causes of death

$p(d)$ disease d prevalence rates (several sources)

$am(d)$ mortality rate attributed to disease d

Mortality rates attributed to diseases are calculated by dividing the cause specific mortality rates registered by Statistics Netherlands by disease specific prevalence rates:

$$am(d) = \frac{c(d)}{p(d)} \quad (6)$$

$c(d)$ cause specific mortality rate disease d

It is assumed that for any disease the attributed mortality is independent from the risk factor levels. This means that the risk factors affect the disease prognosis only through increased

risks for other diseases and mortality from other causes of death. Using the relative risk for the incidence of diseases as an approximation for relative risk for the prevalence of diseases we calculated the relative risks for other causes of death:

$$RR(oc | s_j) = \frac{RR(tot | s_j) * m(tot)_{0S} - \sum_d RR(d | s_j) * am(d) * p(d)_{0S}}{m(oc)_{0S}} \quad (7)$$

$$p(d)_{0S} = \frac{p(d)}{\sum_j RR(d | s_j) * s_j} \quad (8)$$

$$m(oc)_{0S} = \frac{m(oc)}{\sum_j RR(oc | s_j) * s_j} \quad (9)$$

$$m(tot)_{0S} = \frac{m(tot)}{\sum_j RR(tot | s_j) * s_j}$$

$RR(oc | s_j)$ *relative risk for other cause mortality smoking class j*

$m(oc)_{0S}$ *baseline other cause mortality rate for non smoking cohort*

$p(d)_{0S}$ *baseline prevalence rate disease d for non smoking cohort*

$m(tot)_{0S}$ *baseline all cause mortality rate for non smoking cohort*

These equations can be solved for $RR(oc | s_j)$ by substituting equations (8) and (9) into equation (7). In a similar fashion, relative risks for other causes mortality of for overweight and obesity can be derived. Given $RR(oc | s_j)$ and $RR(oc | b_k)$ the baseline other cause mortality rate can be found:

$$RR(oc | s_j, b_k) = RR(oc | s_j) * RR(oc | b_k) \quad (10)$$

$$m(oc)_0 = \frac{m(oc)}{\sum_{j,k} RR(oc | s_j) * RR(oc | b_k) * s_j * b_k} \quad (11)$$

$RR(oc | s_j, b_k)$ *relative risk for other cause mortality smoking class j BMI class k*

$RR(oc | b_k)$ *relative risk for other cause mortality BMI class k*

$m(oc)_0$ *baseline other cause mortality rate for 'healthy living' cohort*

$p(d)_0$ baseline prevalence rate disease d for 'healthy living' cohort

Model simulation part

Risk factors and diseases are linked through relative risks of disease incidence for each risk factor. That is, incidence rates for each risk factor class are found as relative risks times baseline incidence rate. The general assumption used is that conditional on the risk factors included, the disease event rates are independent. For the 'healthy living cohort' relative risks equal one. Formula (12) denotes the change over time in the prevalence rate of disease d for a cohort, homogeneous in its risk factor class prevalence, as a function of relative risks, incidence and mortality rates:

$$\frac{dp(d|t)}{dt} = (i(d)_0 * RR(d|s_j) * RR(d|b_k) - em(d) * p(d|t)) * (1 - p(d|t)) \quad (12)$$

$p(d|t)$ prevalence rate disease d at time t

$i(d)_0$ baseline incidence rate disease d for 'healthy living' cohort

$RR(d|s_j)$ relative risk for disease d for smoking class j

$RR(d|b_k)$ relative risk for disease d for BMI class k

$em(d)$ excess mortality rate disease d

The CDM describes disease prevalence numbers for each disease separately and it is assumed that the disease-specific attributed mortality rates are additive. Given the relations between disease specific attributed mortality, other causes mortality, disease prevalence rates and relative risks we can describe the change in population numbers needed to estimate life expectancy:

$$\frac{dN(t)}{dt} = RR(oc|s_j) * RR(oc|b_k) * m(oc)_0 * N(t) - \sum_d am(d) * p(d|t) * N(t) \quad (13)$$

$RR(oc|s_j)$ relative risk for other causes mortality smoking class j

$RR(oc|b_k)$ relative risk for other causes mortality BMI class k

$m(oc)_0$ baseline other causes mortality rate for 'healthy living' cohort

$am(d)$ mortality rate attributed to disease d

The difference of the mortality rates for persons with and without the disease can be interpreted as the excess mortality rate for that disease. However, in a model with multiple diseases these excess mortality rates cannot be interpreted as mortality uniquely attributable to a disease, since the excess mortality rates can also be caused by other co-morbid chronic diseases, e.g. coronary heart disease being a complication of diabetes. Therefore, in the calculation of the prevalence rates excess mortality rates are used, while in the calculation of the number of survivors disease specific attributed mortality rates are used.

Postprocessing

Calculating life expectancy

Equation (14) displays the formula with which we estimated remaining life expectancy for the different cohorts:

$$LE = \frac{\sum_t n(t)}{n(0)} \quad (14)$$

LE *Life Expectancy*
n(t) *number of survivors of the cohort at time t*
n(0) *initial size of the cohort at time 0*

Calculating health care costs

To assign the Costs of Illness data that give total costs per BMI or smoking related disease to costs of individual patients, total costs per disease were divided by the disease prevalence numbers for 2003:

$$cp(d) = \frac{COI(d)}{p(d)} \quad (15)$$

cp(d) *annual costs per patient having smoking and/or BMI related disease d*
COI(d) *total costs disease d*

Given the estimated annual costs per disease per patient, health care costs are then simply the product of these costs times the prevalence numbers for that disease estimated with the CDM:

$$hc(d | t) = cp(d) * p(d | t) * n(t) \quad (16)$$

hc(d|t) health care costs disease d at time t

To calculate health care costs for all other diseases that are unrelated to smoking and/or BMI the numbers of survivors were multiplied by age and sex specific average health care costs for other diseases (i.e. costs remaining when costs of causally related diseases are subtracted from total costs):

$$co = ac - \frac{\sum_d COI}{pop} \quad (17)$$

co annual health care costs per person for all other diseases

ac average annual health care costs per person

Since the COI study is a top down study in which health care costs are uniquely attributed to diseases we assume that health care costs for patients with more than one CDM disease equals the sum of the disease costs for these individual diseases. The resulting equation for estimating lifetime health care costs then becomes:

$$lhc = \sum_t n(t) * \{co + \sum_d p(d | t) * cp(d)\} \quad (18)$$

lhc lifetime health care costs

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