

Planning Tool Guidance

Project Big Life

2019-06-28

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Chapter 1

Welcome to Project Big Life's Planning Tool

1.1 What is the Project Big Life Planning Tool

To do: Develop video showcasing the platform including what it is and why someone should use the platform

1.2 Who made the Project Big Life Planning Tool?

The Project Big Life Planning Tool was developed by the Project Big Life Team. The Project Big Life Team is part of the ICES. The below video explains what ICES is.

`## PhantomJS not found. You can install it with webshot::install_phantomjs(). If it is installed, please`

Chapter 2

Introduction

The Project Big Life Planning Tool was developed in order to support health professionals: research, plan, develop, and evaluate evidence-based health interventions.

For instance Project Big Life Planning Tool helps:

- Public health professionals: assess the impact of a preventative intervention on a health behaviour
- Health planners: assess the need for palliative care

What types of questions can it answer?

The Project Big Life Planning Tool can answer the following types of questions:

- What is the burden of smoking on life expectancy?
- How many deaths would be prevented if everyone met their daily exercise requirements?

How does it work?

- This tool provides health planners with access to multivariable predictive risk algorithms, created and housed by the Project Big Life Team.
- The multivariable predictive risk algorithms use distinct characteristics and health profiles of groups of people to assess the risk of a health outcome (e.g. Life Expectancy).
- The multivariable predictive risk algorithms are developed and validated using data routinely collected by Statistics Canada and provincial health agencies, and the algorithms have been published in various journals.
- More information about multivariable predictive risk algorithms can be found in the key concepts (Chapter 6).

Why should I use it?

- It is **easy** and **flexible** to use.
 - The user only needs to upload their data and choose which calculation to run.
 - It can be used to assess the future risk of a health outcome.
 - It can be used to assess the effectiveness of different intervention scenarios (e.g. policy) on a health outcome.

- It generates **accurate** predictions.
 - It can be used to accurately assess the risk of a health outcome in populations that were not used in its development, and groups of people that account for only a fraction of the population.
- It is **Private**.
 - Uploaded data remains on your computer and is not uploaded or sent anywhere.

Chapter 3

Getting Started

To help you get started with Project Big Life's Planning Tool quickly, we built a Tutorial directly onto the platform.

The tutorial takes you through step-by-step how to use Project Big Life's Planning Tool. The tutorial will not explain the steps in detail (Chapter 4) nor will it provide reference material (Chapter 7), but it will give you an understanding of how easy it is to use the Planning Tool!

To access the tutorial, go onto Project Big Life's Planning Tool (<http://policy.projectbiglife.ca/>) and click on the Tutorial button in the top right corner!

Chapter 4

How To

These guides will cover similar topics as the tutorials but in greater detail.

- Customize Data
- Load Data
- Select Outcome
- Filter Data
- Stratify Data
- Calculate Outcomes
- Generate Intervention Scenarios
- Visualize Data
- Export Data
- Resolve Error Messages

4.1 Customize Data

Prior to using the Project Big Life Planning Tool you may want to manipulate your dataset. Reasons for data manipulation may include: custom filter or custom stratification.

Data manipulation can occur on any programming software: R, SAS, STATA, etc, provided you output your dataset as a .csv file.

The following table shows the example R, SAS, STATA code for the following:

Put example in to the Consult with Doug To clarify: Not sure whether to continue with this section, as there are multiple ways to manipulate the data. Could provide an example

4.2 Load Data

Data *loaded* to the Project Big Life Planning Tool remains on your computer and is not uploaded or sent anywhere.

Note The Project Big Life Planning Tool can currently only support .csv data files from the 2013/2014 Canadian Community Health Survey *TBD - whether it's shared*.

There are two options for your data: use a sample file or upload your own file.

4.2.1 Use a sample file

If you don't have your own data or want to explore the platform capabilities prior to your data you can use the sample files already on the There are **X** sample files you may use to complete your calculations:

- MockPUMF2013.csv is a subsample of the Public Use Microdata File from the [2013 Canadian Community Health Survey] (<https://www150.statcan.gc.ca/n1/en/catalogue/82M0013X>).

Click on the file name under the **Sample files** to select it.

4.2.2 Load your own file

Click the browse button under **Select a file to use in calculations**. Locate the file on your computer, select, and open.

If the loaded file has all of the variables required and recommended for calculation, you will be able to continue with the planning tool.

- If the loaded file does **not** have all the variables **required** for the calculation you will not be able to continue with the planning tool.
- If the loaded file does **not** have all the variables **recommended** for calculation you will be able to continue with the planning tool, however the calculations may be less accurate.

4.3 Select Outcome

Click on box beside the calculation name under **Select initial calculations**.

TO Do: add details about the calculations can be found in the glossary and key concepts

4.4 Filter Data

Click on the + Add filter button.

Select the variable that you want to filter on.

To Clarify: CAT vs. Continous - what are we letting them filter on *Clarify with Luke if this is done, WKLY ALCOHOL Consumption**

To add another filter repeat the steps above. A maximum of three filters are recommended to maintain statistical power (added filters reduce sample sizes and reduces statistical power).

You are able to filter on all types of variables: required for calculation, recommended for calculation, and ignore variables (includes customized variables).

4.4.1 Remove a filter

To remove a filter entirely, click on the trash can beside the filter you want to delete.

To remove a level within a filtered variable, click on the 'x' beside the variable level.

4.5 Stratify data

Select the variables you want to stratify on Under ‘Stratifications’. A maximum of 3 stratifications are recommended to maintain statistical power (added strata reduce strata sample size and reduces statistical power).

You are able to stratify on categorical variables, but not continuous.

You are able to stratify on all types of categorical variables: required for calculation, recommended for calculation, and ignore variables (includes customized variables).

With Luke - prevent stratification of continuous variables

4.6 Calculate Outcomes

Name your calculation to quickly differentiate multiple calculations.

Note the larger the data file is the longer the calculations will take. It may take a few minutes for the calculation to complete. *TBD: There has been discussion on changing the current method. Once a method has been selected then Indicate how they know that the calculation is being preformed.*

4.7 Generate Intervention Scenarios

TBD: How much of it will be included in the platform. Will inform how to write this section

4.8 Visualize Data

TBD: Need plots on the platform to work through the steps below - export - create your own(?)

4.9 Export Data

Click on the **Download results** button under the **Results** section.

Select which calculations you’d like to download.

To Do: Screenshot of all the calculation options once the platform is fixed.

4.10 Resolve Warning or Error Messages

TO DO: confirm with Luke possible error messages and what they mean

- Invalid category
- Out of range
- Not a number
- Sample Size is too small

Chapter 5

Applications

This chapter provides you with examples of how Project Big Life's Planning Tool can be used in your day-to-day operations. The examples will cover: cause-deletion, generating a health status report and determining the impact of a local and national policy.

5.1 Cause-deleted life expectancy

What would be the life expectancy of a population if be no one in the population ever smoked? This scenario is a cause-deleted scenario.

Cause-deleted life expectancy is the estimated life expectancy of a population if a specific cause (e.g. smoking) did not exist in that population. This population is known as the counterfactual population.

Cause-deleted effect of life expectancy is calculated by comparing the population with the current exposure status of smoking status and pack-years of smoking to a counterfactual population where these two variables are: smoking status = never smoker, and pack-years of smoking = 0. It is measured in life years lost. This calculation can also be preformed with the health outcome: risk of mortality. Further explanation of cause-deleted risk and cause-deleted effect of a risk can be found in key concepts (Chapter ??keyconcepts)).

Lets walk through this scenario step-by-step!

5.2 Health Status Report

5.3 Diet

5.4 Transportation

Chapter 6

Key Concepts

This section explains some key concepts in Project Big Life's Planning Tool. This section will explain how it works rather than how to do things.

- Data and sample files
- Multivariable predictive risk algorithms
- Calculations
 - Risk of health outcome
 - Number of health outcomes
 - Life expectancy
 - Cause-deleted
- Health intervention scenarios

6.1 Data and sample files

The Project Big Life Planning Tool currently accepts **2013/2014 Public Use Microdata File and Shared File of the Canadian Community Health Survey (CCHS) in 'csv' format.**

The CCHS is an annual cross-sectional survey preformed by Statistics Canada. The CCHS collects information related to health status, health care utilization, and health determinants for the Canadian population. Data is shared at the the sub-provincial geographic level (health region or combination of health regions).

- Details about the survey and its design can be found on Statistic Canada website (<http://www23.statcan.gc.ca/imdb/p2SV.pl?Function=getSurvey&Id=144170>).
- Details and access to the Public Use Microdata file (PUMF) can be found on the Odesi website (<https://search2.odesi.ca/#/details?uri=%2Fodesi%2Fcchs-82M0013-E-2013-2014-Annual-component.xml>)

6.1.1 Sample Files

- MockPUMF2013.csv is a subsample of the Public Use Microdata File from the [2013 Canadian Community Health Survey] (<https://www150.statcan.gc.ca/n1/en/catalogue/82M0013X>). It includes variables: age, sex, BMI, smoking habits, alcohol consumption, diet components, physical activity levels, chronic conditions, ethnicity, immigration status, and education.

6.2 Multivariable predictive risk algorithms

Multivariable predictive risk algorithms predict the future risk of health outcomes (e.g. Life Expectancy) for a population using routinely collected health data.

Multivariable predictive risk algorithms can be used to:

- Project the number of new cases of the health outcome
- Estimate the contribution of specific risk factors of the health outcome
- Evaluate effectiveness of health interventions
- Describe the distribution of risk in the population (diffused or concentrated)

Multivariable predictive risk algorithms are able to assess equity issues compared to competing population risk methods (e.g. World Health Organization Global Burden of Disease).

More information on what multivariable predictive risk algorithms are and how they can be used can be found the journal article: *Predictive risk algorithms in a population setting: an overview* (Manuel D, 2012)

6.2.1 Development of multivariable predictive risk algorithms

****Data:****

- Multivariable predictive risk algorithms are created using routinely collected data that includes information about risk factors (exposure) and health events (outcomes).
- Data is collected at an individual level through population health surveys (e.g. Canadian Community Health Survey) and administrative databases (e.g. Vital Statistics). Data sources are linked together when the individual has given permission too.
- Individuals are followed overtime until the health event (e.g. death or disease) occurs.
- Separate data is collected to create a derivation cohort and validation cohort(s).
 - Note: The risk factors that are collected are from population health surveys and are self-reported; no clinical data (e.g. blood pressure) is collected. Risk factors focus on health behaviours (e.g. smoking) and sociodemographic factors, commonly associated with health outcome.

Algorithm generation:

- Multivariable predictive risk algorithms are cox proportional hazard models that analyze time to health outcome (e.g. death) *Question for Carol - The models are not cox-proportional hazard models but they are similar?*
- Multivariable predictive risk algorithms are developed and validated in 4 stages:
 - Algorithm derivation: the predictive risk algorithm is created using data from the derivation cohort
 - Algorithm validation: the predictive risk algorithm is applied to the validation cohort
 - Final algorithm generation: validation and derivation cohorts are combined to estimate the final application of the predictive risk algorithm
 - Derivation of the application algorithm: creation of a parsimonious (fewer predictors) algorithm that maintained discrimination, calibration, and overall algorithm performance
- In each stage of the algorithm development and validation, algorithm performance is assessed using measures of discrimination and calibration.

6.2.2 Multivariable predictive risk algorithms built in Project Big Life Planning Tool

- There is currently 1 multivariable predictive risk algorithm is built into to Project Big Life planning tool.

Title

Outcomes

Information

Mortality Population Risk Tool

5 year risk of death, Life Expectancy, Cause deleted

Appendix A

6.3 Calculations

6.3.1 Risk of health outcome

Risk of the health outcome (e.g. risk of dying) is the outcome of the multivariable predictive risk algorithm. An example of the multivariable risk algorithm is:

$$\text{Risk} = \sum_t h_0(t) * e^{\beta_{pred.smoking} * x_{smoking} + \beta_{pred.cancer} * x_{cancer} + \beta_{pred.male} * x_{male} + \dots}$$

Where:

- t = survival time
- $h_0(t)$ = the baseline hazard
- β_{pred} = predictive hazard ratios for the exposures
- x = the exposure (e.g. $x_{smoking}$ can have the value: current smoker, former smoker ≤ 5 years, former smoker > 5 years, or never smoker).

6.3.2 Number of health outcomes

The number of health outcomes (e.g. Summary Deaths) is calculated through the following steps:

1. Risk of the health outcome is calculated for each individual in the dataset using the multivariable predictive risk algorithm.
2. The weighted mean of all the risks is calculated using individual risk values and corresponding study weights (CCHS variable: WTS_M).
3. The weighted mean is then multiplied with the total number of individuals in the population to generate the number of health outcomes (e.g. number of deaths in 5 years).

6.3.3 Life expectancy calculation

Life expectancy is calculated using abridge life tables:

1. The mortality risk for each individual is calculated using the multivariable predictive risk algorithm for mortality (MPoRT - details can be found in Appendix A).

2. Mortality risks for each individual are weighted using the corresponding survey weights (CCHS variable - WTS_M).
3. Weighted mortality risks are then used in sex-specific 5-year abridge period life tables.
4. Life expectancy was then calculated using these 5-year abridge period life tables.

6.3.4 Cause-deletion calculations

What would be the life expectancy of a population if be no one in the population ever smoked? This scenario is a cause-deleted scenario.

To Do: Insert video of cause deleted calculations that explains Method 1 vs. Method 2

There are two parts to the cause-deleted calculations: (A) calculate the risk, and (B) calculate the health outcome: life expectancy or number of deaths.

Part A: Risk calculations

The original multivariable predictive risk algorithm is:

$$\text{Risk} = \sum_t h_0(t) * e^{\beta_{pred.smoking} * x_{smoking} + \beta_{pred.cancer} * x_{cancer} + \beta_{pred.male} * x_{male} + \dots}$$

Step 1. Modify the original algorithm to include the external coefficient(s). This means replacing all predictive hazard ratios/betas related to the health behaviour to the causal hazard ratios/betas.

- Remove the original regression coefficient(s) for the health behaviour.
- Add the new external coefficient(s) to the algorithm. External coefficients are generated from either: causal models, or from systematic reviews or meta-analysis.

$$\text{External coefficient risk} = \sum_t h_0(t) * e^{\beta_{causal.smoking} * x_{smoking} + \beta_{causal.cancer} * x_{cancer} + \beta_{pred.male} * x_{male} + \dots}$$

Step 2. Risk is calculated using the modified algorithm created in Step 1 and the respondent's original profile (e.g. current smoker). This is the “external coefficient risk”.

$$\text{External coefficient risk} = \sum_t h_0(t) * e^{\beta_{causal.smoking} * (\text{current smoker}) + \beta_{causal.cancer} * x_{cancer} + \beta_{pred.male} * x_{male} + \dots}$$

Step 3. “Cause-deleted risk” is calculated by setting an exposure to a reference (non-exposed) value (all other risk exposures remain unchanged).

$$\text{Cause-deleted risk}' = \sum_t h_0(t) * e^{\beta_{causal.smoking} * (\text{never smoker}) + \beta_{causal.cancer} * x_{cancer} + \beta_{pred.male} * x_{male} + \dots}$$

Step 4. The “cause-deleted effect external” is calculated as “external coefficient risk” (Step 2) minus the “cause-deleted risk” (Step 3).

$$\text{Cause-deleted effect}_{external} = \text{External coefficient risk} - \text{Cause-deleted risk}'$$

Step 5. Original risk is calculated using the original algorithm and the original respondent's profile.

only -cbf.bb

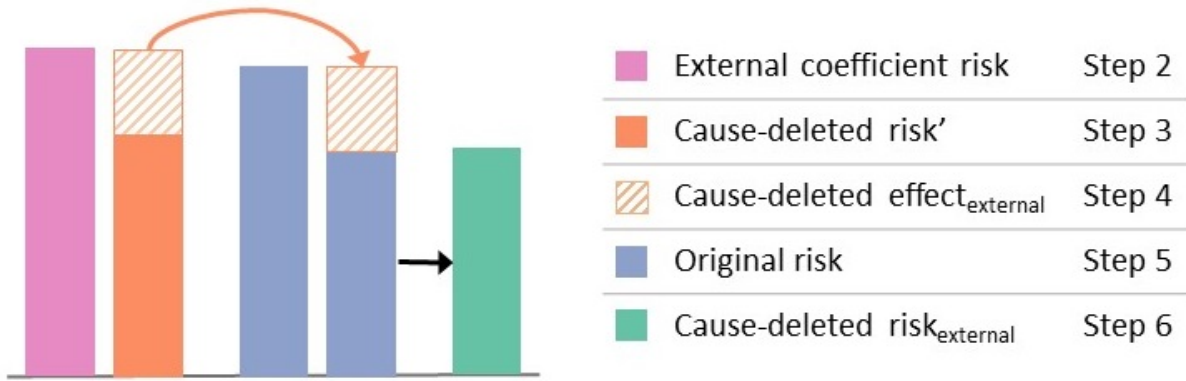


Figure 6.1: Risk portion of cause-deleted calculations

$$\text{Original risk} = \sum_t h_0(t) * e^{\beta_{\text{pred.smoking}} * (\text{current smoker}) + \beta_{\text{pred.cancer}} * x_{\text{cancer}} + \beta_{\text{pred.male}} * x_{\text{male}} + \dots}$$

Step 6. The “cause-deleted risk external” is calculated by “original risk” (Step 5) minus the “cause-deleted effect external” (Step 4).

$$\text{Cause-deleted risk}_{\text{external}} = \text{Original risk} - \text{Cause-deleted effect}_{\text{external}}$$

Part B: Health outcome calculations

Using risks generated above you can then calculate:

- cause-deleted life expectancy or life years lost attributable to a health behaviour (exposure)
- cause-deleted number of deaths or number of deaths attributable to a health behaviour (exposure)

Life expectancy calculations

Step I: Calculate the original life expectancy by using the original risk (Step 5 above) in the sex-specific 5-year abridge period life tables.

Step II: Calculate the cause-deleted life expectancy by using the cause-deleted risk external (Step 6 above) in the sex-specific 5-year abridge period life tables.

Step III: Calculate life years lost attributable to a health behaviour by: original life expectancy (Step I) minus cause-deleted life expectancy (Step II):

$$\text{Life years lost} = \text{Original life expectancy} - \text{Cause-deleted life expectancy}$$

Number of deaths calculations

Step I: Calculate the number of deaths that would occur using the original risk (Step 5 above).

Step II: Calculate the number of deaths that would occur using the cause-deleted risk external (Step 6 above).

Step III: Calculate the number of deaths that are attributable to a health behaviour (exposure) by: original number of deaths (Step I) minus cause-deleted number of deaths (Step II):

$$\text{Deaths due to exposure} = \text{Original number of deaths} - \text{Cause-deleted number of deaths}$$

6.4 Health interventions scenarios

TBD: whether this will be built into the platform or not. Depending on the outcome will dictate how I write the following section (e.g. do they have to manipulate thier data)

Chapter 7

Glossary

5-year mortality risk

The probability that an individual will die in the next 5 years.

Body Mass Index (BMI)

A weight-to-height ratio used as an indicator of obesity and underweight. BMI is calculated by dividing an individual's body weight in kilograms by the square of height in metres (kg/m²).

Burden

The impact or size of a health problem in an area, measured by cost, mortality, morbidity or other indicators. The burden of unhealthy behaviour is calculated by the differences in life expectancy based on individuals' exposure to four health behavioural risks for poor health relative to the healthy category.

By Row Measures

When selected, the output 'csv' file will include the result of the calculation for each row (e.g. individual) of the dataset.

Calibration

The agreement between predicted risk generated from the model and observed risk generated from the data.

Canadian Community Health Survey

An annual survey performed by Statistics Canada that collects information related to health status, health care utilization and health determinants for the Canadian population. Details about the survey can be found on Statistic Canada website (<http://www23.statcan.gc.ca/imdb/p2SV.pl?Function=getSurvey&Id=144170>).

Cause-deleted life expectancy

A cause-deleted health outcome is the estimated health outcome of a population if a specific cause (e.g. smoking) did not exist in that population.

Discrimination

The ability of the model to differentiate between high risk individuals and low risk individuals.

Error Message

Error messages will occur when variables that are “**Required for Calculation**” are missing in the data. If the entire column for the variable is missing then the calculation cannot be performed on the data. If there are missing row entries for the variable then the entire row will not be used in the calculation.

Exposure

In the risk algorithms the exposure refers to the level of the predictor variable, e.g. for the predictor variable EDUDR04 there are four levels of exposure: (1) Post-Secondary, (2) Some Post-Secondary, (3) High School, (4) Less than High School.

Filter

Chooses part of your dataset for analysis. If you filter on ‘Sex’ and then ‘Male’, calculations will only be performed on individuals that are ‘Male’ and ‘Females’ will be excluded. For example, when calculating Life Expectancy on the filter variable ‘Sex’ then ‘Male’ there will be a Life Expectancy estimate for ‘Males’ and *no* Life Expectancy estimate for ‘Females’.

Health Behaviour

Actions people do that may affect their health, positively or negatively. Health behaviours are among the determinants of health and are influenced by the social, cultural and physical environments in which people live and work. (Statistics Canada, 2010) They are also shaped by individual choices and external constraints. (Statistics Canada, 2010) The four health behaviours of **smoking**, **alcohol consumption**, **diet**, and **physical activity** are specified in Project Big Life’s planning tool.

Ignored Variables

Are not included in the calculation. It does not matter if your dataset includes these variables or not. Ignored variables can be used for filter and stratification.

Life Expectancy (LE)

Life expectancy is a calculation of how long a person or population would be expected to live, on average, given unchanging risk of death from a specific point in time.

Metabolic Equivalent of Task (MET)

The metabolic equivalent of task (MET) is a measure of the rate of energy expenditure from an activity; a measure of calories burned by type, duration and frequency of physical activity. The reference value of 1 MET is defined as the energy expenditure rate at rest which is equal to 1kcal/kg/day.

Predictor

A variable that is used in the algorithm to predict the outcome.

Recommend for calculation

Variables that are included in the calculation but not necessary for the calculation to run. Rather these variables increase the accuracy of the results.

Required for calculation

Variables that are included in the calculations and are necessary for the calculation to run. If a dataset does not have these variables then the calculation will not run.

Risk

The probability of a health event occurring at some point of time in the future.

Socioeconomic Position

People in poorer socioeconomic circumstances generally have poorer health. Deprivation measures identify those who experience material or social disadvantage compared to others in their community. The Deprivation Index for Health in Canada developed by the Institut national de santé publique du Québec (INSPQ) (Pampalon R, Raymond G, 2000) is used in this planning tool. The index includes education, employment and income as measures of material deprivation; and single-parent families, living alone, or being divorced, widowed or separated as measures of social deprivation. The deprivation index was used to assign geographical areas into socioeconomic position groups (low, middle and high) based on material and social quintiles. High-deprivation neighbourhoods were those in the top two quintiles for both social and material deprivation. Low-deprivation neighbourhoods were those in the bottom two quintiles.

Stratification

The separation of data into smaller strata (levels or classes which individuals are assigned too). If the variable 'Sex' is stratified it creates two strata: 'Male' and 'Female'. Calculations are performed on each strata (level or class) and the outcome will be specific to that strata. For example, when calculating Life Expectancy on the stratified variable 'Sex' there will be a Life Expectancy estimate for 'Males' and a different Life Expectancy estimate for 'Females'.

Summary Measures

When selected, the output .csv file will include the calculation result for the entire population of the dataset.

Warning Message

Warning messages will occur when variables that are **“Recommended for Calculation”** are missing in the data. If the entire column for the variable is missing the calculation will still be performed on the data. If there are missing row entries for the variable the row will still be used in the calculation.

Appendix A

Mortality Population Risk Tool (MPoRT)

Outcomes: 5-yr risk of death, Life Expectancy, Cause-deleted Life Expectancy

Calculations

Using MPoRT you are able to calculate:

- 5 year mortality risk
- Number of deaths
- Life Expectancy
- Cause-deleted deaths and life expectancy
- Burden of health behaviour in deaths and on life expectancy

Types of Questions

- What is the burden of smoking on life expectancy?
- How many deaths would be prevented if everyone met their daily exercise requirements?

Description: A multivariable predictive risk model that estimates the future risk of all-cause death in Canada. It adjusts for health behaviours: smoking, unhealthy alcohol consumption, poor diet, and physical inactivity, and a wide range of other risk factors.

Versions of MPoRT have been developed since 2012 and used in various studies. Each version of MPoRT (v1.0, v1.2, v2.0) used the Ontario subset of the Canadian Community Health Survey (CCHS) for development and the survey respondents were linked to personal death records. In later versions of MPoRT (v1.2, v2.0) the following changes were made: (a) algorithm variables were adjusted to improve predictions, and (b) the algorithms were validated using: the Ontario subset of CCHS of the years that were not used in development and the National CCHS dataset (excluding Ontario).

MPoRTv1.0 Was used in the “Seven More Years” report, a joint report with Public Health Ontario and IC/ES (<https://www.ices.on.ca/Publications/Atlases-and-Reports/2012/Seven-More-Years>). In summary, the algorithm estimated the risk of death associated with health behaviours: smoking, unhealthy alcohol consumption, poor diet, physical inactivity and stress. There were approximately 550,000 person-years of follow up and over 6000 deaths in the development dataset. The algorithm used categorical predictor variables for health behaviours and sociodemographic factors.

MPoRTv1.2 Was published in PLoS (<https://journals.plos.org/plosmedicine/article?id=10.1371/journal.pmed.1002082>). In summary, the algorithm estimated the risk of death associated with health behaviours:

smoking, unhealthy alcohol consumption, poor diet, and physical inactivity (stress was removed due to its low prediction ability). There were approximately 1 million person-years of follow up and over 9000 deaths in the development and validation datasets. The algorithm used multiple continuous predictor variables, and added chronic disease predictor variables and interaction terms.

MPoRTv2.0 - The version used in Project Big Life’s Planning Tool This version of MPoRT has not yet been published.

Development: This predictive risk model was developed using Ontario subsets of the 2001 to 2008 CCHS and participants were linked to personal health records. There were approximately 1.3 million person-years of follow-up and over 15,000 deaths in the developmental dataset.

Validation: This predictive risk model was validated using three different datasets: Ontario subset of the 2009 to 2012 CCHS, National dataset (except Ontario) of the 2003 to 2008 CCHS, and the National dataset of the 2000 and 2005 National Health Interview Survey in the United States of America. In all validation datasets individuals were linked to personal health records.

Parameters: The parameters used in this predictive risk model are:

Category

Variable

Scale

Description

Demographic

Age*

Continuous

5 knot spline. Valid range 20 to 102

Sex

Dichotomous

Stratified Female/Male

Health Behaviour

Pack years of smoking

Continuous

3 knot spline. Valid range: 0 to 78 (Female), 0 to 112.5 (Male)

Smoking Status

Categorical

Non-smoker

Current Smoker

Former Smoker ≤ 5 years

Former > 5 years

Alcohol (number of drinks per week)

Continuous

4 knot spline (Females) and 3 knot spline (Males). Valid range: 0 to 25 (Female), 0 to 50 (Male)

Former/non-drinker

Dichotomous

Yes/No

Simplified diet score

Continuous

3 knot spline. Valid range: -18.9 to 20.7 (Female), -16.8 to 18.4 (Male)

Leisure physical activity (MET)

Continuous

3 knot spline. Valid range: 0 to 12.4 (Female), 0 to 16 (Male)

Socio-demographic

Ethnicity

Categorical

White

Black

Chinese

Arab; South Asian; West Asian

Filipino; Japanese; Korean; Southeast Asian

Other; Indigenous; Latin American; Multiple origin; unknown

Immigrant

Dichotomous

Yes/No

Fraction of lifetime in Canada

Continuous

3 knot spline[†]. Valid range: 0 to 1

Education

Categorical

Less than secondary

Secondary School Graduation

Some Post-Secondary

Post-Secondary Graduation

Neighbourhood social and material deprivation

Ordinal

Low (1st or 2nd quantile)

High (4th or 5th quantile)

Moderate (all others)

Chronic Conditions

Diabetes

Dichotomous

Yes/No

High Blood Pressure

Dichotomous

Yes/No

Chronic Respiratory Disease

Dichotomous

Yes/No

Mood Disorder

Dichotomous

Yes/No

Cancer

Dichotomous

Yes/No

Dementia

Dichotomous

Yes/No

Heart Disease

Dichotomous

Yes/No

Stroke

Dichotomous

Yes/No

Epilepsy

Dichotomous

Yes/No[‡]

BMI

Continuous

3 knot spline. Valid range: 8.9 to 47.2 (Female), 8.6 to 43.7 (Male)

* Age interaction included for all variables except immigrant, fraction of time in Canada, and ethnicity † Excluded in the male model, remains in the female model ‡ Excluded in the female model, remains in the male model

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