

# Introduction to the Diabetes Population Risk Tool (DPoRT)



Evidence Guiding Health Care



# Learning Objectives

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1. To understand the principles of risk prediction algorithms
2. To understand the development and validation of DPoRT
3. To identify DPoRT's applications to population based diabetes risk assessment and public health planning

# Diabetes Population Risk Tool (DPoRT)

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A decision-support tool that uses routinely collected population characteristics applied to a validated risk prediction algorithm to estimate the number of new and existing diabetes cases in a population of interest for the purpose of:

- ▶ Understanding distribution of risk in the population
- ▶ Prevention
- ▶ Resource planning
- ▶ Facilitating decision-making and priority setting

# DPoRT Knowledge-to-Action Work

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## OVERALL GOAL

For researchers and decision-makers in varied health related settings to **work collaboratively** to **build capacity** and **facilitate the application of DPoRT** as a strategic aid for population-based risk assessment, intervention and planning decisions.



# Examples of DPoRT in Action

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## Practice

- **Public Health Units** integrated DPoRT into staff training and operational protocols, public reporting on websites, and annual internal reporting

## Advocacy

- **Medical Officers of Health** used DPoRT to estimate impact of “The Big Move” on decreasing diabetes incidence through active transportation
- DPoRT findings presented to **Mississauga City Council** to advocate for active transport investment

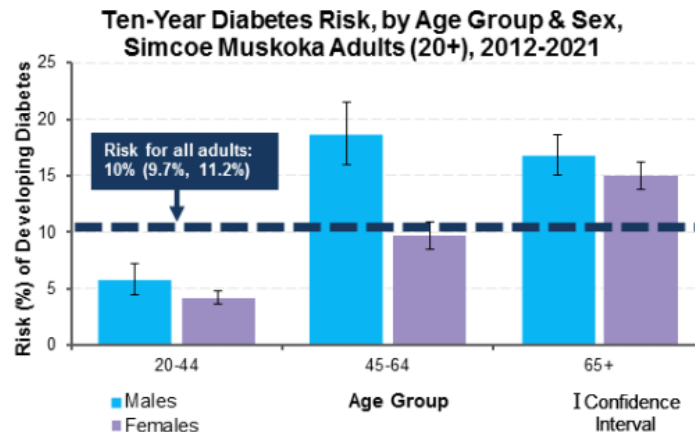
## Program/ Policy

- **Ministry of Health and Long-Term Care** used DPoRT to estimate impact of scaling up the Primary Care Diabetes Prevention Program

# Examples of DPoRT in Action



According to the Diabetes Population Risk Tool (DPoRT) an adult (20+) living in Simcoe Muskoka that did not have diabetes in 2011 has a 10% (9.7%, 11.2%) chance of developing diabetes in the ten year time period from 2012 to 2021. The risk of developing diabetes is highest among males aged 45 to 64 years (19% (15.9%, 21.5%)) and highest among females aged 65 years and older (15% (13.7%, 16.2%)).



Data Source: Rosella LC et al. 2010. A population based risk algorithm for the development of diabetes: development and validation of the Diabetes Population Risk Tool (DPoRT). J Epidemiology Community Health, doi:10.1136/jech.2009.102244.

- Simcoe Muskoka District Health Unit, Health Stats webpage
- An example of using DPoRT for public health reporting

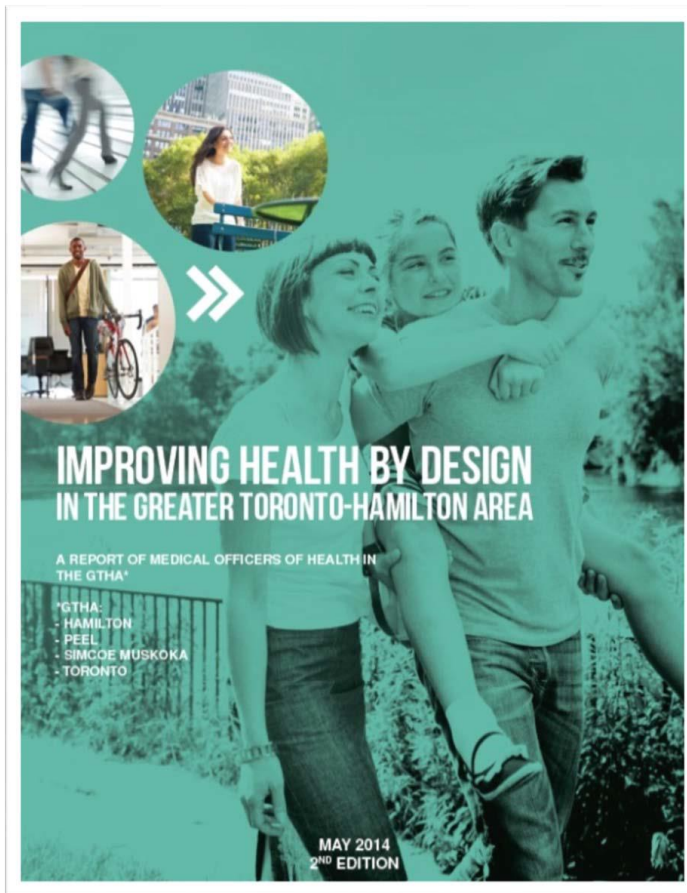


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Simcoe Muskoka District Health Unit, Health Stats webpage: <http://www.simcoemuskokahealthstats.org/topics/chronic-diseases/diabetes/prevalence-and-incidence#Incidence>.

# Examples of DPoRT in Action

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- Improving Health by Design, Medical Officers of Health report
- An example of using DPoRT for advocacy

# Risk Algorithms

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## Risk algorithms

- ▶ Predicts risk of an outcome – usually a disease state
- ▶ Traditionally calculated for individuals
- ▶ Typically used when there are multiple factors that contribute to risk

## Population-based risk tool

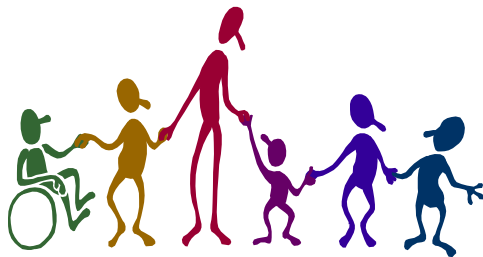
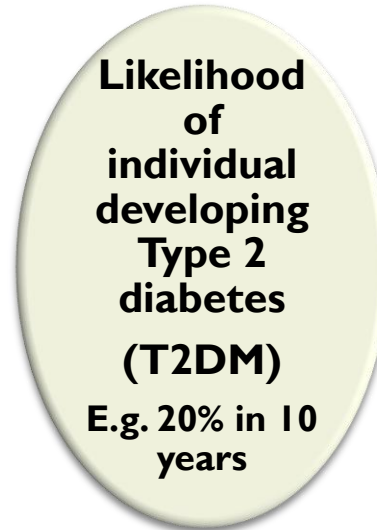
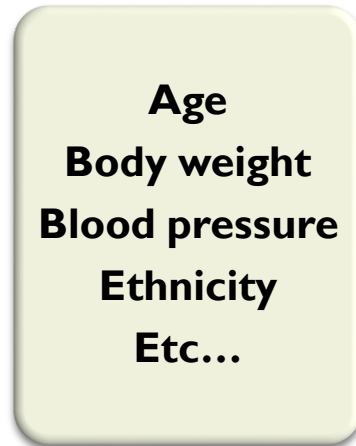
- ▶ Can be summarized for groups
- ▶ Can be applied as a decision-support and planning tool
- ▶ Have unique methodological and data challenges



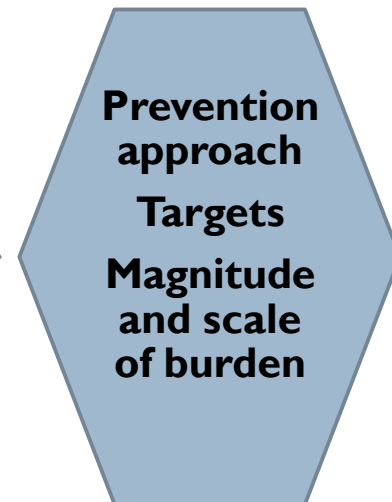
# Individual vs. Population-based Risk Prediction Tools



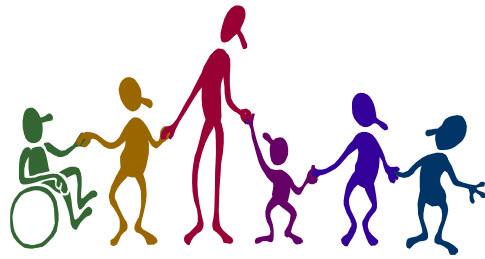
E.g., Diabetes Risk Score  
Farmingham Risk Score



E.g., DPoRT



# Improving Population Health



**Age distribution**  
**% obese**  
**% hypertensive**  
**Ethnic distribution**  
**Etc...**



**Ten-year T2DM risk in a region**  
**(country, province, health region)**



**Prevention approach**  
**Targets**  
**Magnitude and scale of burden**



Inform changes to programs and policy:  
Home, school and work environment  
Food production  
Walkability and active transport  
Socioeconomic status

# Development and Validation of DPoRT

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**Objective:** To develop a population based risk tool for Type 2 Diabetes Mellitus that is valid, reliable and accessible for various levels of health

## Validity in this context:

1. With the available factors is this the best model that can be found? (statistical)
2. Does the model predict accurately for its intended purpose? (policy relevant)



## KEY CHALLENGE

Balancing accessibility, relevance and model performance

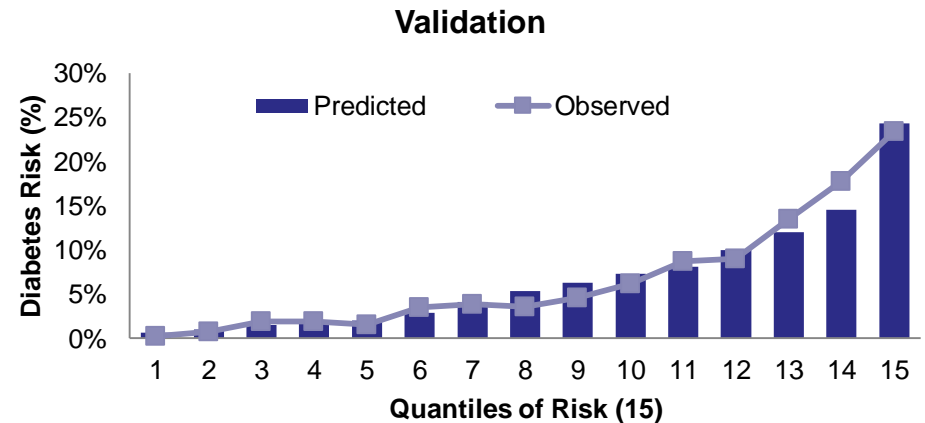
# Development and Validation of DPoRT

## Linkage

- ▶ External validation in two provinces and two time points

## Complex survey design + prediction

- ▶ Parametric survival models applied to survey data
- ▶ Optimal predictor determination
- ▶ Bootstrap variance
- ▶ Incorporate survey weights in prediction



# Development and Validation of DPoRT

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**Development Cohort:** Linked 1996/97 NPHS in ON (N=23,403)

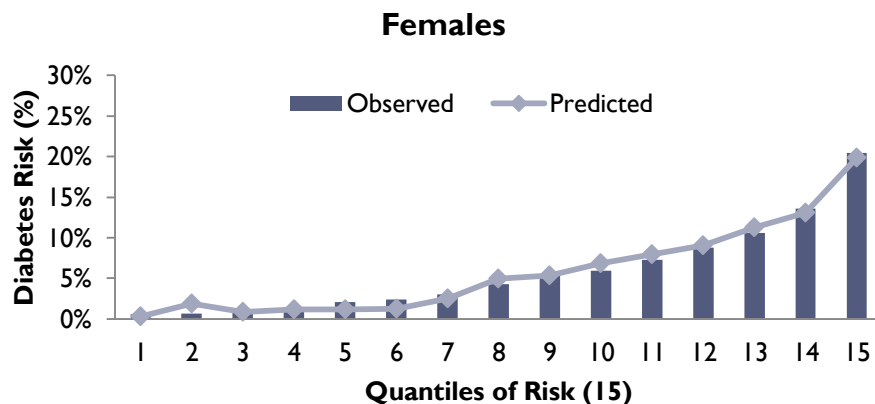
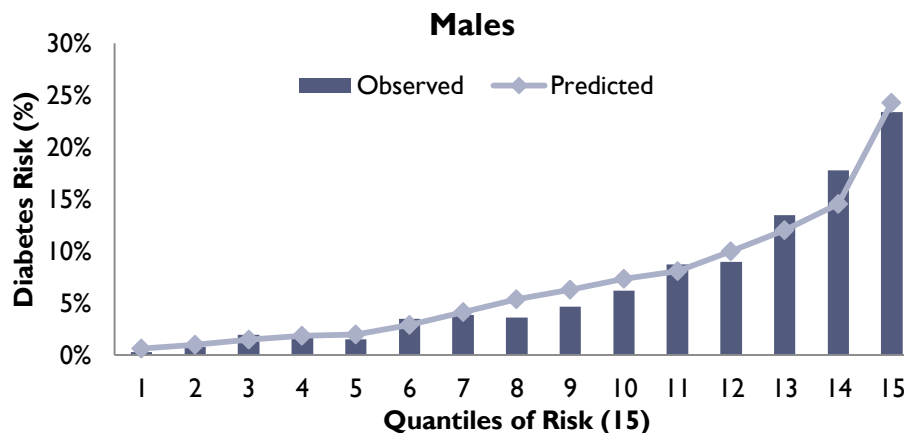
**Validation Cohort 1:** Linked 2000/01 CCHS in ON (N=37,463)

**Validation Cohort 2:** Linked 1996/97 NPHS in MB (N=10,118)

- ▶ Risk variables: only those that are routinely and publicly available (in the NPHS and CCHS)
- ▶ Outcome: physician-diagnosed type 2 diabetes (Ontario Diabetes Database & MB version)

# DPoRT 2.0 – algorithm was recently updated

- ▶ Predicted versus observed incidence of diabetes for men and women in Ontario, validation datasets across quintiles (15) of risk



# DPoRT Risk Algorithm

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- ▶ Sex specific DPoRT models are applied to data in population health surveys (e.g., CCHS) for those who are  $\geq 20$  years and free from diabetes at baseline

DPoRT Risk Factor Variables	
Body mass index (BMI)	Income
Age	Immigrant status
Sex	Hypertension
Ethnicity	Heart disease
Education	Smoking status

# Canadian Community Health Survey (CCHS)

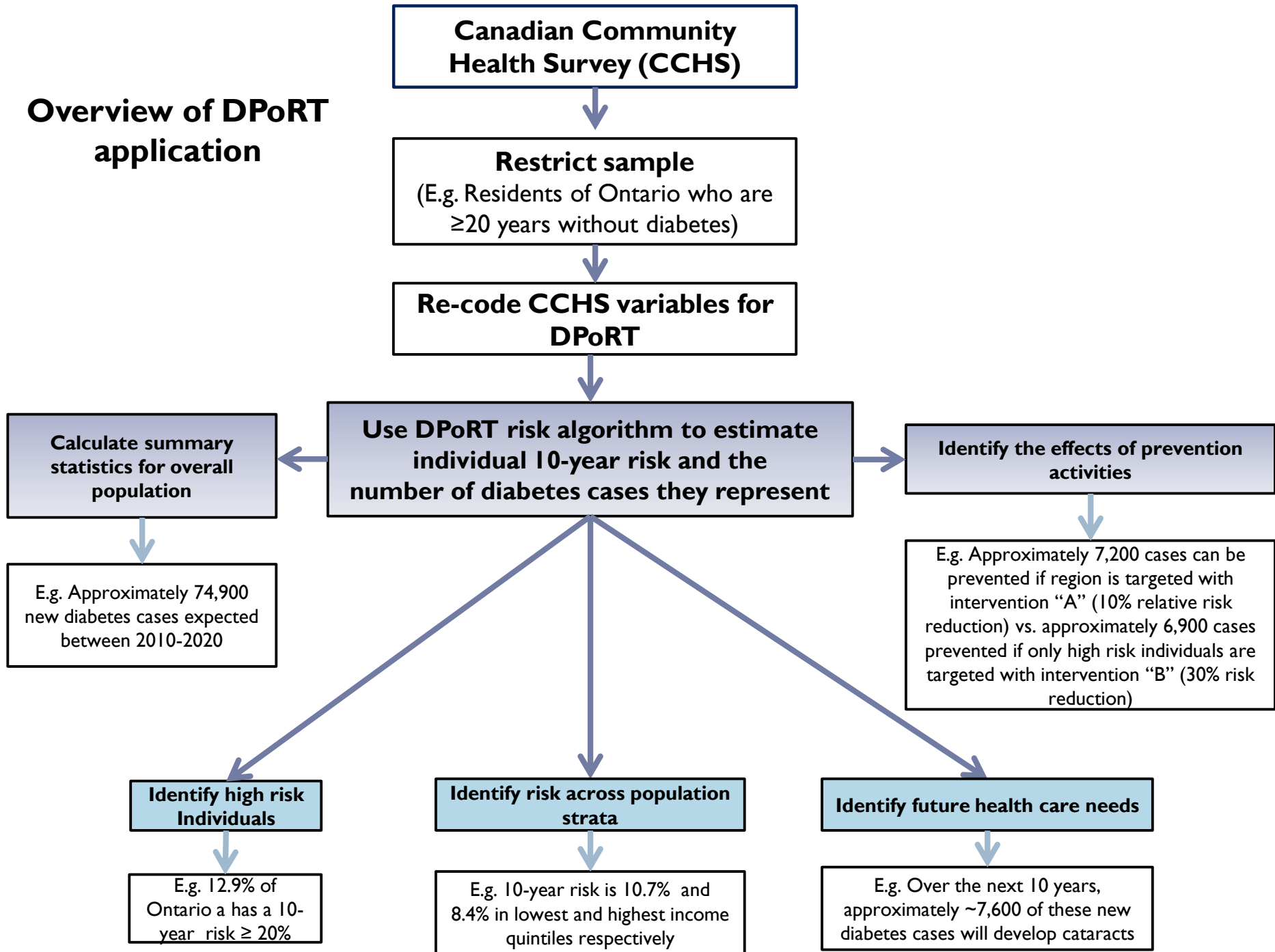
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- ▶ A cross sectional survey administered by Statistics Canada with questions on health status, determinants of health, and health care utilization
- ▶ Representative of 98% of the Canadian population 12 years and over





# Overview of DPoRT application



# DPoRT Risk Algorithm 2.0

Males:

$$\mu = 10.3062 - 0.3629 \times \text{hypertension} - 0.3483 \times \text{heart disease} - 0.5697 \times \text{non-white ethnicity} - 0.0585 \times \text{smoker} + 0.1884 \times \text{attended post-secondary} + 0.1173 \times \text{top income quintile} - 0 \times (\text{BMI} < 23 \ \& \ \text{age} < 45) - 0.5520 \times (23 \leq \text{BMI} < 25 \ \& \ \text{age} < 45) - 0.9521 \times (25 \leq \text{BMI} < 30 \ \& \ \text{age} < 45) - 1.7162 \times (30 \leq \text{BMI} < 35 \ \& \ \text{age} < 45) - 2.3310 \times (35 \leq \text{BMI} \ \& \ \text{age} < 45) - 1.3602 \times (\text{BMI} < 23 \ \& \ \text{age} \geq 45) - 1.6537 \times (23 \leq \text{BMI} < 25 \ \& \ \text{age} \geq 45) - 2.0563 \times (25 \leq \text{BMI} < 30 \ \& \ \text{age} \geq 45) - 2.5513 \times (30 \leq \text{BMI} < 35 \ \& \ \text{age} \geq 45) - 2.9353 \times (35 \leq \text{BMI} \ \& \ \text{age} \geq 45).$$

Scale = 0.7994

Females:

$$\mu = 10.5777 - 0.4098 \times \text{hypertension} - 0.4528 \times \text{non-white ethnicity} - 0.1477 \times \text{immigrant} + 0.1939 \times \text{attended post-secondary} - 0 \times (\text{BMI} < 23 \ \& \ \text{age} < 45) - 0.7432 \times (23 \leq \text{BMI} < 25 \ \& \ \text{age} < 45) - 1.1521 \times (25 \leq \text{BMI} < 30 \ \& \ \text{age} < 45) - 1.8479 \times (30 \leq \text{BMI} < 35 \ \& \ \text{age} < 45) - 2.0562 \times (35 \leq \text{BMI} \ \& \ \text{age} < 45) - 1.5832 \times (\text{BMI} = \text{missing} \ \& \ \text{age} < 45) - 0.7100 \times (\text{BMI} < 23 \ \& \ 45 \leq \text{age} < 65) - 1.2338 \times (23 \leq \text{BMI} < 25 \ \& \ 45 \leq \text{age} < 65) - 1.8357 \times (25 \leq \text{BMI} < 30 \ \& \ 45 \leq \text{age} < 65) - 2.3742 \times (30 \leq \text{BMI} < 35 \ \& \ 45 \leq \text{age} < 65) - 2.6631 \times (35 \leq \text{BMI} \ \& \ 45 \leq \text{age} < 65) - 2.1988 \times (\text{BMI} = \text{missing} \ \& \ 45 \leq \text{age} < 65) - 1.5956 \times (\text{BMI} < 23 \ \& \ \text{age} \geq 65) - 1.6144 \times (23 \leq \text{BMI} < 25 \ \& \ \text{age} \geq 65) - 1.9830 \times (25 \leq \text{BMI} < 30 \ \& \ \text{age} \geq 65) - 2.2148 \times (30 \leq \text{BMI} < 35 \ \& \ \text{age} \geq 65) - 2.6448 \times (35 \leq \text{BMI} \ \& \ \text{age} \geq 65) - 2.4209 \times (\text{BMI} = \text{missing} \ \& \ \text{age} \geq 65).$$

Scale = 0.8419

Note: All independent variables are centered on the mean value for each province.

Note: Women with missing BMI have an elevated risk of diabetes. They are included in the model.

For both males and females:

$$m = \frac{\log(\text{follow-up time in days}) - \mu}{\text{scale}}$$

$$p = 1 - \exp(-m^{\text{exp}})$$

Number of diabetes cases =  $p \times$  survey weight.

# Interpretation of DPoRT Estimates

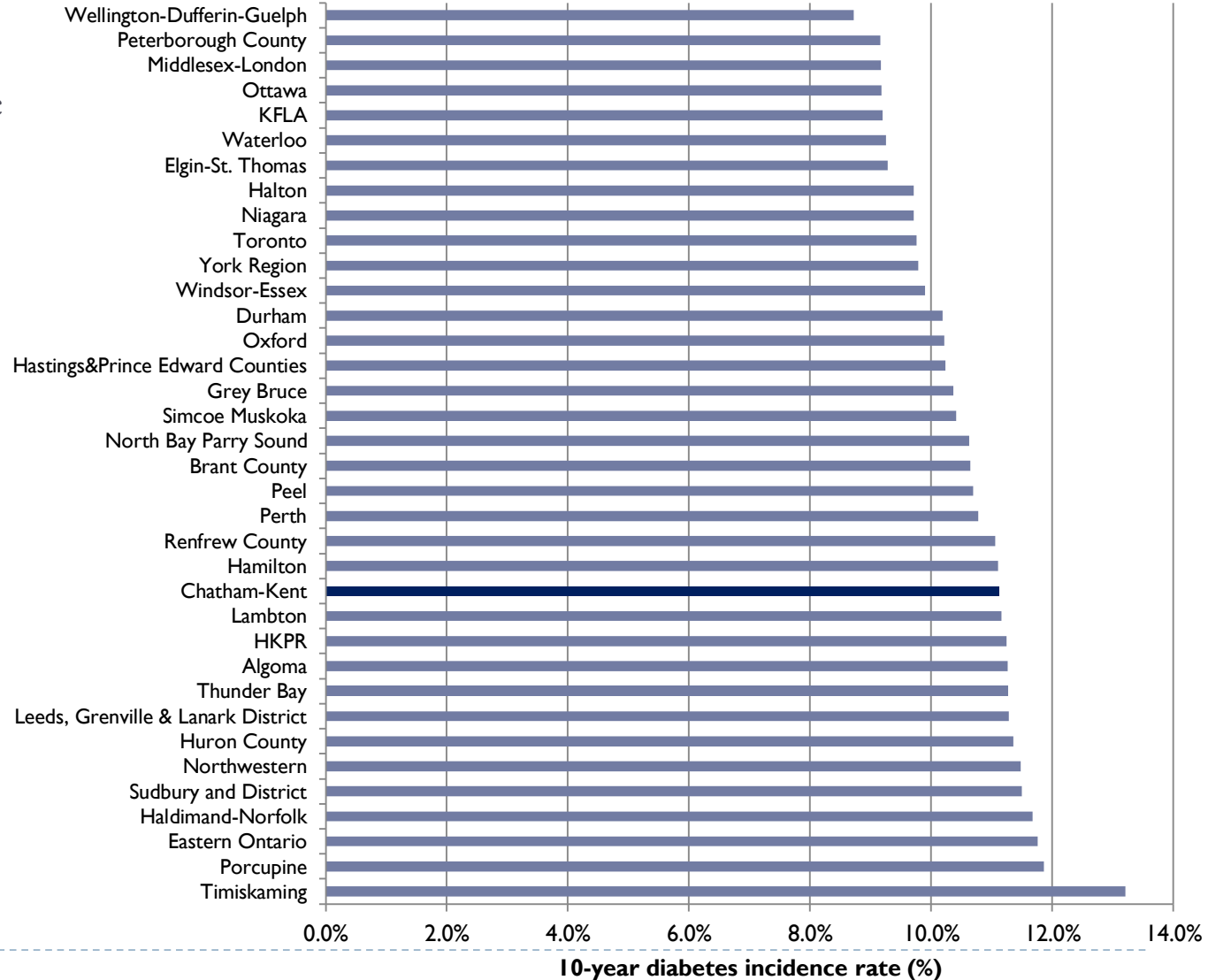
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- ▶ Interpretation of the baseline estimate of the number of new diabetes cases

*Between 2007 and 2017, **1.9 million** Canadians aged 20 and older will be newly diagnosed with diabetes, based on 2007 BMI levels and other risk factors. Canadians' average baseline risk for developing diabetes in 2007 was **8.9%**. This means that about nine out of every 100 Canadians are predicted to develop diabetes during the 10-year period.*

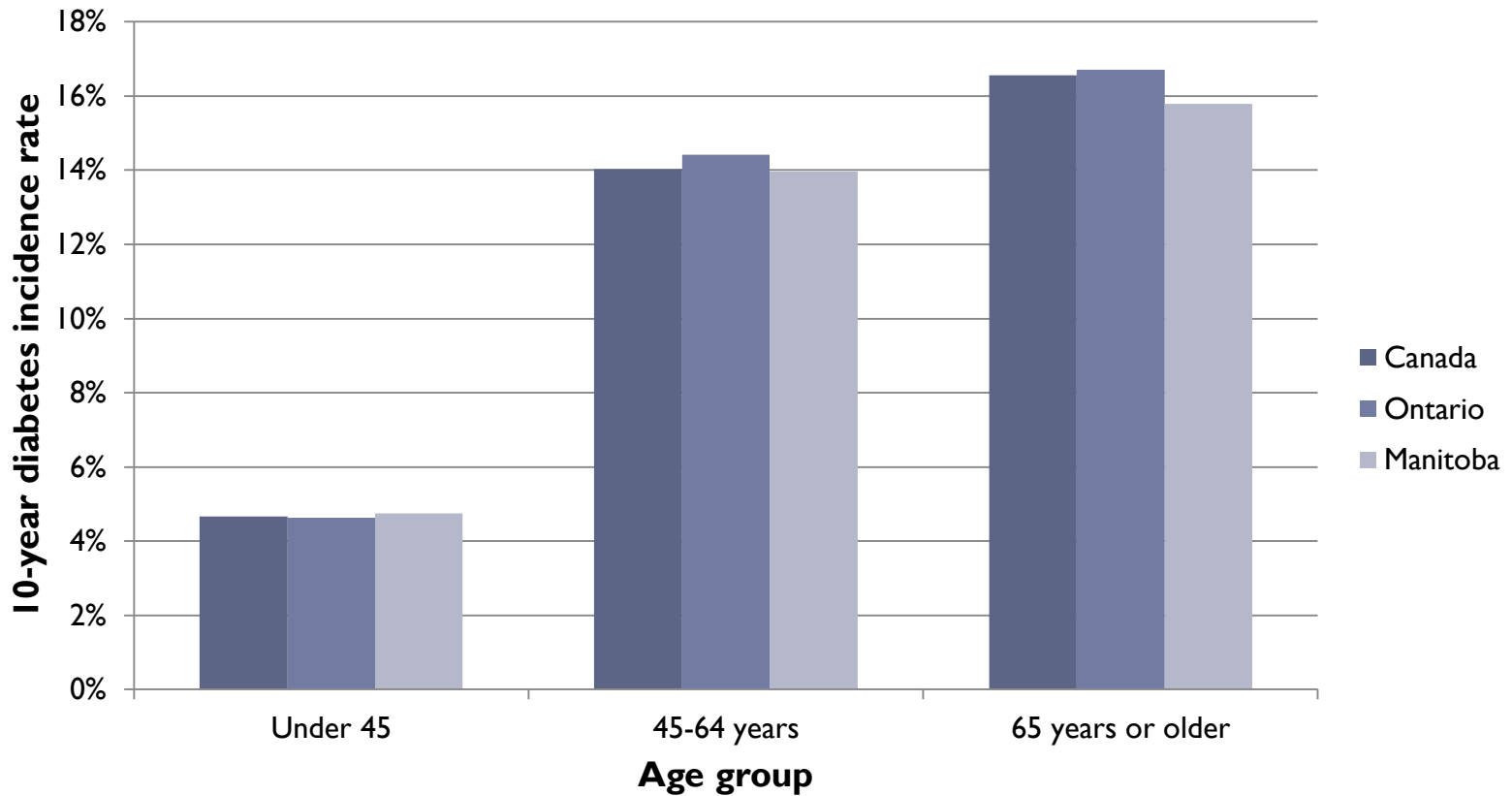
# Project Diabetes Incidence by Geographic Region

Ten-year diabetes incidence rate by public health unit in Ontario (2011/12–2021/22)



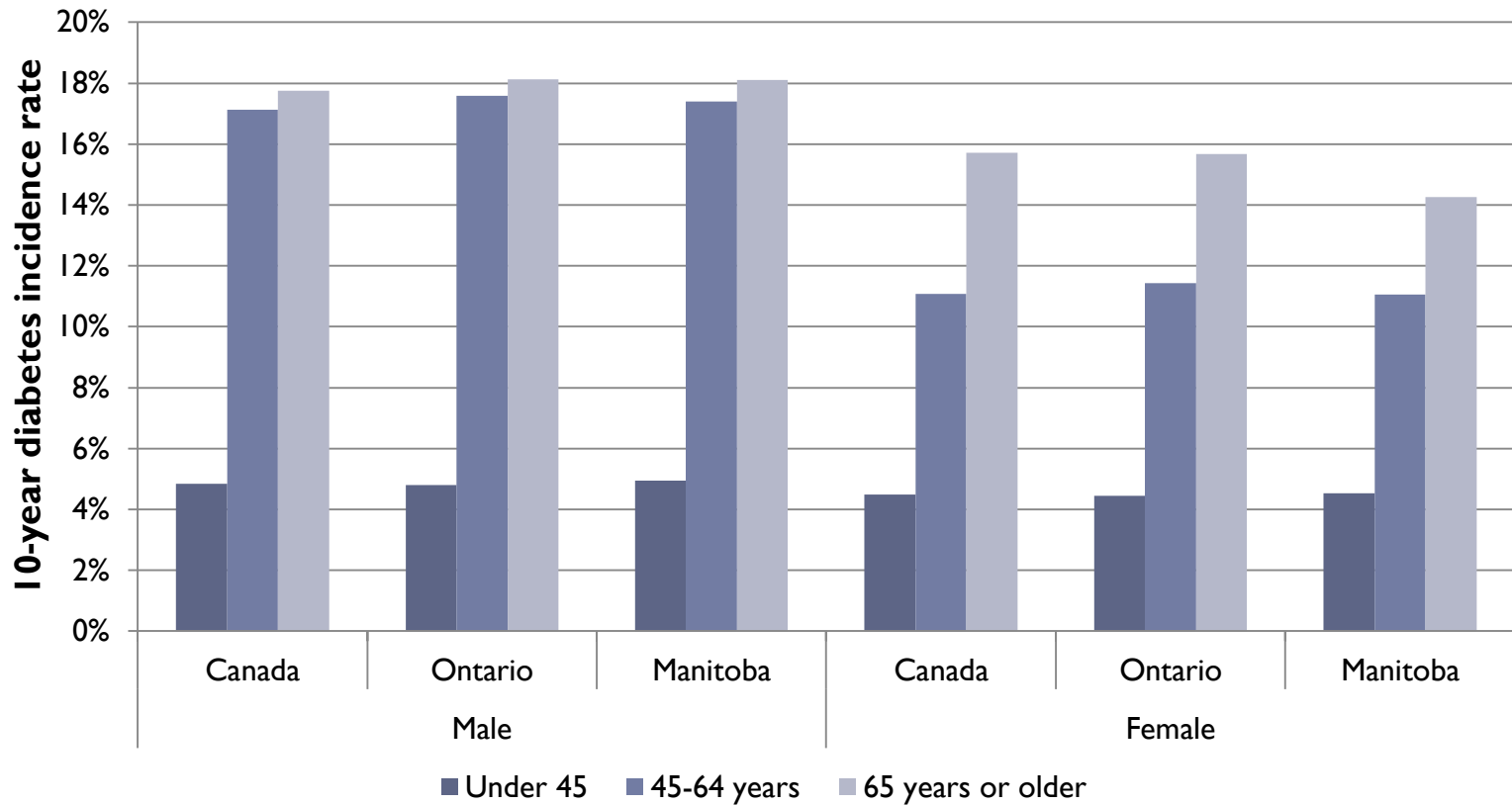
# Project Diabetes Incidence by Target Groups

Ten-year diabetes incidence rate by age group and geographic region (2011/12–2021/22)



# Project Diabetes Incidence by Target Groups

Ten-year diabetes incidence rate by age group, sex and geographic region (2011/12–2021/22)



# Growing burden of diabetes in Peel

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1 in 6 individuals in Peel will be living with type 2 diabetes



2027

2012

Approximately 140 thousand new diabetes cases will develop

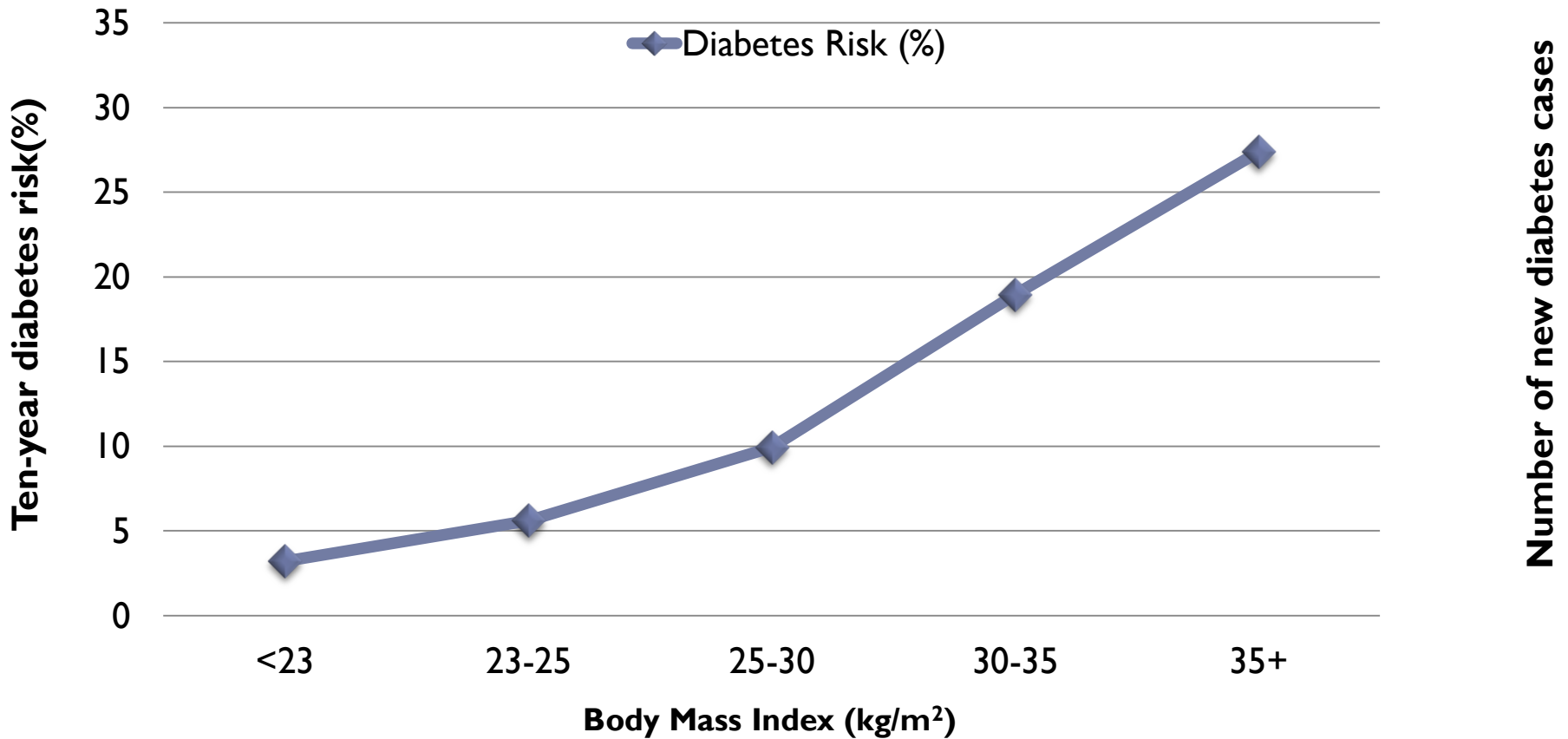
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1 in 10 individuals in Peel are living with type 2 diabetes



# Identify Targets for Diabetes Prevention Approaches

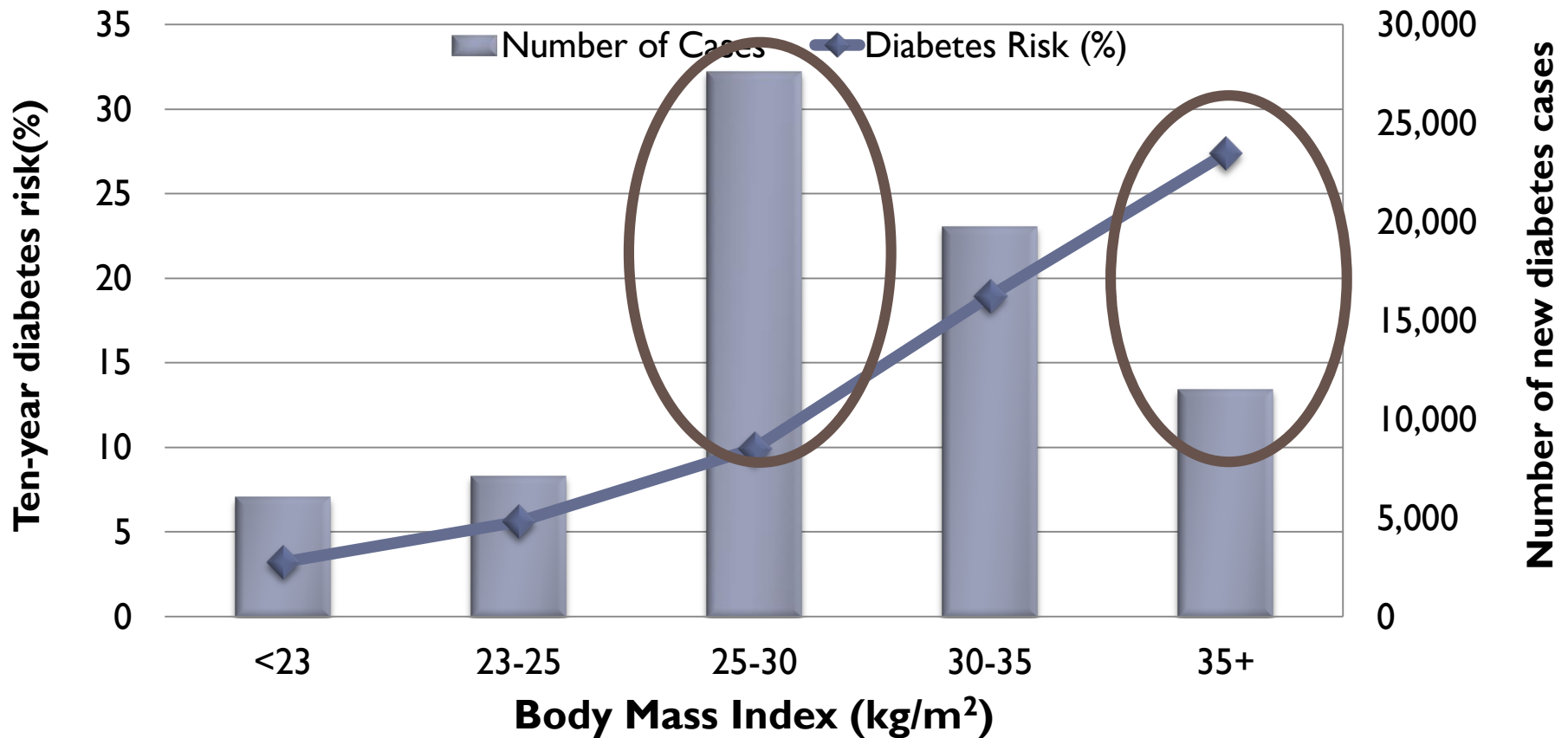
Ten-year diabetes risk according to DPoRT and number of new diabetes cases in Manitoba by BMI category (2011/12–2021/22)

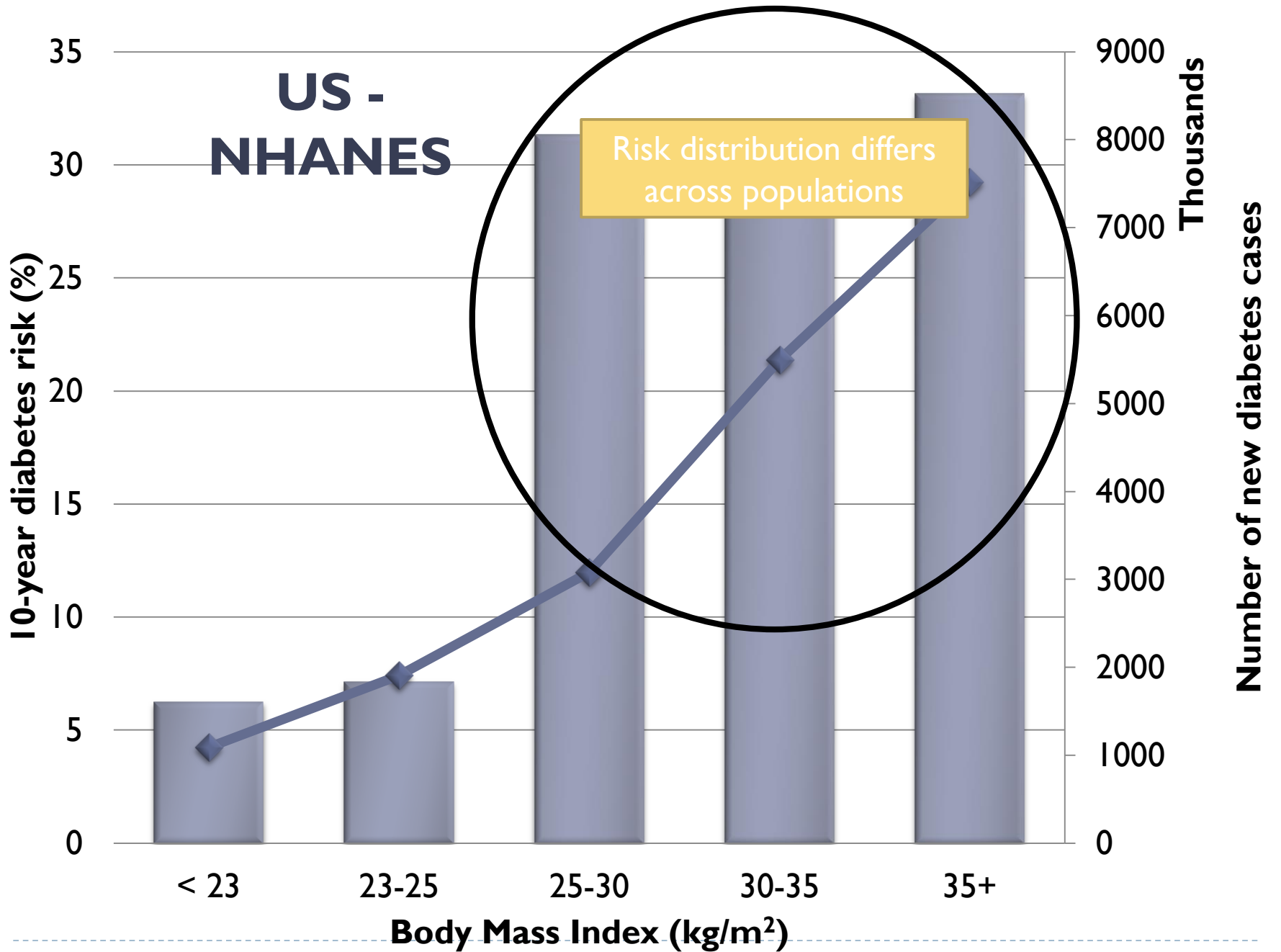




# Identify Targets for Diabetes Prevention Approaches

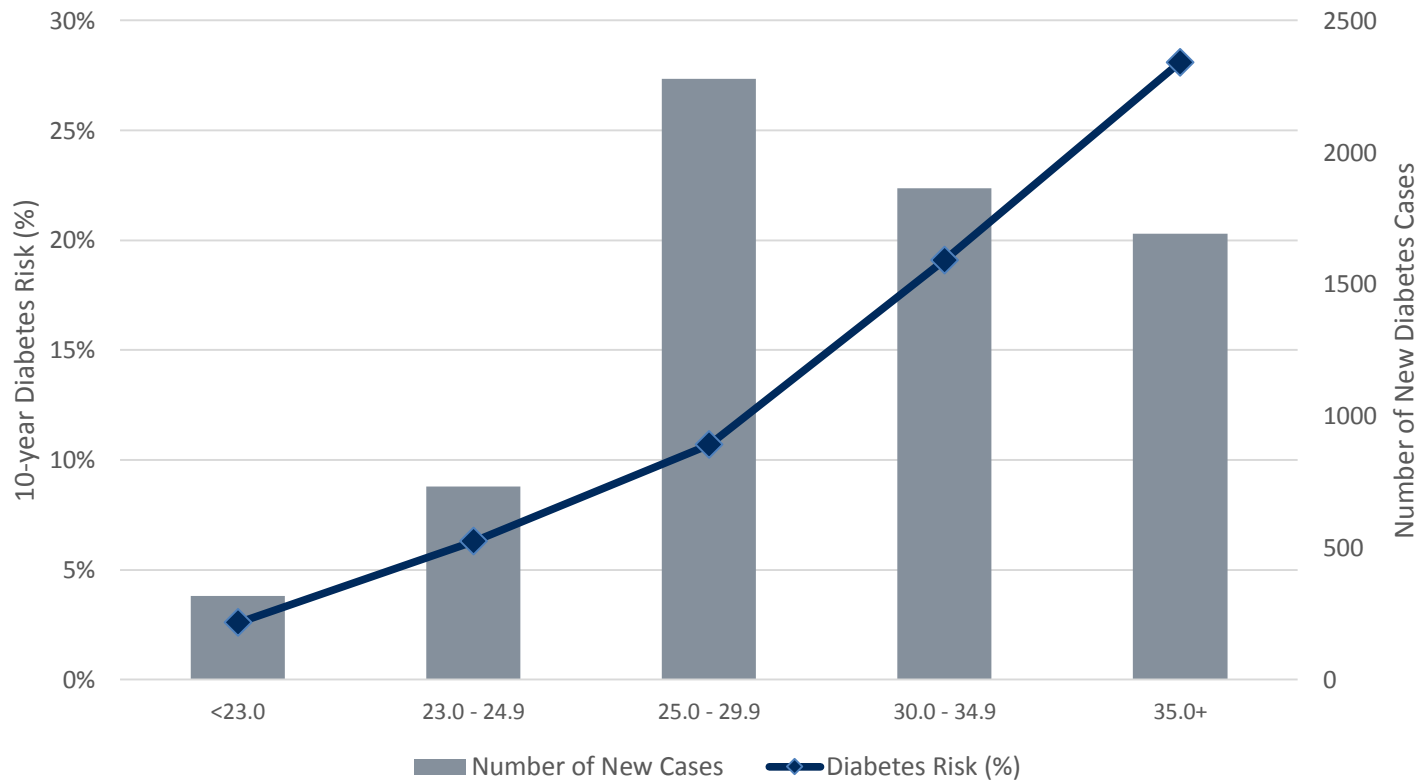
Ten-year diabetes risk according to DPoRT and number of new diabetes cases in Manitoba by BMI category (2011/12–2021/22)





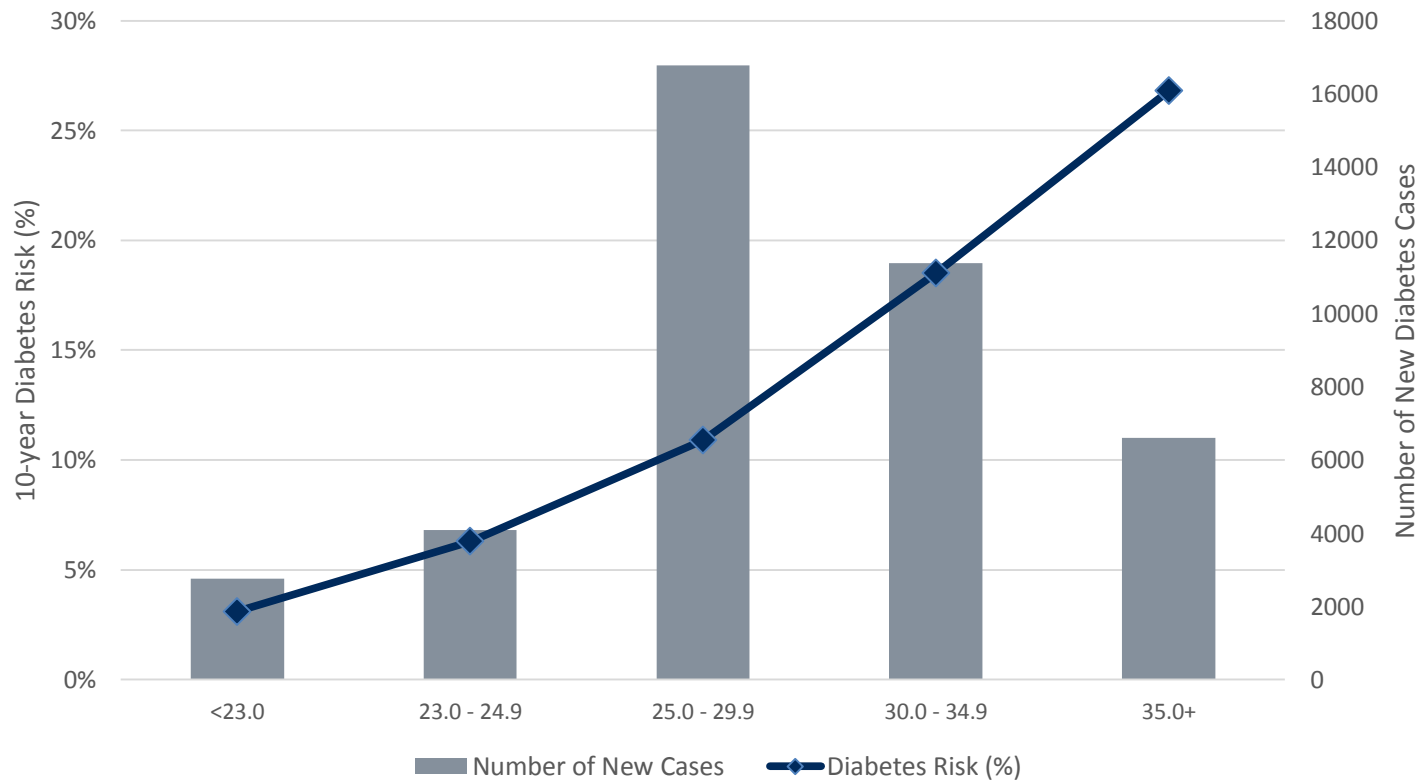
# Estimate the Impact of Diabetes Prevention Approaches

**Chatham-Kent Public Health Unit** region baseline estimated ten-year diabetes incidence rate and number of new cases (2012/13 – 2022/23)



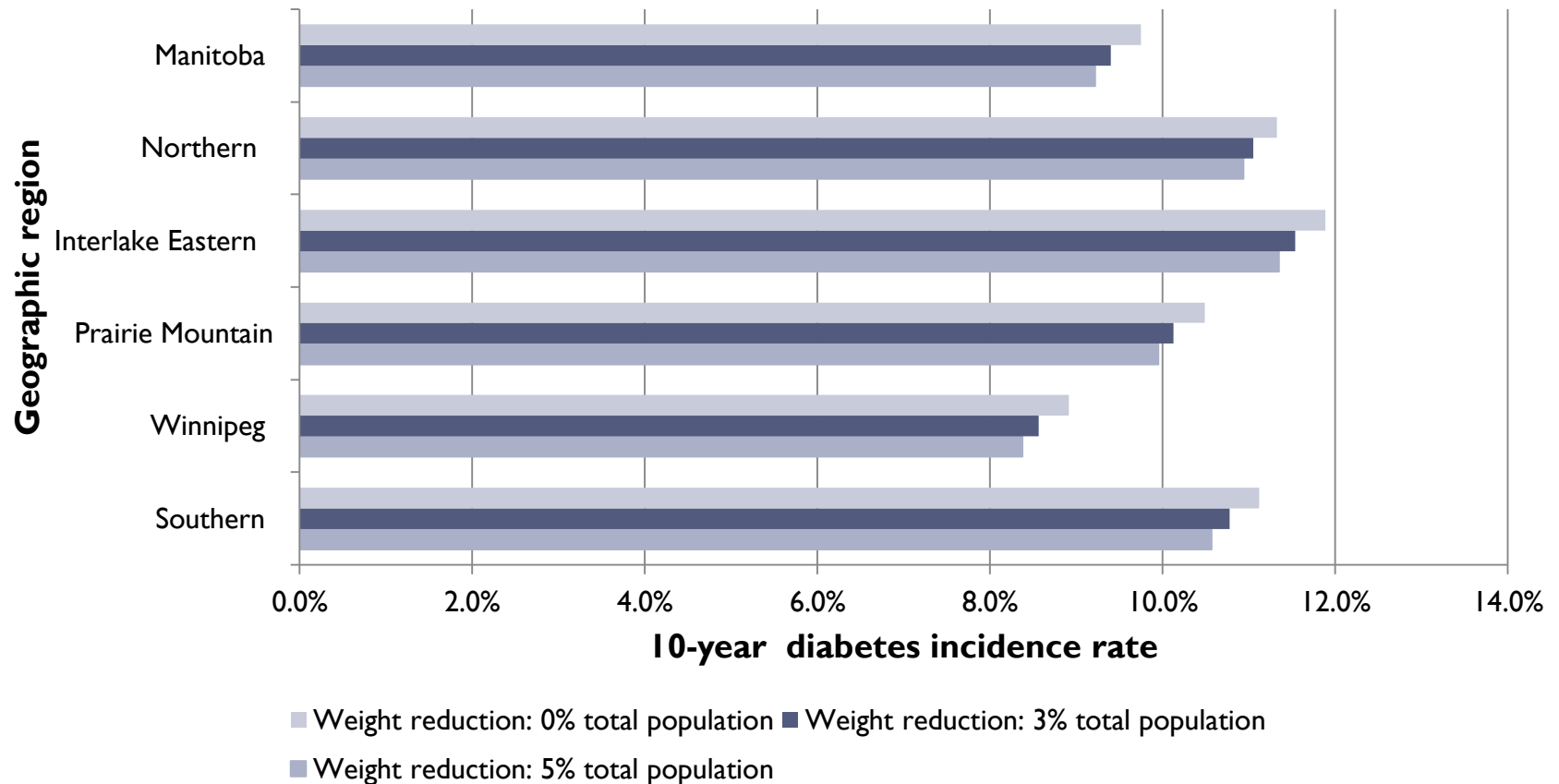
# Estimate the Impact of Diabetes Prevention Approaches

**Erie St. Clair Local Health Integration Network** region baseline estimated ten-year diabetes incidence rate and number of new cases (2012/13 – 2022/23)



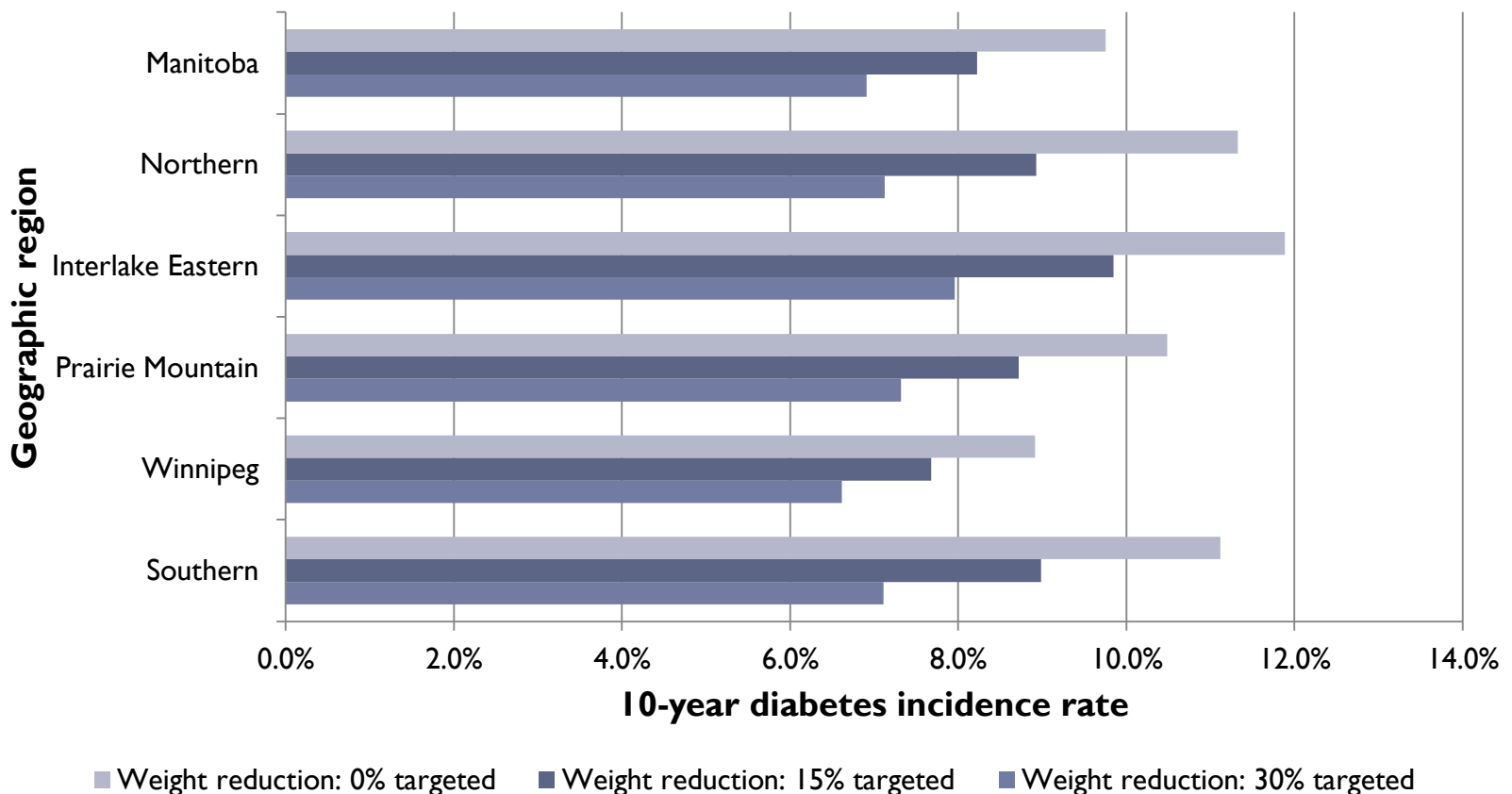
# Estimate the Impact of Diabetes Prevention Approaches

Expected change in ten-year diabetes incidence rate in Manitoba as a result of population-level weight loss (2011/12–2021/22)



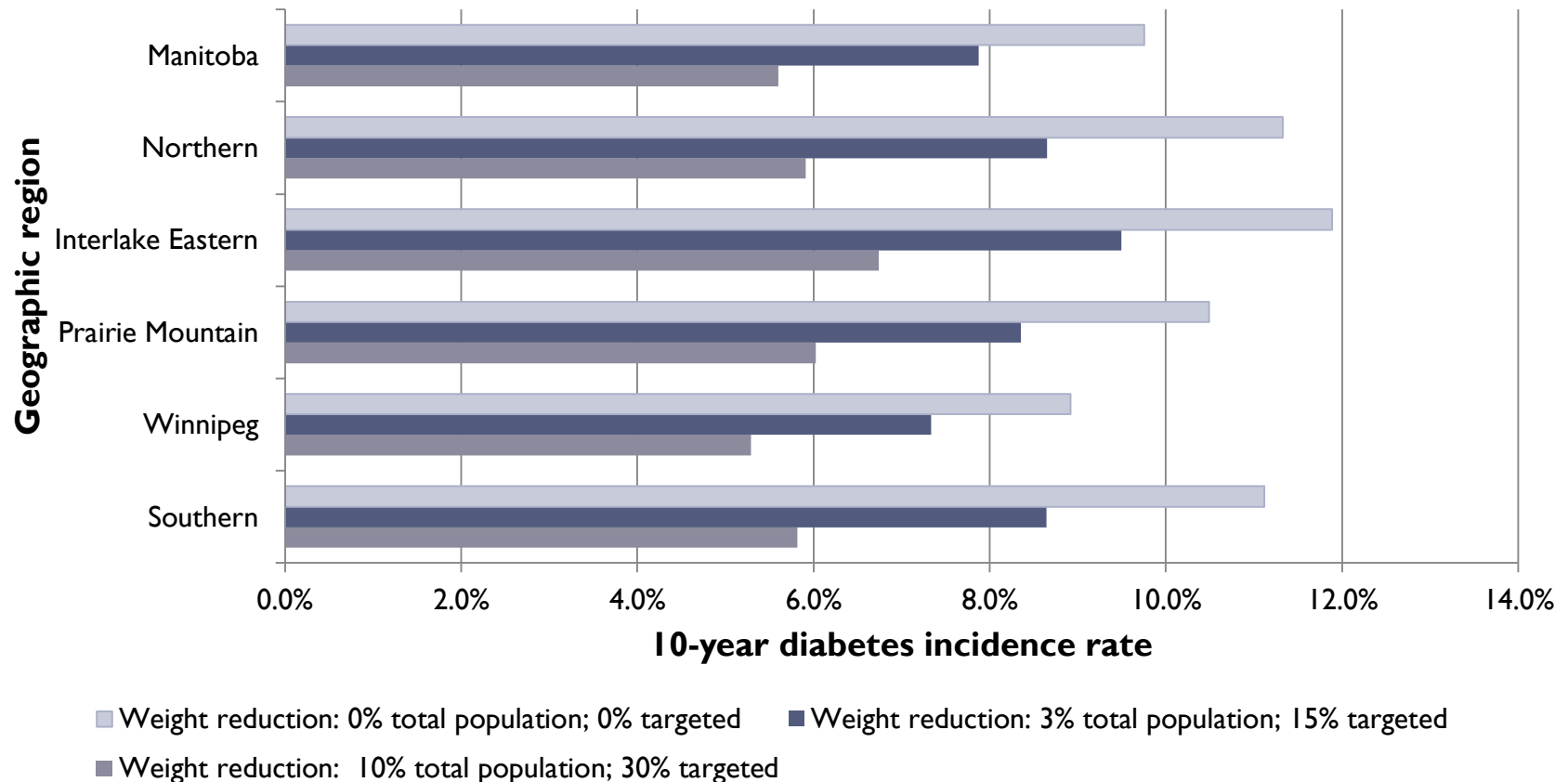
# Estimate the Impact of Diabetes Prevention Approaches

Expected change in ten-year diabetes incidence rate in Manitoba as a result of weight loss among obese individuals (2011/12–2021/22)



# Estimate the Impact of Diabetes Prevention Approaches

Estimated ten-year diabetes incidence rate in Manitoba as a result of a combination of population-level weight loss and targeted weight loss among obese individuals (2011/12–2021/22)



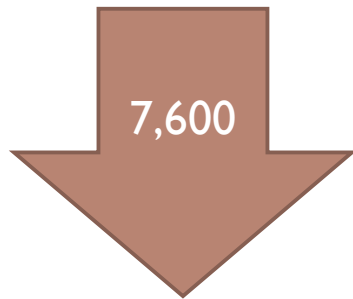
# Making the Case for Funding Prevention

## Example from Peel Public Health, Ontario

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**A lifestyle program implemented to all individuals at high risk (i.e. 10% of the population)**

Reduction in future diabetes cases within 10 years



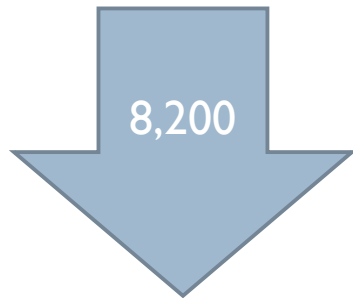
and

Total health care savings



**A combination of population wide policies that result in a 2.5% reduction in weight**

Reduction in future diabetes cases within 10 years



and

Total health care savings





# Interpretive Cautions

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- ▶ **Diabetes definition:** DPoRT estimates the number of individuals who will develop physician-diagnosed diabetes.
- ▶ DPoRT does not consider individuals with diabetes not recognized by themselves or their doctor.
- ▶ The estimates reflect cases identified in the National Diabetes Surveillance System (NDSS). There may be provincial differences in how people with diabetes are included in the NDSS.

# Interpretive Cautions

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**Study population:** DPoRT estimates represent community-dwelling Canadians living in the 10 provinces during data collection year of CCHS.

- ▶ DPoRT estimates do not represent:
  - ▶ residents of First Nation reserves,
  - ▶ people who live in institutions such as nursing homes,
  - ▶ full-time members of the Canadian Forces,
  - ▶ residents of certain remote regions, and
  - ▶ people who may immigrate to Canada during ten-year period following data collection year of CCHS

# Self-Reported Height/Weight & Ethnicity

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## Height/weight:

- ▶ Shields *et al.* (2008) examined agreement between self-reported and measured BMI in a sub-sample of the CCHS population
- ▶ DPoRT's discrimination and calibration would be minimally affected at these levels

## Ethnicity:

- ▶ Influence of ethnic groups was tested by examining modified versions of DPoRT models
- ▶ All models produced similar C statistics (differing only at the 0.01 place)

# Summary

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- ▶ DPoRT was successfully validated in two external validation cohorts and demonstrated good discrimination and calibration
- ▶ DPoRT allows us to empirically estimate the future risk and number of new cases of Type 2 diabetes in a population
- ▶ DPoRT can quantify the impact that changes in baseline risk factors will have on future diabetes incidence