

BC Chronic Disease Capstone Proposal

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Contents

1	Executive Summary	2
2	Introduction	3
2.1	Dashboard	3
2.2	Temporal Modeling	3
3	Data Science Techniques	3
3.1	Dashboard	3
3.2	Temporal Modeling	4
4	Timeline	4
5	Conclusion	4

1 Executive Summary

The Chronic Disease Registry is a data product that captures information about the rate of new and persistent cases of chronic diseases in each CHSA (community health service area), as well as demographic information such as sex and age standardized rates. The Registry has 3 different types of rates available. Incidence Rate is the rate at which new cases occur in a specified population during a specified time period; Lifetime Prevalence is the proportion of individuals who have had the condition for at least part of their lives, and Active Healthcare Contact Prevalence are the cases for which a patient seeks healthcare services for relapsing - remitting conditions. In this project we aim to create an interactive dashboard that will allow users of all technical expertise to explore and visualize spatiotemporal information of the diseases rates in the data, and to develop an analysis pipeline that will describe the spatiotemporal trends in the data. This proposal will outline the approach we will take to tackle this problem and achieve the project goals.

2 Introduction

Millions of people in BC live with a chronic disease, so it's important to understand and interpret the distribution of disease prevalence throughout the province for a variety of reasons. We may want to know how to best allocate healthcare resources, or to identify if a specific region is experiencing rapid growth of a disease.

2.1 Dashboard

To visualize the spatial and temporal trends of disease rates in the province, we will build an interactive dashboard that will allow users to compare the rates of various diseases in one specific health region over time, as well as to compare how the rates of one disease has varied across several health regions over time. The users will be able to compare between multiple Health Authorities (HA) or multiple Community Health Service Areas (CHSA). We also plan to create an information page in the dashboard containing definitions and descriptions of variables and diseases to increase usability for less technical users. Lastly, we will have a page displaying the data table with filters applied, with the option to download the data as a report. A sketch of the proposed dashboard design is shown below in Figure 1.

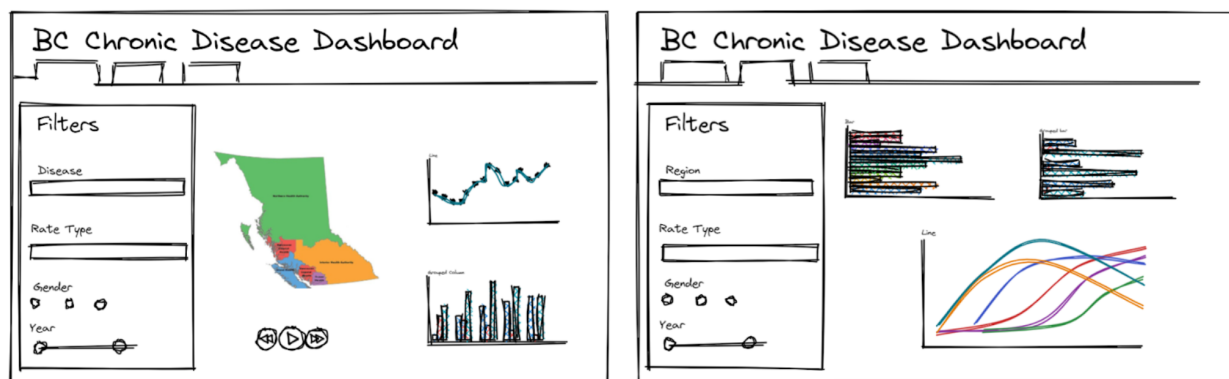


Figure 1: Proposed Dashboard Design Layout

2.2 Temporal Modeling

The Office of BC Provincial Health Officer curates data of ongoing surveillance of 25 Chronic Diseases throughout the whole region (British Columbia province). Currently the dataset has data on Incidence Rate, Prevalence Rates, Geographical and administrative geo-tiers and age-groups. To observe the trend of changes throughout the years of such surveillance data, temporal analysis is the standard choice. We will explore various methods as per nature of data, established evidences, stakeholder discussions and variables of interest. Eventually, the outputs will be incorporated with the primary R Shiny app.

3 Data Science Techniques

Several different data science tools and techniques will be used throughout the project to accomplish the project deliverables. Some tools are familiar to the team, while others tools we will need to learn. The tools and techniques for each aspect of the project are described in this section.

3.1 Dashboard

We will be using R Shiny to build the framework of the interactive dashboard, and we will be utilizing the leaflet and ggplot packages to assist in creating visually appealing graphs and maps. For data wrangling we will use the tidyverse set of packages to tidy the data in preparation for plotting.

3.2 Temporal Modeling

We will use a Bayesian temporal smoothing model for the inferential analysis of chronic disease incidence and prevalence over time. A Bayesian approach was selected in order to integrate prior knowledge about the disease rates and to generate 95% credible intervals from the posterior distribution. Temporal smoothing is appropriate because we are assuming that the disease incidence and prevalence within a given region will be correlated over time. We propose to use the INLA package for Bayesian analysis for computational efficiency compared to traditional MCMC approaches.

For each disease, we will compare 3 temporal models for each of the 3 disease rates: an autoregressive model with order 1 (AR(1)), a random walk model with order 1 (RW(1)), and a random walk with order 2 (RW(2)). To select our final model, we will use the Deviance Information Criterion (DIC) and the Widely Applicable Information Criterion (WAIC), where a smaller DIC/WAIC indicates a better fit. After the best model is determined for that disease and metric, we will fit that model on all CHSA regions and generate plots with both observed and smoothed disease incidence/prevalence rates for the Shiny app.

4 Timeline

We will form two sub-teams to work in parallel to develop the two final deliverables. The milestones descriptions and respective target week of completion for each sub-team are showed below in Figure 2.

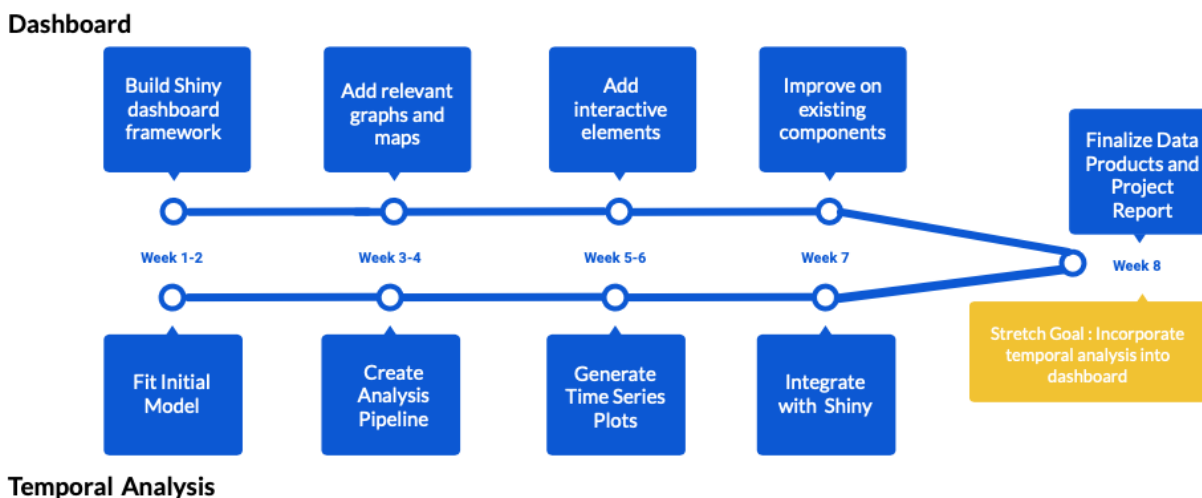


Figure 2: Timeline of Capstone project milestones

5 Conclusion

Each year, the Office of British Columbia Provincial Health Officer tracks persistent and new cases of 25 chronic diseases throughout the whole region. This document has proposed methods to accommodate the possible inquiries about the prevalence of chronic conditions and their spatiotemporal trends from health professionals and non-experts. The proposed methods include: (1) an interactive and informative dashboard using R Shiny for users to visualize the distribution of diseases with filters for measure, disease, sex, and health authority regions, and (2) a data analysis pipeline using the Bayesian temporal smoothing model to infer and display temporal trends of diseases in the dashboard. Two sub-teams will work in parallel to develop the final products for each method in 8 weeks. The project will include a formal presentation on June 16 and a final report on June 29.