# Alberta Mortality 2000 to 2022\*

My subtitle if needed

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First sentence. Second sentence. Third sentence. Fourth sentence.

#### 1 Introduction

# 2 Data

Our data is from Alberta Government opendata, the dataset includes a ranking of the 30 most common causes of death each year in Alberta starts from 2000 to 2022, by ranking and total number of deaths (Government 2015).

#### 3 Model

#### 3.1 Model set-up

Two models generated, one follows poisson distribution and another follows negative binomial distribution. We are interested in how total number of deaths differs by different causes of deaths

Define  $y_i$  as the total number of deaths for the i-th observation. Then  $\beta_0$  is the expected log count of total deaths when none of the causes in the model are present. It's the starting point of the model's prediction.

<sup>\*</sup>Code and data are available at: https://github.com/peachvegetable/Alberta-mortality

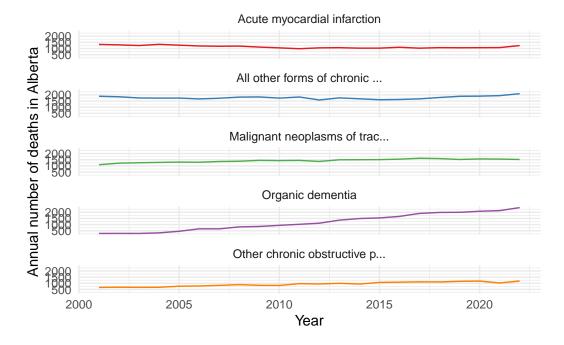


Figure 1: Top 5 causes of deaths from 2000 to 2022, for Alberta, Canada

$$\begin{aligned} y_i | \lambda_i &\sim \operatorname{Poisson}(\lambda_i) \\ \log(\lambda_i) &= \beta_0 + \beta_1 \times \operatorname{cause}_i \\ \beta_0 &\sim \operatorname{Normal}(0, 2.5) \\ \beta_1 &\sim \operatorname{Normal}(0, 2.5) \end{aligned}$$

Define  $y_i$  as the total deaths for the i-th observation.  $\theta$  is the additional parameter to model overdispersion.  $\mu_i$  is the mean of the Negative Binomial distribution for the i-th observation,  $\beta_0$  is the intercept, and  $\beta_1$  represents the effect of each cause of death.

$$\begin{split} y_i | \lambda_i, \theta &\sim \text{NegativeBinomial}(\mu_i, \theta) \\ \log(\mu_i) &= \beta_0 + \beta_1 \times \text{cause}_i \\ \beta_0 &\sim \text{Normal}(0, 2.5) \\ \beta_1 &\sim \text{Normal}(0, 2.5) \\ \theta &\sim \text{SomePrior}(.) \end{split}$$

We run the model in R (R Core Team 2023) using the rstanarm package of (Goodrich et al. 2022). We use the default priors from rstanarm.

Table 1: Modeling the cause deaths in Alberta, 2000 - 2022

	Poisson	Negative binomial
(Intercept)	7.037	7.037
		(0.070)
causeAll other forms of chronic	0.446	0.448
		(0.101)
$cause {\bf Malignant\ neoplasms\ of\ trac}$	0.223	0.226
		(0.102)
causeOrganic dementia	0.046	0.048
		(0.101)
causeOther chronic obstructive p	-0.206	-0.202
		(0.101)
Num.Obs.	110	110
Log.Lik.	-5718.182	-810.965
ELPD	-5906.6	-815.4
ELPD s.e.	1211.7	10.5
LOOIC	11813.2	1630.9
LOOIC s.e.	2423.5	21.1
WAIC	11965.6	1630.8
RMSE	325.38	325.38

Table 2: ?(caption)

elpd\_diff se\_diff neg\_binomial\_model 0.0 0.0 poisson\_model -5091.2 1201.6

We calculate the total number of deaths by the formula

total deaths = 
$$e^{\beta_0 + \sum_{i=0}^5 \beta_i X_i}$$
 (1)

Where  $X_1$ ,  $X_2$ ,  $X_3$ ,  $X_4$ , and  $X_5$  represent  $X_{\rm Acute\ myocardial\ infarction}$ ,  $X_{\rm All\ other\ forms\ of\ chronic}$ ,  $X_{\rm Malignant\ neoplasms}$ ,  $X_{\rm Organic\ dementia}$ , and  $X_{\rm Other\ chronic\ respectively}$ .

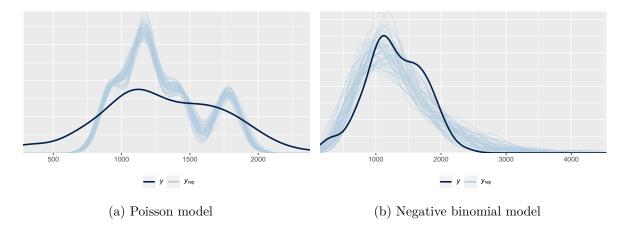


Figure 2: Comparing Poisson and negative binomial models

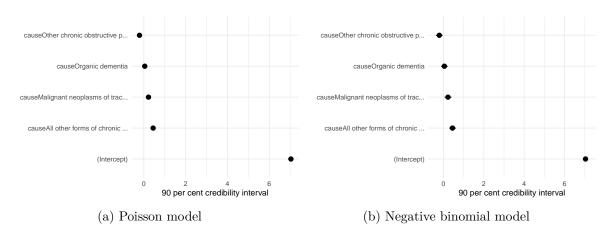


Figure 3: 90% Confidence Interval

# 3.2 Model summary

#### 4 Results

# 5 Discussion

#### 5.1 First discussion point

If my paper were 10 pages, then should be be at least 2.5 pages. The discussion is a chance to show off what you know and what you learnt from all this.

#### 5.2 Second discussion point

#### 5.3 Third discussion point

### 5.4 Weaknesses and next steps

Weaknesses and next steps should also be included.

# **Appendix**

- A Additional data details
- **B** Model details
- **B.1** Posterior predictive check
- **B.2 Diagnostics**

# References

- Goodrich, Ben, Jonah Gabry, Imad Ali, and Sam Brilleman. 2022. "Rstanarm: Bayesian Applied Regression Modeling via Stan." https://mc-stan.org/rstanarm/.
- $Government, \ Alberta. \ 2015. \ Leading \ Causes \ of \ Death. \ https://open.alberta.ca/dataset/03339dc5-fb51-4552-97c7-853688fc428d/resource/1a10c821-7399-4d0f-95fb-f96728d01fae/download/deaths-leading-causes.xlsx.$
- R Core Team. 2023. R: A Language and Environment for Statistical Computing. Vienna, Austria: R Foundation for Statistical Computing. https://www.R-project.org/.