My title*

My subtitle if needed

Rahma Binth Mohammad

August 8, 2024

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

1 Introduction

Consumption is the biggest indicator of a person's needs and desires. Our interest is to inquire into whether tracking consumption, along with income and prices of items through inflation, will tell us a better story about a household's financial behaviour and broader economic trends. By assessing these variables, we can uncover how inflation influences spending habits and whether an increase or decrease in income can mitigate or exacerbate the effects of rising or falling prices. The relationship between these variables can provide information in regards improving economic stability and effective social/governmental assistance to support households in managing their spending in the face of inflationary pressures. Policymakers can use this information to employ strategies for the wellbeing of the state.

Recently, inflation has become serious problem. It diminishes purchasing power, which is a problem at the individual level but also state level. On an individual level, it affects costs of living, impeding the ability to live luxuriously. On a state level, it can lead to increased costs for goods and services, reducing the competitiveness of a state's exports and making it harder for businesses to thrive on a domestic and global market. Thus, it is crucial states manage inflation as it will strength consumer and investor confidence, leading to economic stability and growth.

This study focuses specifically on the total expenditure, household income, and the inflationary changes over a period of several years.

We used a simple linear regression model and a multiple linear regression for our analysis. For the simple linear model, we used the the change in prices over several years against the average household spending during these years. For the multiple regression model, we used the average

---

*Code and data are available at: https://github.com/rahmabinth/consumer-behaviour

1

household income, along with the change in the prices to predict household spending. Our estimand is the

Based on our model and analysis, we found that

This paper was designed to communicate our model and analysis in a structured manner. Section 2 will provide the datasets used for this analysis. Section 3 will explore the simple and multiple linear regression models used for this analysis, along with the justification for the use of such models. Section 4 will provide the results from that analysis. A discussion of the results and the weaknesses and biases of this type of analysis will be provided in Section 5.

2 Data

2.1 Methodology

The language and environment used for this analysis is R (R Core Team 2023), alongside the tidyverse (Wickham et al. (2019)), rstanarm (Goodrich et al. (2022)), cansim (von Bergmann and Shkolnik 2021), ggplot2 (Wickham 2016), scales (Wickham, Pedersen, and Seidel 2023), modelsummary (Arel-Bundock 2022), and marginaleffects (Arel-Bundock 2024) packages.

We downloaded the data directly from Statistics Canada using the cansim package. We cleaned it, identified missing values and wrote csv files to make it easily accessible using R studio.

2.2 Sources and Variables

A good indicator of inflation is the Consumer Price Index (CPI), as it measures the change in price using a fixed basket of goods and services. Statistics Canada is a government agency that collects information on Canada's economy (cite). Thus, it has the data that provides the CPI for numerous years. Statistics Canada also provides us with other datasets such as the average household income and the average household expenditures, in a year for all the provinces. We have chosen to look at Ontario solely, as we hold firmly Ontario behaves very differently due to its high population in relation to the other provinces. The data for Canada as a whole is complicated as there are many missing values for some provinces. Our analysis and the data provided by Statistics Canada makes it possible to reproduce our model for different provinces using the github repository. While there are other datasets available, these three datasets are provided by the same source and has data for provinces across Canada. The expenditure and cpi data also provides the information for three additional specific categories, food, shelter, and recreation, making it easier to isolate and answer our research question. Section 2.2 will illustrate the visualised data.

This paper's objective is to address the following questions using data collected from a government census: (1) Does changes in prices affect expenditure? (2) Does changed in prices affect expenditure as income changes?

2.2.1 Spending Data

`spending.csv`

This dataset contains the average total expenditures of the Ontario resident for the years 2010 to 2021, excluding years 2018 and 2020. It also contains the breakdown of the expenditures, food purchased (from restaurant and stores), shelter (rented and owned), utilities such as water, fuel, and electricity for the principal accommodation, clothing, transportation (private and public), healthcare, recreational goods and services.

From this dataset, we chose to select the values for all food expenditures, all shelter expenditures, recreation expenditures, and total expenditures. This choice was made because these are principal expenditures for everyone. Recreation was added to this mix, to identify how consumers behave when inflation is high. Figure 5 indicates this relationship. In Figure 1, we observe a steady increase in recreation expenditures, while food, shelter, and overall expenditures show a more rapid growth. Note for simplicity, we did not include the other categories used for total expenditures. For all four of the categories, the mean and median values were close to each other, indicating the data is not skewed and did not have many outliers. However, the standard deviation for shelter and total expenditures were higher which indicates more money was spent and thus, more variation over the years. The only variable we use for our models from this dataset is `total expenditure`. However, the model can be used for the other categories. The model is made to be easily reproducible and can be found in the github link provided on page 1.
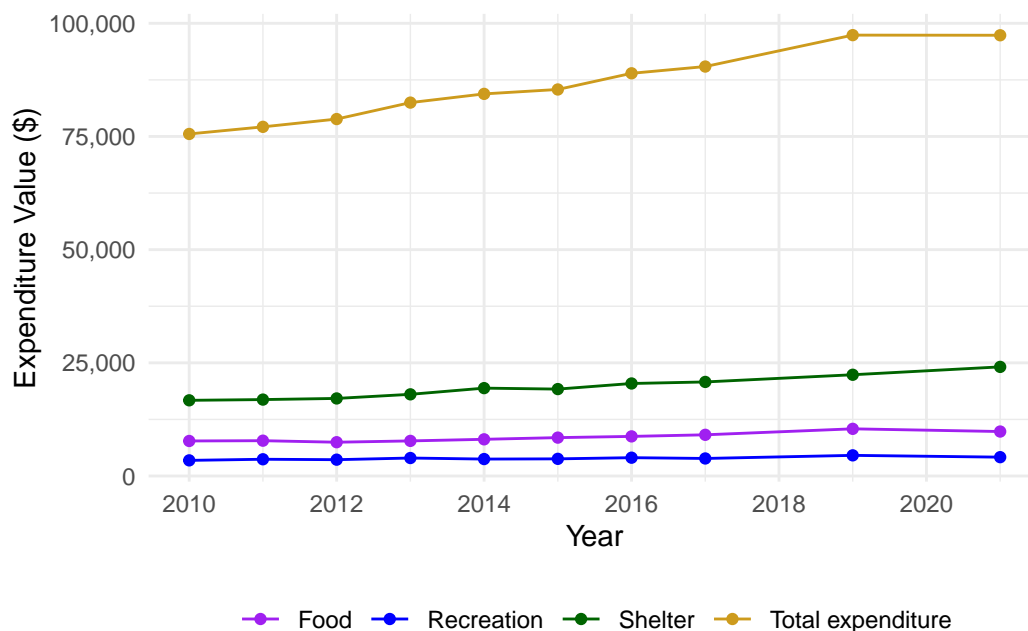

Figure 1: Household spending per year

2.2.2 Inflation Data

`cpi.csv`

The dataset provides CPI for every year. It is not seasonally adjusted for changes as it is an annual average. Thus, it is actual price paid by consumers and not adjusted for seasonal fluctuations. The base year for the CPI calculation is 2002 where CPI is 100. Thus, if our CPI is 120 for a year, it indicates a 20% increase since 2002. This value is useful as an indicator for inflation, as it provides the increase and comparison between years, rather than a absolute price values.

From this dataset, we chose to look at the annual average to keep it consistent with the annual average of expenditures. The data provides cpi for specific categories, as well as all-items. We chose to select cpi for all-items, food, shelter, and recreations, for comparison with the household spending. The prices for food and shelter increased almost 35% in the 10 years illustrated in Figure 2. However, the prices for recreation saw a decrease in 2020 and 2021, which indicates a relation to COVID-19. Our models will use the 'all-items' CPI to ensure consistency.
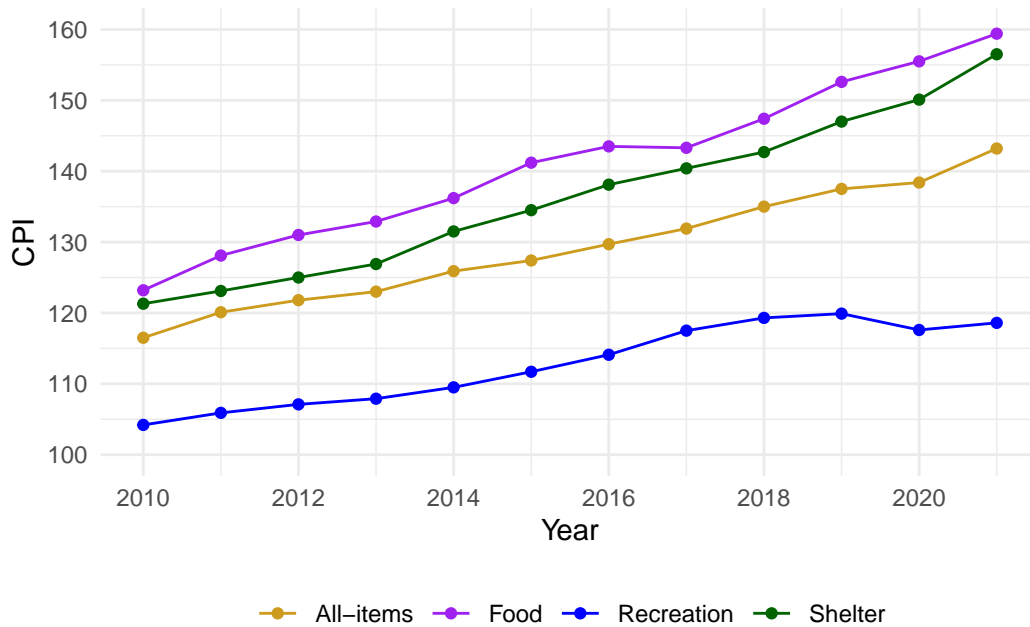

Figure 2: Consumer Price Index (CPI) per Year

2.2.3 Income Data

`income.csv`

Throughout the province of Ontario, there exits people in many different financial situations. Statistics Canada provides us with this dataset that contains the average number of people in an income category. For example, "persons with income under $5000", "persons with income of $5000 and over", "persons with income of $10 000 and over." Note for 2020, COVID-19 government income support and benefits are included in income estimates.

From this dataset, we chose to isolate the values for the category "all persons with income". This indicate all tax fillers and dependants that earn income from all sources. This was done to keep our analysis consistent, as we could not separate who was spending. Our model will use the all persons with income variable as well. Figure 3 shows a significant increase in the number of income earners. This could be due to a population increase, specifically an increase in tax fillers and tax dependants.
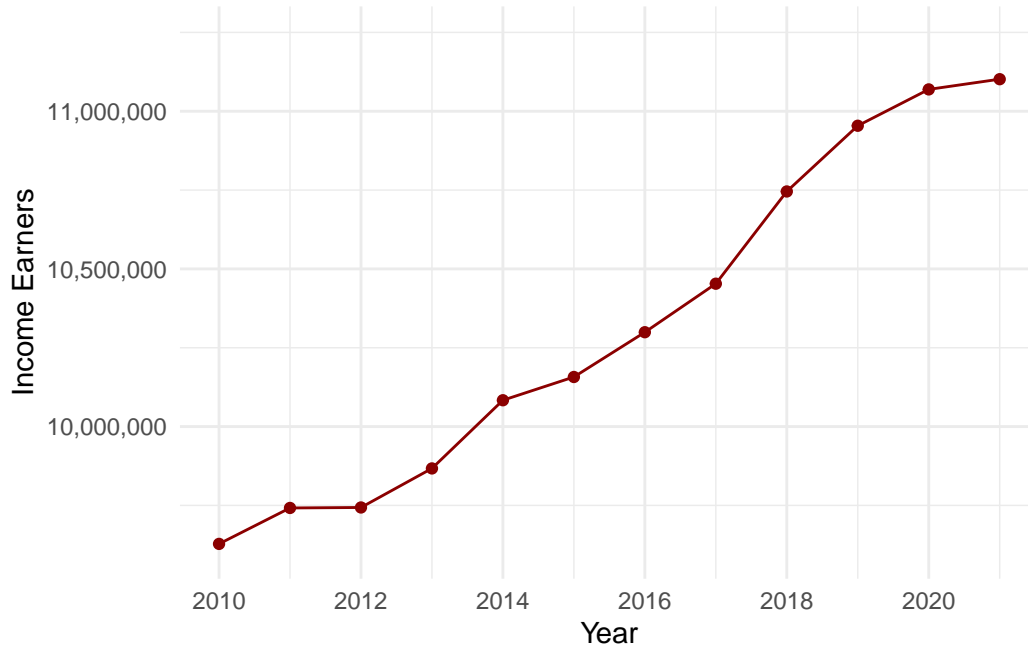

Figure 3: Number of Income Earners per Year

For our analysis, we have to use all three dataset values and thus, we chose to use data for the years 2010 to 2021, excluding 2018 and 2020, given the data was only available in all given datasets for these years. However our model does predict the behaviour in 2018 and 2020, using the other years' data. We also created numerous csv files that isolated the specific data. Section A thoroughly explains the structure of the csv files.

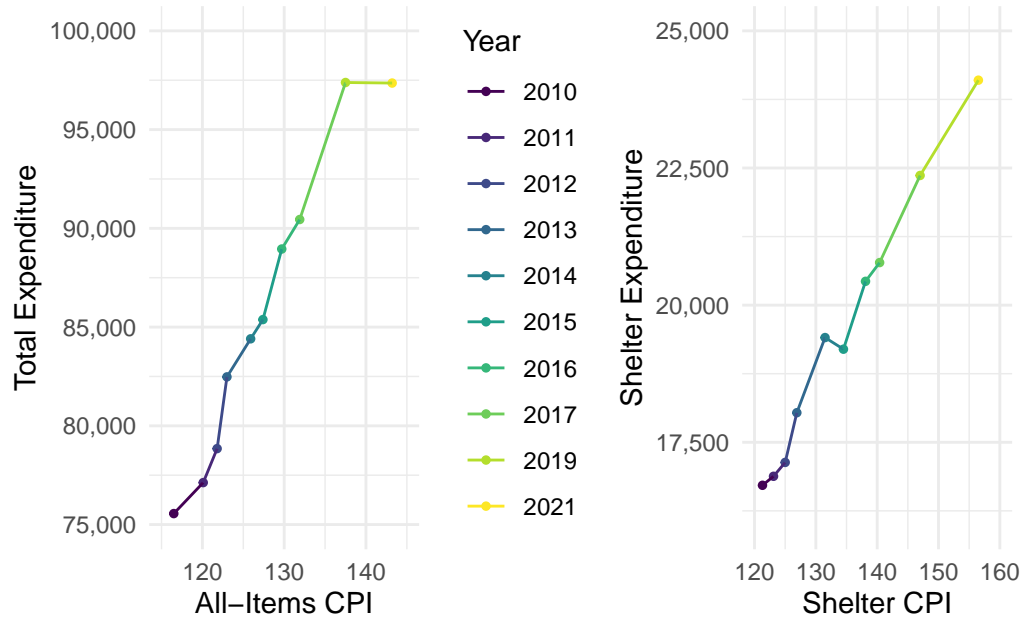## 2.3 Measurement

A discussion about measurement.

Figure 4: All-items CPI and total expenditure over the years (on the left), shelter CPI and shelter expenditure over the years (on the right)
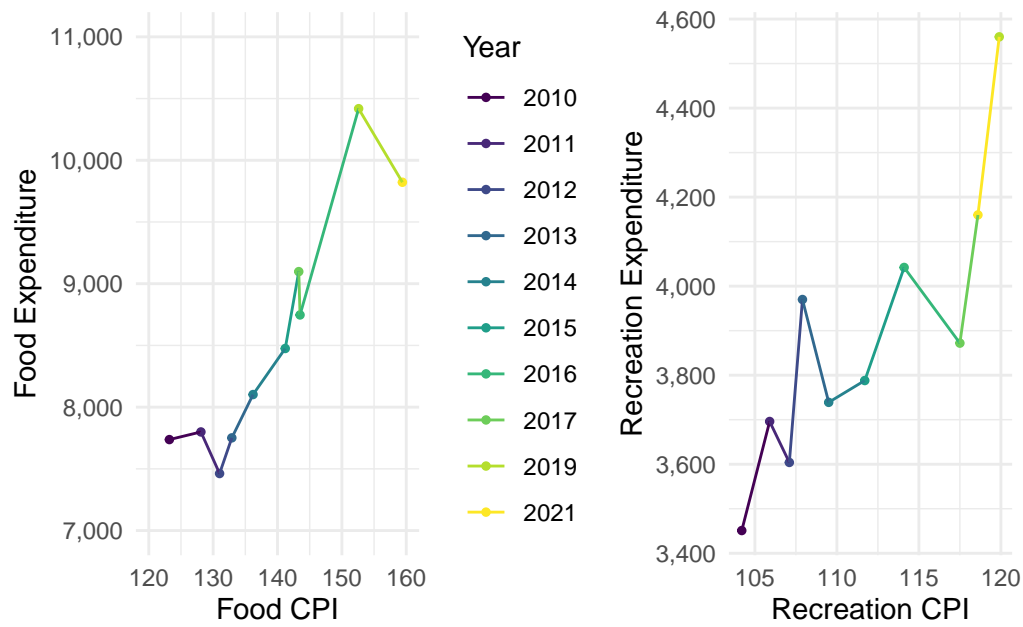

Figure 5: Food CPI and food expenditure over the years (on the left), recreation CPI and recreation expenditure over the years (on the right)

3 Model

The goal of our model is to forecasting the total expenditure in a household in a year, based solely on a the Consumer Price Index, and the number of people in Ontario earning an income. 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`.

Here we briefly describe the Bayesian analysis model used to investigate... Background details and diagnostics are included in Appendix C.

3.1 Model set-up

3.1.1 Simple linear regression

The model we employ is:

$$y_i | \mu_i, \sigma \sim \text{Normal}(\mu_i, \sigma) \tag{1}$$
$$\mu_i = \beta_0 + \beta_1 \times \text{cpi}_i + \epsilon \tag{2}$$
$$\beta_0 \sim \text{Normal}(0, 2.5) \tag{3}$$
$$\beta_1 \sim \text{Normal}(0, 2.5) \tag{4}$$
$$\sigma \sim \text{Exponential}(1) \tag{5}$$

Where:

- y_i is the continuous outcome variable, representing the total expenditure in Canadian dollars in year i.
- cpi_i is the predictor variable, representing the annual Consumer Price Index (CPI) for year i

3.1.2 Multiple linear regression

The model we employ is:

$$y_i|\mu_i, \sigma \sim \text{Normal}(\mu_i, \sigma) \tag{6}$$
$$\mu_i = \beta_0 + \beta_1 \times \text{cpi}_i + \beta_2 \text{income}_i + \epsilon \tag{7}$$
$$\beta_0 \sim \text{Normal}(0, 2.5) \tag{8}$$
$$\beta_1 \sim \text{Normal}(0, 2.5) \tag{9}$$
$$\beta_2 \sim \text{Normal}(0, 2.5) \tag{10}$$
$$\sigma \sim \text{Exponential}(1) \tag{11}$$

Where:

- y_i is the continuous outcome variable, representing the total expenditure in Canadian dollars in year i.
- cpi_i is the predictor variable, representing the annual Consumer Price Index (CPI) for year i
- $income_i$ is the number of people earning an income in year i.

3.1.3 Model justification

We expect a positive relationship between the size of the wings and time spent aloft. In particular...

We can use maths by including latex between dollar signs, for instance θ.

4 Results

Our results are summarized in Table 1.

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.
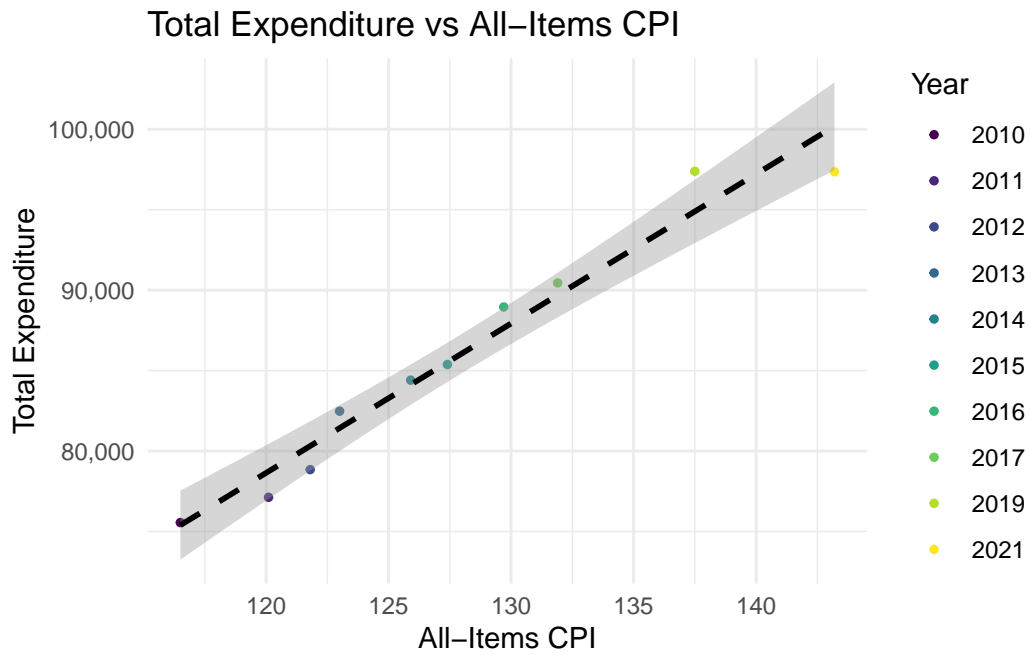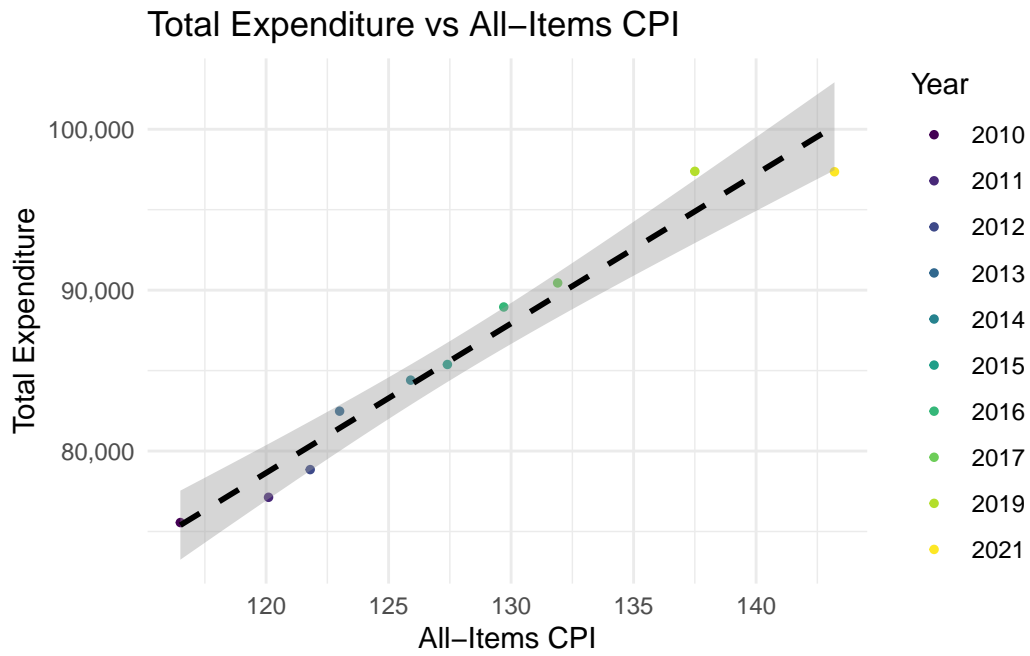
Figure 6: Linear model

Figure 7: Linear model

Table 1: Explanatory models of

	First model	Second model
(Intercept)	−32 899.89	−59 979.55
	(9842.07)	(18 660.19)
cpi	928.77	225.52
	(76.80)	(437.28)
income		0.01
		(0.01)
Num.Obs.	10	10
R2	0.944	0.955
R2 Adj.	0.920	0.932
Log.Lik.	−88.550	−87.273
ELPD	−92.0	−91.1
ELPD s.e.	3.1	2.3
LOOIC	184.1	182.1
LOOIC s.e.	6.1	4.6
WAIC	182.4	180.5
RMSE	1501.29	1326.89

5.2 Second discussion point

5.3 Third discussion point

5.4 Weaknesses and next steps

Weaknesses and next steps should also be included.

A Appendix

B Additional data details

C Model details

Posterior predictive check

Arel-Bundock, Vincent. 2022. "modelsummary: Data and Model Summaries in R." *Journal of Statistical Software* 103 (1): 1–23. https://doi.org/10.18637/jss.v103.i01.

———. 2024. *Marginaleffects: Predictions, Comparisons, Slopes, Marginal Means, and Hypothesis Tests.* https://CRAN.R-project.org/package=marginaleffects.

Goodrich, Ben, Jonah Gabry, Imad Ali, and Sam Brilleman. 2022. "Rstanarm: Bayesian Applied Regression Modeling via Stan." https://mc-stan.org/rstanarm/.

R Core Team. 2023. *R: A Language and Environment for Statistical Computing.* Vienna, Austria: R Foundation for Statistical Computing. https://www.R-project.org/.

von Bergmann, Jens, and Dmitry Shkolnik. 2021. *Cansim: Accessing Statistics Canada Data Table and Vectors.* https://CRAN.R-project.org/package=cansim.

Wickham, Hadley. 2016. *Ggplot2: Elegant Graphics for Data Analysis.* Springer-Verlag New York. https://ggplot2.tidyverse.org.

Wickham, Hadley, Mara Averick, Jennifer Bryan, Winston Chang, Lucy D'Agostino McGowan, Romain François, Garrett Grolemund, et al. 2019. "Welcome to the tidyverse." *Journal of Open Source Software* 4 (43): 1686. https://doi.org/10.21105/joss.01686.

Wickham, Hadley, Thomas Lin Pedersen, and Dana Seidel. 2023. *Scales: Scale Functions for Visualization.* https://CRAN.R-project.org/package=scales.