# My title\*

# My subtitle if needed

First author

Another author

April 13, 2024

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

## 1 Introduction

You can and should cross-reference sections and sub-sections. We use R Core Team (2023) and Wickham et al. (2019).

The remainder of this paper is structured as follows. Section ??....

## 2 Data

Table 1: Sample of Cleaned Elections Data

Ward ID	Eligible Voter Turnout (%)
1	24
2	30
3	33
4	38
5	28

<sup>\*</sup>Code and data are available at: LINK.

Table 2: Sample of Cleaned Toronto Ward Profile Data

Ward ID	Uneducated Population (%)	Unemployment Rate (%)	Income
1	18.997568	16.5	95200
2	11.053754	12.8	146600
3	9.269583	11.8	127200
4	9.072244	12.9	127200
5	21.750594	16.4	88700

Table 3: Sample of Combined Ward Election, Income, Employment, and Education Data

Ward ID	Ward Name	Uneducated Population (%)	1 0		Eligible Voter Turnout (%)
1	Etobicoke North	18.997568	16.5	95200	24
2	Etobicoke	11.053754	12.8	146600	30
	Centre				
3	Etobicoke-	9.269583	11.8	127200	33
	Lakeshore				
4	Parkdale-High	9.072244	12.9	127200	38
	Park				
5	York	21.750594	16.4	88700	28
	South-Weston				
6	York Centre	13.692888	14.1	107500	25
7	Humber	23.080036	17.8	85700	22
	River-Black				
	Creek				
8	Eglinton-	10.490333	12.2	176400	32
	Lawrence				
9	Davenport	18.012986	13.1	107300	31
10	Spadina-Fort	4.841211	9.8	118200	24
	York				

Table 4: Summary Statistics

			Standard.		
Variable	Mean	Median	Deviation	$\operatorname{Min}$	Max
Income	120096.00	107300.00	33980.64	85700.00	224800.00
Voter Turnout (%)	29.48	29.00	4.22	22.00	38.00
Uneducated	12.43	11.95	4.93	4.84	23.08
Population (%)					

Table 4: Summary Statistics

Variable Mean Median		Standard. Deviation Min		Max	
Unemployment Rate (%)	14.13	14.10	2.11	9.80	17.80

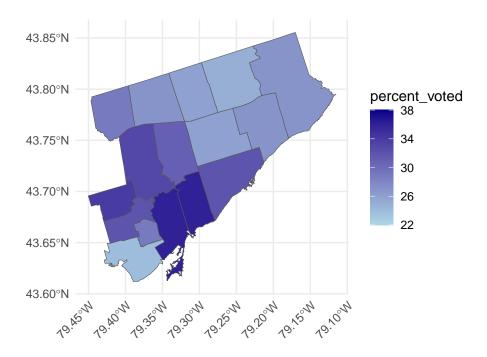


Figure 2: Map of Toronto highlighting the voter turnout across wards

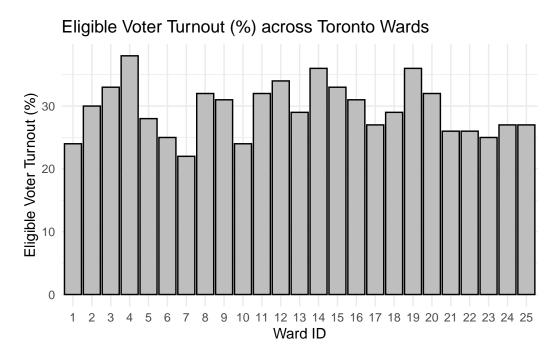


Figure 1: Voter Turnout (%) across Toronto Wards

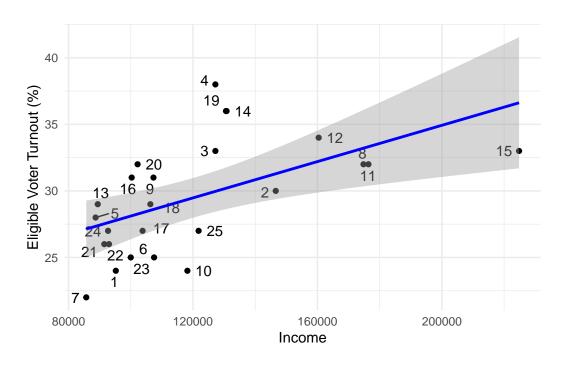


Figure 3: Correlation between Eligibile Voter Turnout and Ward's Income

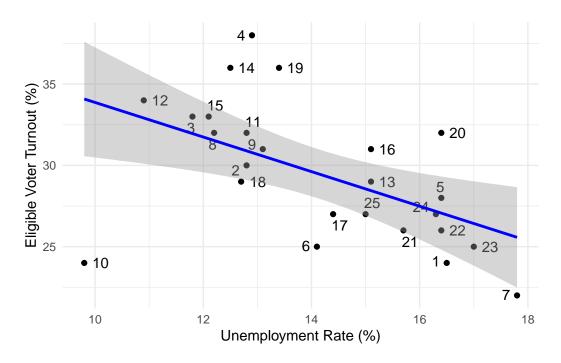


Figure 4: Correlation between Eligibile Voter Turnout and Ward's Unemployment Rate

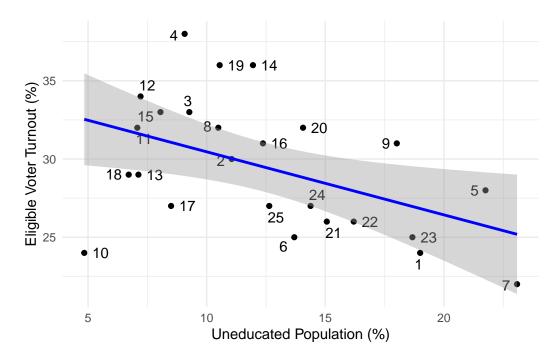


Figure 5: Correlation between Eligibile Voter Turnout and Ward's Level of Education

#### 3 Model

The goal of our modelling strategy is twofold. Firstly,...

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

#### 3.1 Model set-up

Define  $y_i$  as the number of seconds that the plane remained a loft. Then  $\beta_i$  is the wing length, both measured in millimeters.

$$y_i|\mu_i, \sigma \sim \text{Normal}(\mu_i, \sigma)$$
 (1)

$$\mu_i = \alpha + \beta_i + \gamma_i \tag{2}$$

$$\alpha \sim \text{Normal}(0, 2.5)$$
 (3)

$$\beta \sim \text{Normal}(0, 2.5)$$
 (4)

$$\gamma \sim \text{Normal}(0, 2.5)$$
 (5)

$$\sigma \sim \text{Exponential}(1)$$
 (6)

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.

#### 3.1.1 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  $\theta$ .

#### 4 Results

Our results are summarized in Table ??.

Table 5: Explanatory models of flight time based on wing width and wing length

	First model
(Intercept)	3.46
	(0.49)
$percent\_uneducated$	0.00
	(0.01)
income	0.00
	(0.00)
$unemployment\_rate$	-0.01
	(0.03)
Num.Obs.	25
Log.Lik.	-70.304
ELPD	-73.4
ELPD s.e.	2.1
LOOIC	146.9
LOOIC s.e.	4.1
WAIC	146.5
RMSE	3.37

#### 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**

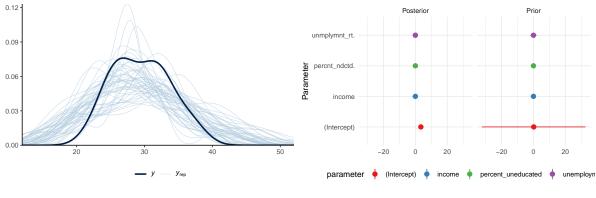
# A Additional data details

#### **B** Model details

#### **B.1** Posterior predictive check

In Figure ?? we implement a posterior predictive check. This shows...

In Figure ?? we compare the posterior with the prior. This shows...



- (a) Posterior prediction check
- (b) Comparing the posterior with the prior

Figure 6: Examining how the model fits, and is affected by, the data

#### **B.2 Diagnostics**

Figure ?? is a trace plot. It shows... This suggests...

Figure ?? is a Rhat plot. It shows... This suggests...

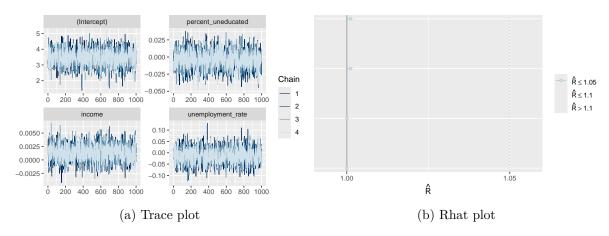


Figure 7: Checking the convergence of the MCMC algorithm

## References

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/.

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.