

# Predicting probability of life satisfaction level among adult Canadians with multiple factors in life

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## Abstract

This report analyzes how mature Canadians feel about their life, provided certain factors in life. Deriving data from the 2017 General Social Survey, this analysis utilizes a dependent variable that has two categorical levels- whether the respondent individual answered feeling generally happy, or the respondent was overall not content with life. Income of the respondent illustrated to play a role in influencing the life rating, whereas other factors in life did not portray a significance in affecting the rating. This was a meaningful discovery as higher income was found to be associated with higher life happiness level in general.

## Introduction

Here is where you should give insight into the setting and introduce the goal of the analysis. Here you can introduce ideas and basic concepts regarding the study setting and the potential model. Again, this is the introduction, so you should be explaining the importance of the work that is ahead and hopefully build some suspense for the reader. You can also highlight what will be included in the subsequent sections.

The goal of this analysis is to discover a relationship between the outcome variable of contentedness of life and the auxiliary variables of factors in life by utilizing a logistic regression model. In this analysis, the study variable is feelings about life, and five auxiliary variables were selected to test if they indeed have a relationship with the outcome variable. Feelings about life were measured as generally happy if the rating was five or higher out of 10, while they were grouped as overall not content if the rating was lower than 5. The chosen dependent variables are age at first birth, current age of the respondent, income of the respondent, place of birth whether it is within or outside of Canada, and future intention of having any children. While there were numerous variables in the data that were available to select, these five specific variables were intuitively picked to check the hypothesis that these will have an association with life rating. The importance of the work is that through the logistic regression model, an analysis can be built about how adult individuals in Canada feel about their life, considering the input variables. More precisely, the probability of an adult Canadian having a contented life can be calculated. Ultimately, the main goal is to observe if any of these life factors affect life rating, and if so, how large the influence is. In the subsequent sections, the model section is covered and explained using tables and the results section through a graph, while weaknesses and next steps are also discussed.

## Data

Introduce the data, explain why it was selected. Make sure to comment on important features and highlight any potential drawbacks to the data.

The selected output variable is feelings about life, and the auxiliary variables are age of the respondent, age at first birth, place of birth (whether it was in Canada or not), income of the respondent, and intention of having future children. We chose this study variable in the intent to scrutinize the life satisfaction rating of Canadians, whether this is dependent on certain factors in life. Also, we decided on the specific input variables because intuitively, they are the most suitable data for our hypothesis and to prove that indeed,

there is a relationship between the y and x variables. All of the data were pulled from the General Social Survey - Family (GSS) that was conducted in 2017. The 2017 survey data set was selected because it was a relatively recent set of data, therefore more interesting and purposeful to observe as it reflects comparatively newer and current responses. Survey data includes a target population of non-institutionalized individuals who are 15 years of age or older in the 10 provinces of Canada. The survey's primary objectives are to gather data on social trends to monitor changes in the living conditions of Canadians throughout life, and to provide information on specific social policy issues of current interest.

20,602 respondents had responded to the survey, and the number of respondents is also the sample size. This survey is a cross-sectional design and uses a two-stage sampling design. During the first stage, the sampling groups are the groups of telephone numbers, and the second stage units are people in the identified households. The sampling frame is Statistics Canada's common telephone frame, which combines landline and cellular telephone numbers that are from the Address Register. Survey data were collected through self-completed online questionnaires and telephone interviews, handled both in English and French. Since responding to the survey was voluntary, there were non-responses, and survey respondents were allowed to have non-responses within their survey answers as well. Also, although there were households without telephone services- generally, these exclusions are small, and so introduced biases were expected to be minor too.

In the data set, there were numerous variables and responses, and some variables were very similar to one another. For instance, there were two questions regarding income, one for income of the respondent and another for income of the entire family. Since income of the respondent would usually be more direct and influential to the respondent, this variable was chosen instead of the other. There were no new variables that were constructed by combining multiple others, as each variable by itself normally had several response options, and it would be confusing to mix up the variables. The key features of the survey are that it covers various important life events and factors that influence life, and so the primary objectives of the survey are well covered and represented. Its strengths are that therefore, the survey is fairly accurate in terms of representing all the variables that affect life, and the large number of respondents and the appropriate survey designs also add to the accuracy of the survey.

A drawback to the data is that because of the numerous questions and responses in the survey and gathered data, it was time consuming to review all these to choose our variables. Another drawback to the data is that there are numerous vague responses such as "Don't know" or "Unsure", which are difficult to place in order which should be placed before the other. Furthermore, some weaknesses about the survey in general are that the questions could have been phrased more inclusively, and that the answer options could have been broader. For example, the variable "age at first birth" does not include males, and all males responded with "NA" to the question. However, males are broadly still affected by first child's birth, as they participate in nurturing the child. A more precise question would be with the phrase "age at first child's birth," which would permit males to answer. Moreover, the variable "sex" only represents female or male- such option for "intersex" is not included. Perhaps another category could have been added for "gender," in which respondents could have more comfortably represented themselves within the LGBTQ+ group.

Figures 1-6 show various visualizations of the raw data. We see that most respondents have a positive feeling about life with a bimodal (at 8 and 10), left-skewed distribution of `feelings_life`. An interesting result seen in Figure 2 is that most respondents are older than 50 and no respondents were younger than 20, despite the target population including persons 15 years of age or older. Moreover, most respondents make less than the national mean of around \$50,000 per year as seen in Figure 4. Most respondents were born in Canada and are not confident about whether they will have children in the future.

Figures 7-9 show the aggregated means for all respondents across the variables `feelings_life`, `age`, and `age_at_first_birth` by province. We see from the limited range of the variables for each plot that the provinces are not too different from one another for these factors.

Figure 1: Distribution of feelings\_life

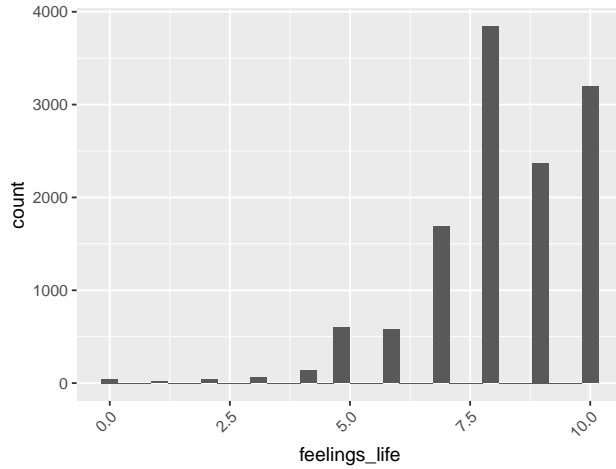


Figure 2: Distribution of age

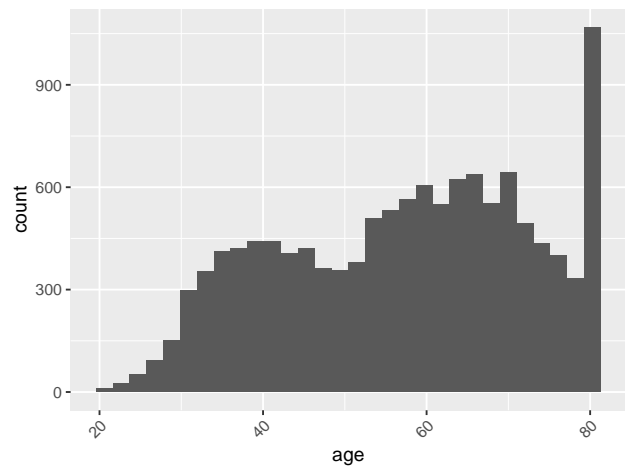


Figure 3: Distribution of age\_at\_first\_birth

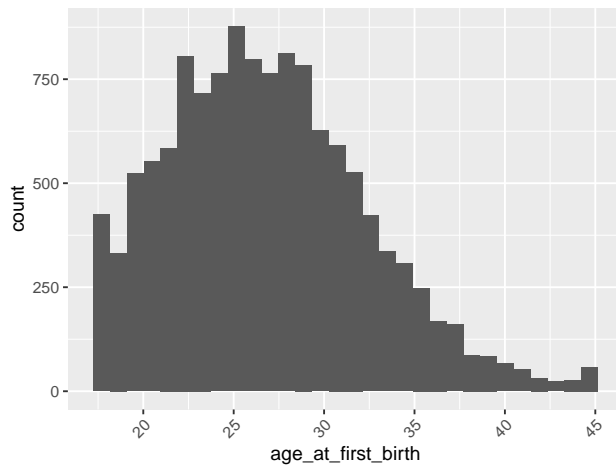


Figure 4: Distribution of income\_respondent

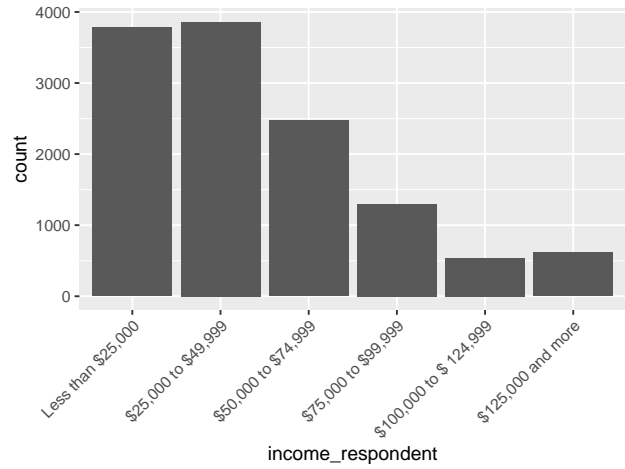


Figure 5: Distribution of place\_birth\_canada

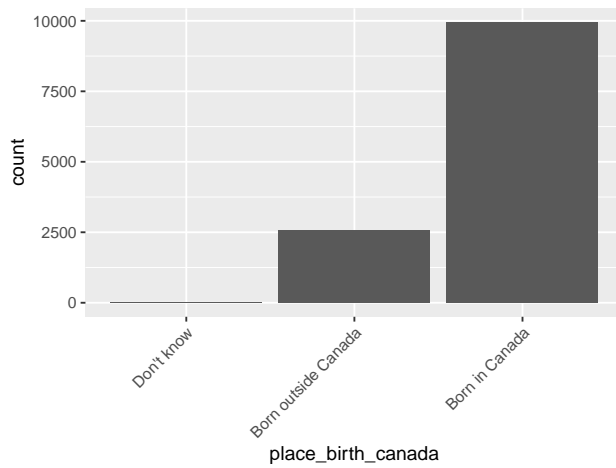


Figure 6: Distribution of future\_children\_intention

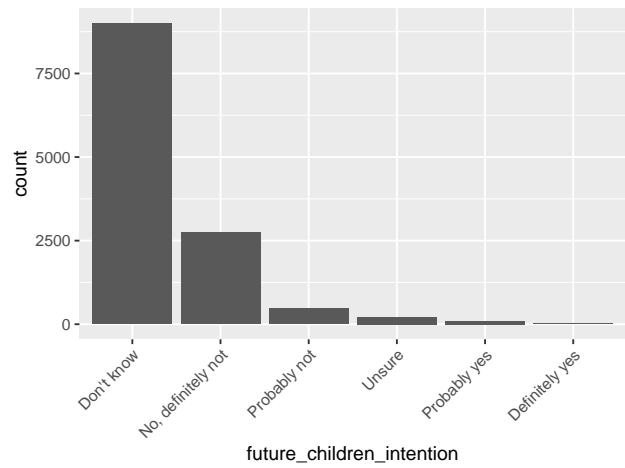


Figure 7: Mean feelings\_life by province

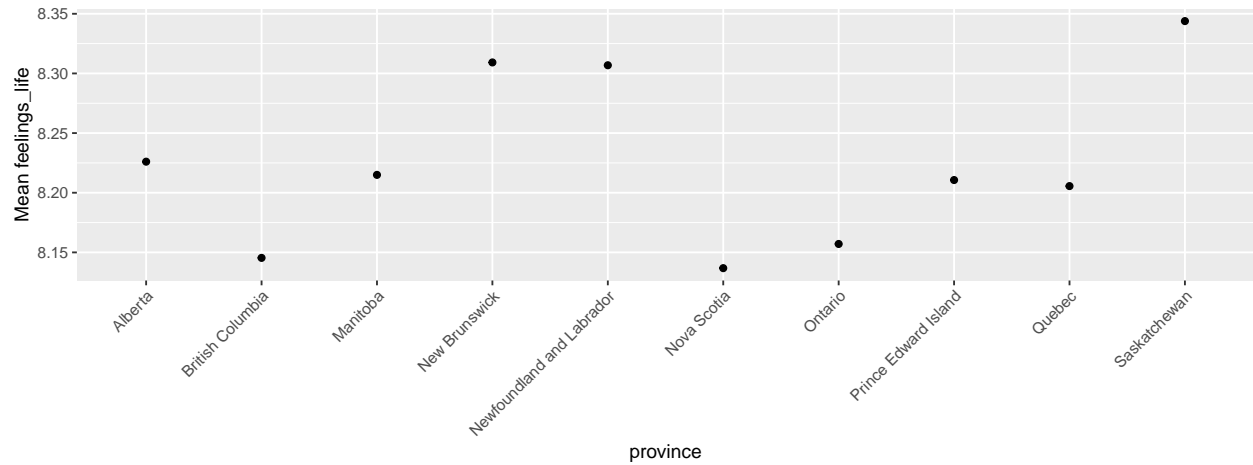


Figure 8: Mean age by province

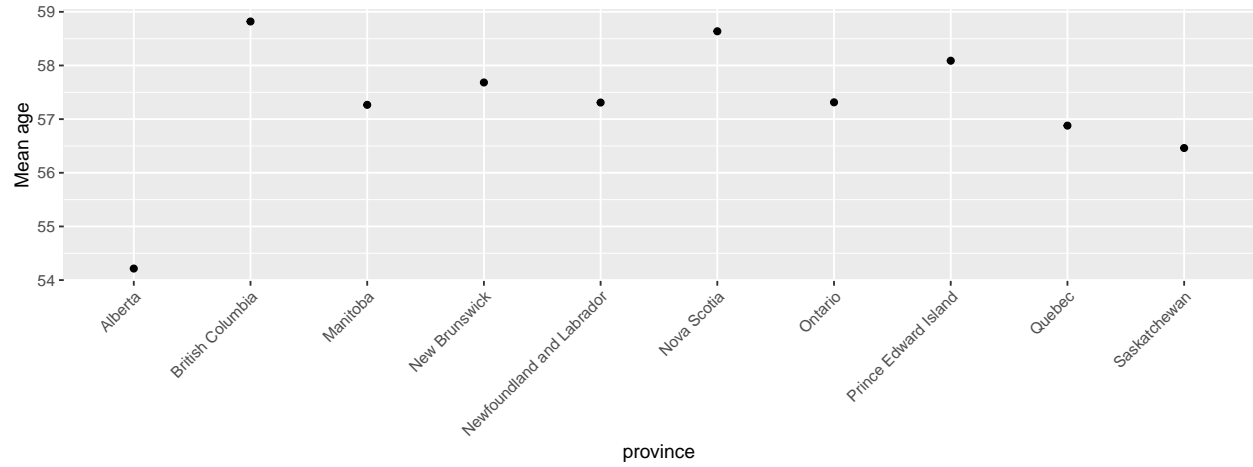
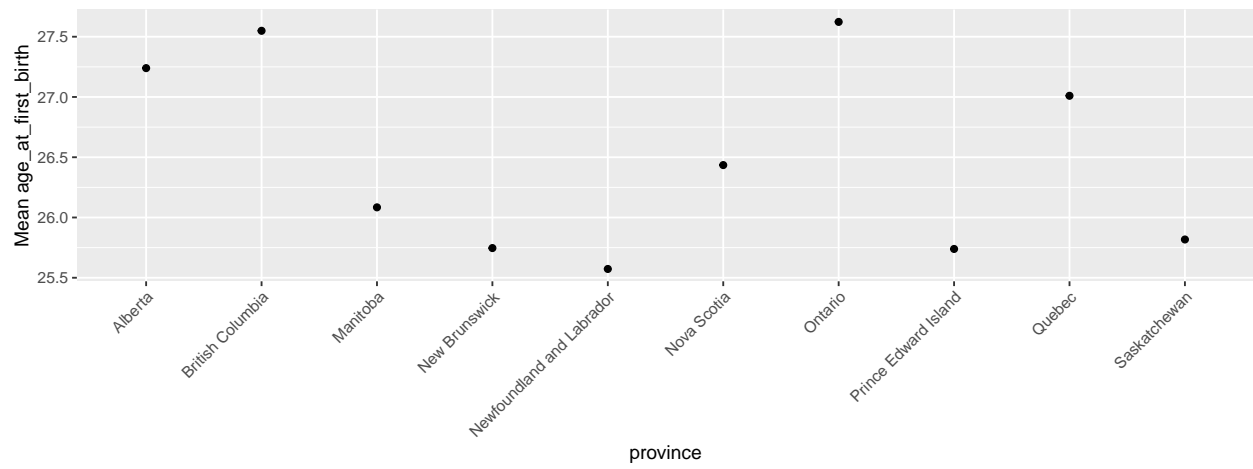
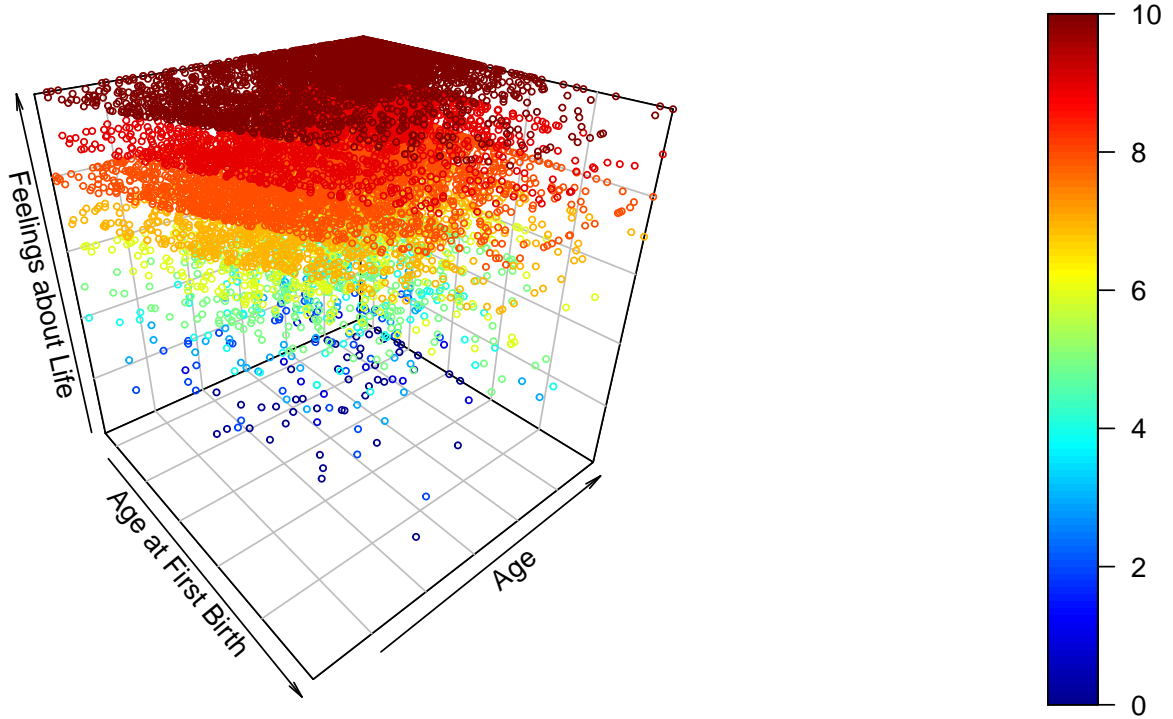


Figure 9: Mean age\_at\_first\_birth by province



**Figure 10: Age at First Birth & Age vs Feelings About Life**



## Model

Introduce the selected model here. It is expected that you will use some mathematical notation here. If you do please ensure that all notation is explained. You may also want to discuss any special (hypothetical) cases of your model here, as well as any caveats.

To predict the probability of a person being happy, which we define as:

$$\text{Prob}(h) := \begin{cases} 1, & \text{if feelings\_life} \geq 6 \\ 0, & \text{otherwise} \end{cases}$$

we fit a logistic regression model with some independent/predictor variables.

Table 1: Categorical Variables

Income of Respondent	Future Children Intention	Born in Canada?
Less than \$25,000	Don't know	Don't know
\$25,000 to \$49,999	No, definitely not	Born outside Canada
\$50,000 to \$74,999	Probably not	Born in Canada
\$75,000 to \$99,999	Unsure	
\$100,000 to \$ 124,999	Probably yes	
\$125,000 and more	Definitely yes	

The other predictor variables are `age_at_first_birth` and `age`.

Using the GSS data, we replicated the approach used in the original survey. A single-stage stratified sampling approach by applying finite population correction to the sample was employed, adjusting each observation by

the corresponding provincial population to reduce the variation. This yielded the following model:

$$\begin{aligned} \log\left(\frac{h}{1-h}\right) = & \beta_0 + \beta_1 x_1 + \beta_2 x_2 \\ & + \underbrace{\beta_{3a}x_{3a} + \beta_{3b}x_{3b} + \beta_{3c}x_{3c} + \beta_{3d}x_{3d} + \beta_{3e}x_{3e}}_{\text{dummy coding for income}} \\ & + \underbrace{\beta_{4a}x_{4a} + \beta_{4b}x_{4b}}_{\text{dummy coding for place birth Canada}} \\ & + \underbrace{\beta_{5a}x_{5a} + \beta_{5b}x_{5b} + \beta_{5c}x_{5c} + \beta_{5d}x_{5d} + \beta_{5e}x_{5e}}_{\text{dummy coding for future children intention}} \end{aligned}$$

The functional form of the logistic model gives the logarithm of the odds of the outcome variable – in this case, the binary variable **happy** ( $h$ ). For each of the categorical variables in Table 1, we use dummy variable coding with the variable at the top representing the baseline, in order to be able to assess what effect, if any, moving to a category would have compared to the baseline.

## Results

The coefficients fitted using the logistic regression model are given in Table 2. The *OR* column gives the odds ratio, *Beta (SE)* gives the logarithm of the odds ratio and corresponding standard error of the estimate. *P* gives the p-value for the significance test which evaluates the probability of encountering data as or more significant than the Wald test statistic under the assumption that the population parameter coefficient of interest is equal to 0. The Wald test is relevant because of the log-transformed odds ratio; the model summary summarizes the results of this test to each of the parameters.

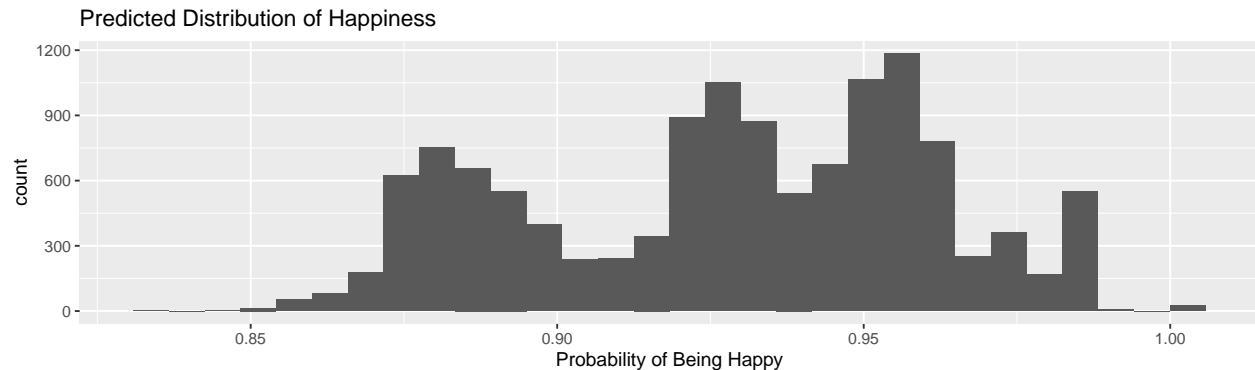
Table 2: Logistic Regression Model Summary

	Variable	OR	Beta (SE)	P
$\beta_0$	Intercept	-	2.12 (0.88)	0.02
$\beta_1$	age_at_first_birth	1.02	0.02 (0.01)	0.01
$\beta_2$	age	1.00	-0.00 (0.00)	0.35
income cat. var	as.factor(income_respondent)			
income baseline	Less than \$25,000 (ref)	-	-	-
$\beta_{3a}$	\$25,000 to \$49,999	1.62	0.48 (0.09)	<0.001
$\beta_{3b}$	\$50,000 to \$74,999	2.37	0.86 (0.12)	<0.001
$\beta_{3c}$	\$75,000 to \$99,999	2.82	1.04 (0.17)	<0.001
$\beta_{3d}$	\$100,000 to \$ 124,999	4.39	1.48 (0.31)	<0.001
$\beta_{3e}$	\$125,000 and more	7.65	2.03 (0.35)	<0.001
birthplace cat. var	as.factor(place_birth_canada)			
birthplace baseline	Don't know (ref)	-	-	-
$\beta_{4a}$	Born outside Canada	0.61	-0.50 (0.82)	0.55
$\beta_{4b}$	Born in Canada	0.69	-0.37 (0.82)	0.65
children int. cat. var	as.factor(future_children_intention)			
children int. baseline	Don't know (ref)	-	-	-
$\beta_{5a}$	No, definitely not	0.95	-0.05 (0.13)	0.67
$\beta_{5b}$	Probably not	0.95	-0.05 (0.24)	0.84
$\beta_{5c}$	Unsure	0.75	-0.28 (0.30)	0.34
$\beta_{5d}$	Probably yes	1.20	0.18 (0.45)	0.69
$\beta_{5e}$	Definitely yes	188128.04	12.14 (0.25)	<0.001

The model has the following interpretations:

- The intercept term has no meaningful interpretation but it is statistically significant with a very small p-value. It shifts the overall model prediction by  $e^{\beta_0}$ .
- The continuous variables **age\_at\_first\_birth**, **age** don't affect the model by much with odds-ratios slightly off from 1.  $\beta_1$  is statistically significant but  $\beta_2$  is not, with a p-value of 0.35, and so we fail to reject the null hypothesis that  $\beta_2 \neq 0$ .
- We see a very large prediction of the odds ratio for  $\beta_{5e}$  (**future\_children\_intention**=*Definitely yes*) which is likely caused by the large number of people who voted *Don't know*. The other parameters  $\beta_{5i}$  for other levels of **future children intention** are not statistically significant, which is also due to the same reason. However, we included this factor in the model because it is intuitive that living a happy life would be necessary to raising and taking care of children. Despite this, there are certainly outliers in reality in low-income households, wherein there are many children in the same household.
- **income\_respondent** is a meaningful variable in this model. For all six levels, the odds ratio is greater than 1 which implies that each increase in the respondent's income results in an increase in the log odds of being happy compared to the baseline.

Using a variety of parameters with this model where each combination has a uniform probability of being chosen, we obtain the following histogram (Figure 11) that shows the class distribution of happiness across Canada. Note that the bins are lower-bound inclusive, explaining a number of predictions appearing beyond 1 (these probabilities are 1). The probabilities are very high, but that may be due to an imbalance in the survey data – this will be discussed in the Weaknesses section of the report.



## Discussion

## Weaknesses

## Next Steps

## References

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