

The Gender Gap Unveiled:*

Tracing Unemployment Trends from 2018 to 2023

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This paper examines the impact of gender on unemployment rates across various demographics from 2018 to 2023, employing a comprehensive dataset from the Labor Force Survey to unravel the interplay between gender and overall economic participation. We also employ the use of a linear regression model to help dissect the influence of gender, among other predictors, on the unemployment rates of Generation Z. Our findings reveal significant disparities in unemployment trends, with distinct patterns emerging between different genders and age groups, highlighting the persistent gender gap in the labour market. This research is important because it sheds light on the underlying factors that contribute to unemployment disparities, offering insights into the effectiveness of current policies and the need for targeted interventions. Ultimately, this paper enhances our understanding of the labour market's complexities, advocating for inclusive economic strategies to bridge the gender divide in employment opportunities.

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*Code and data are available at: <https://github.com/fatimahsy/Gender-Unemployment-.git>.

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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](#)...

2 Data

2.1 Data source

2.2 Datat Features

2.3 Data Analysis

3 Model

3.1 Model Set-up

$$\begin{aligned}
 \text{Gen Z Unemployment} = & \beta_0 + \beta_1 \cdot \text{Sex} \\
 & + \beta_2 \cdot \text{Reference Period} \\
 & + \beta_3 \cdot \text{Adult Unemployment} \\
 & + \beta_4 \cdot \text{Seniors Unemployment} + \epsilon
 \end{aligned}$$

Where:

- **Gen z Unemployment** is
- **Sex** is
- **Reference Period** is another independent variable that represents the time frame of the data collection (e.g., year).
- **Adult Unemployment** and **Seniors Unemployment** are independent variables
- β_0 is the y-intercept, representing the expected value of **Gen z Unemployment** when all the independent variables are 0.
- $\beta_1, \beta_2, \beta_3, \beta_4$ are the coefficients for each independent variable, representing the change in **Gen z Unemployment** for a one-unit change in the respective independent variable, holding all other variables constant.
- ϵ represents the error term, accounting for the variability in **Gen z Unemployment** not explained by the model.

Talk more about it.

3.2 Model Justification

4 Results

4.1 Generation Z unemployment

4.2 Seniors Unemployment

4.3 Adults Unemployment

4.4 Discussion

4.5 Implications of these results

4.6 Possible Causes

4.7 Weaknesses and next steps

5 Conclusion

Appendix

A Additional data detail

References

- 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 Grolmund, et al. 2019. “Welcome to the tidyverse.” *Journal of Open Source Software* 4 (43): 1686. <https://doi.org/10.21105/joss.01686>.