

# Bayesian Insights into Aerial Bombing Strategies: An Ordered Logistic Regression Analysis of WWII Target Prioritization Against Germany\*

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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....

## 2 Data

Some of our data is of penguins (**?@fig-bills**), from Horst, Hill, and Gorman (2020).

Talk more about it.

And also planes (**?@fig-planes**). (You can change the height and width, but don't worry about doing that until you have finished every other aspect of the paper - Quarto will try to make it look nice and the defaults usually work well once you have enough text.)

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\*Code and data are available at: [https://github.com/yunzhaol/aerial\\_bomb\\_priority.git](https://github.com/yunzhaol/aerial_bomb_priority.git).

Table 1: 2022 CES (Cooperative Election Study) Data (Cultural)

```
# analysis_data |>
#   ggplot(aes(x = width, y = length)) +
#   geom_point(alpha = 0.8) +
#   theme_minimal() +
#   labs(x = "Wing width (mm)",
#         y = "Wing length (mm)")
```

Talk way more about it.

### 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 B](#).

#### 3.1 Model set-up

Define  $y_i$  as the number of seconds that the plane remained aloft. Then  $\beta_i$  is the wing width and  $\gamma_i$  is the wing length, both measured in millimeters.

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

$$\mu_i = \alpha + \beta_i + \gamma_i \quad (2)$$

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

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

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

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

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`.

The Bayesian model for analyzing the prioritization of aerial bombing targets during WWII is formulated as follows:

Let  $(y_i)$  be the ordered target priority level for each bombing mission, which can take values from  $\{1, 2, 3, 4\}$ , representing the levels from ‘target of last resort’ to ‘primary target’.

The likelihood of  $(y_i)$  given the predictor variables is modeled by an ordered logistic regression:

Table 2: Explanatory models of flight time based on wing width and wing length

$$[ \text{logit}( \hat{y}_i^{(k)} ) = \eta_k - ( \eta_0 + \eta_1 \times \text{tgt\_industry}_i + \eta_2 \times \text{country\_flying\_mission}_i + \eta_3 \times \text{total\_tons}_i + \eta_4 \times \text{ac\_attacking}_i ) ]$$

where -  $( \hat{y}_i^{(k)} )$  is the probability of  $( y_i )$  being at least level  $( k )$ , -  $( \eta_k )$  are the cutpoints between the categories of target priority (with  $( K-1 )$  cutpoints for  $( K )$  categories), -  $( \text{tgt\_industry}_i )$  is the industry type of the target, -  $( \text{country\_flying\_mission}_i )$  represents the country conducting the bombing mission, -  $( \text{total\_tons}_i )$  is the amount of bombs dropped in tons, -  $( \text{ac\_attacking}_i )$  is the number of aircraft involved in the mission.

Priors for the model parameters are set as follows: -  $( \eta_0, \eta_1, \eta_2, \eta_3, \eta_4 \sim \text{Normal}(0, 2.5) )$ , - Cutpoints  $( \eta_k )$  are ordered with priors  $( \eta_k \sim \text{Normal}(0, 2.5) )$  ensuring  $( \eta_1 < \eta_2 < \dots < \eta_{K-1} )$ .

### 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 2.

## 5 Discussion

### 5.1 First discussion point

If my paper were 10 pages, then should 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 `?@fig-ppcheckandposteriorvsprior-1` we implement a posterior predictive check. This shows...

In `?@fig-ppcheckandposteriorvsprior-2` we compare the posterior with the prior. This shows...

Examining how the model fits, and is affected  
by, the data

#### B.2 Diagnostics

`?@fig-stanareyouokay-1` is a trace plot. It shows... This suggests...

`?@fig-stanareyouokay-2` is a Rhat plot. It shows... This suggests...

Checking the convergence of the MCMC algo-  
rithm

## References

- Goodrich, Ben, Jonah Gabry, Imad Ali, and Sam Brilleman. 2022. “Rstanarm: Bayesian Applied Regression Modeling via Stan.” <https://mc-stan.org/rstanarm/>.
- Horst, Allison Marie, Alison Presmanes Hill, and Kristen B Gorman. 2020. *Palmerpenguins: Palmer Archipelago (Antarctica) Penguin Data*. <https://doi.org/10.5281/zenodo.3960218>.
- 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 Golemund, et al. 2019. “Welcome to the tidyverse.” *Journal of Open Source Software* 4 (43): 1686. <https://doi.org/10.21105/joss.01686>.