



University  
of Glasgow



# Factors Influencing the Duration of Stay for Animals in Shelters

Group17



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## 1.1 Problem statement

- **Problem Statement:**
- Which factors influence the number of days an animal spends in the shelter before its final outcome is decided?

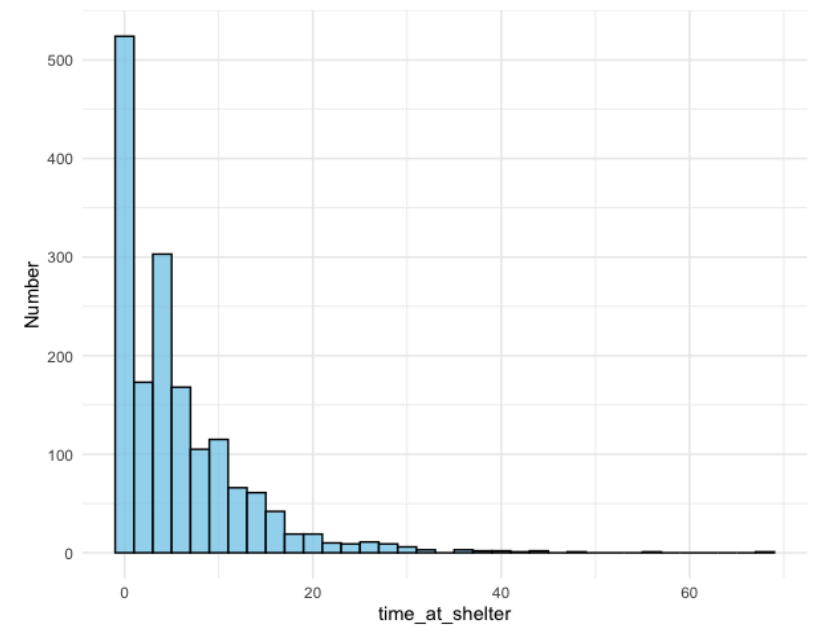


## 1.2 Data Overview

### Key variables:

- Animal\_type – The type of animal admitted to the shelter
- Month – Month the animal was admitted, recorded numerically with January=1
- Year. – Year the animal was admitted to the shelter.
- Intake\_type – Reason for the animal being admitted to the shelter
- Outcome\_type – Final outcome for the admitted animal
- Chip\_Status – Did the animal have a microchip with owner information?
- Time\_at\_Shelter – Days spent at the shelter between being admitted and the final outcome.

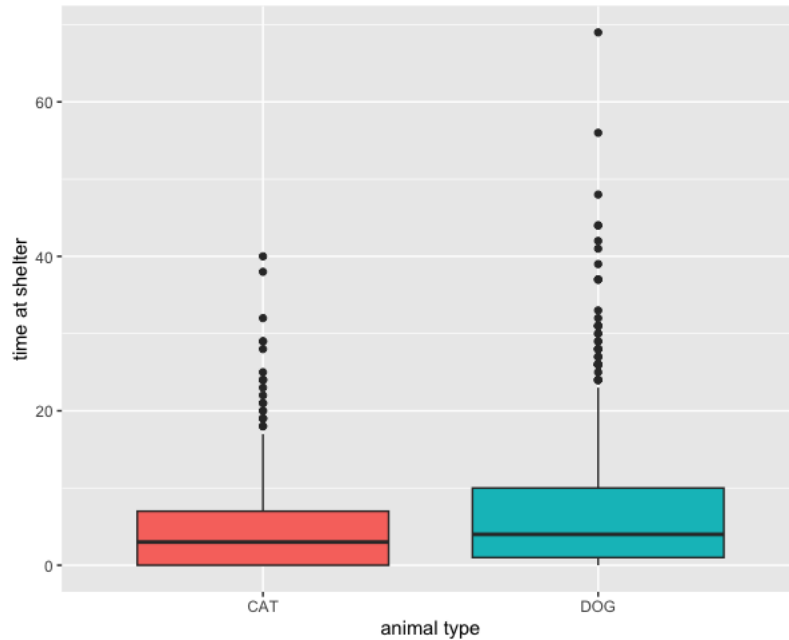
### 1. Histogram of the number of days an animal spends in the shelter



The chart shows a right-skewed distribution, with most animals staying 0-10 days and few staying beyond 30 days.

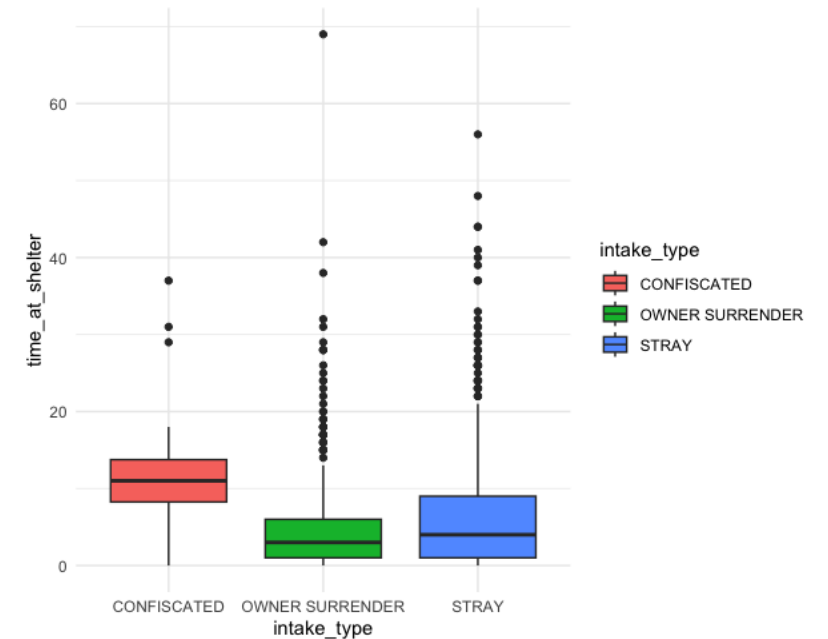
## 1.2 Data Overview

### 2. Boxplot of time\_at\_shelter and animal\_type



Cats have a lower median stay than dogs, but dogs show a wider distribution with longer stays.

### 3. Boxplot of time\_at\_shelter and intake\_type

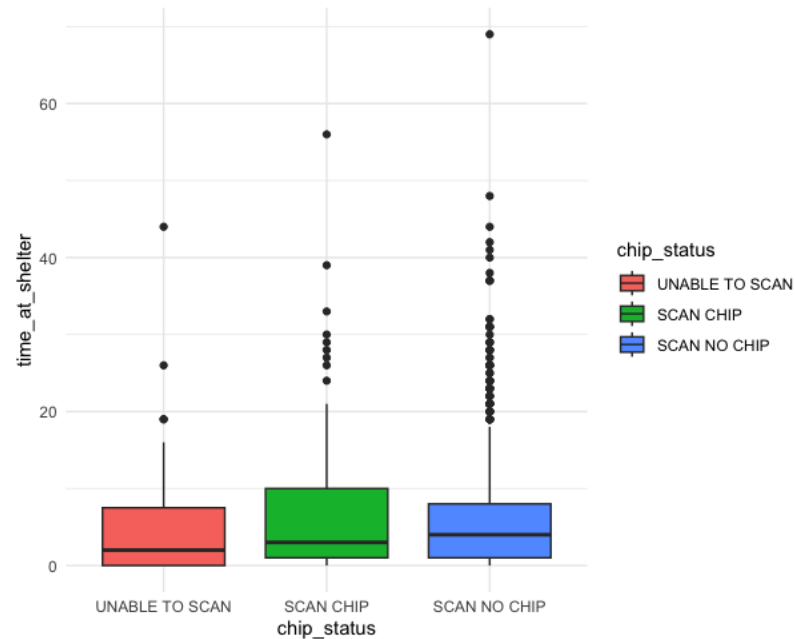


The median of confiscated animals was significantly higher than that of the other two groups, indicating that confiscated animals stayed concentrated and longer.



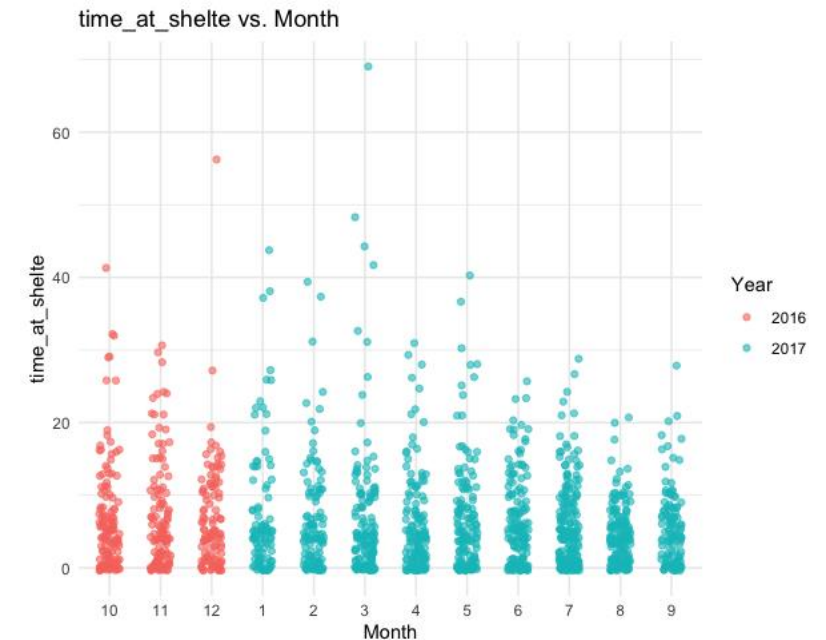
## 1.2 Data Overview

### 4. Boxplot of time\_at\_shelter and chip\_status



The medians were similar across chip statuses, suggesting comparable shelter stays. However, the IQR was wider for scanned animals, indicating greater data dispersion.

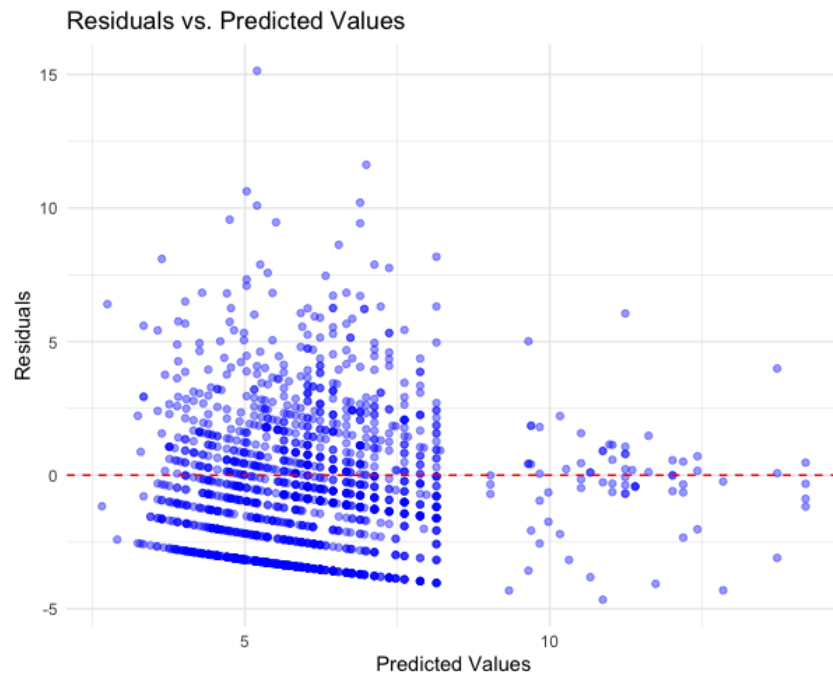
### 5. Scatter plot of time\_at\_shelter and intake\_type



The dataset spans from October 2016 to September 2017 with complete monthly data. Animal shelter stays were mostly short (under 20 days) and showed no significant seasonal variation.

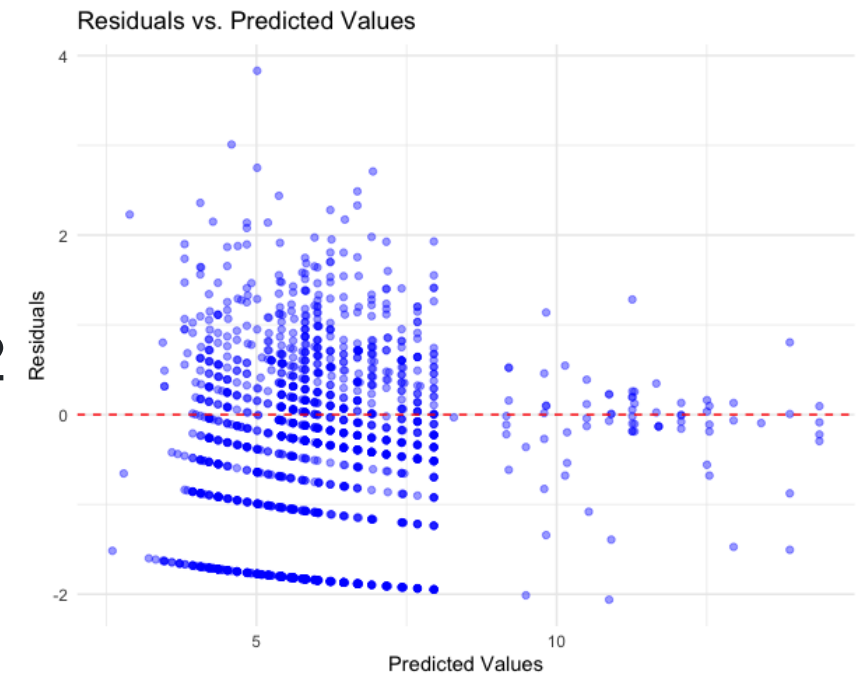
## 2. Residual Plot

### Model1 : Poisson



Residual variability is higher at lower predicted values, indicating excessive dispersion, consistent with the previously calculated dispersion parameter.

### Model2 : nb\_model\_2

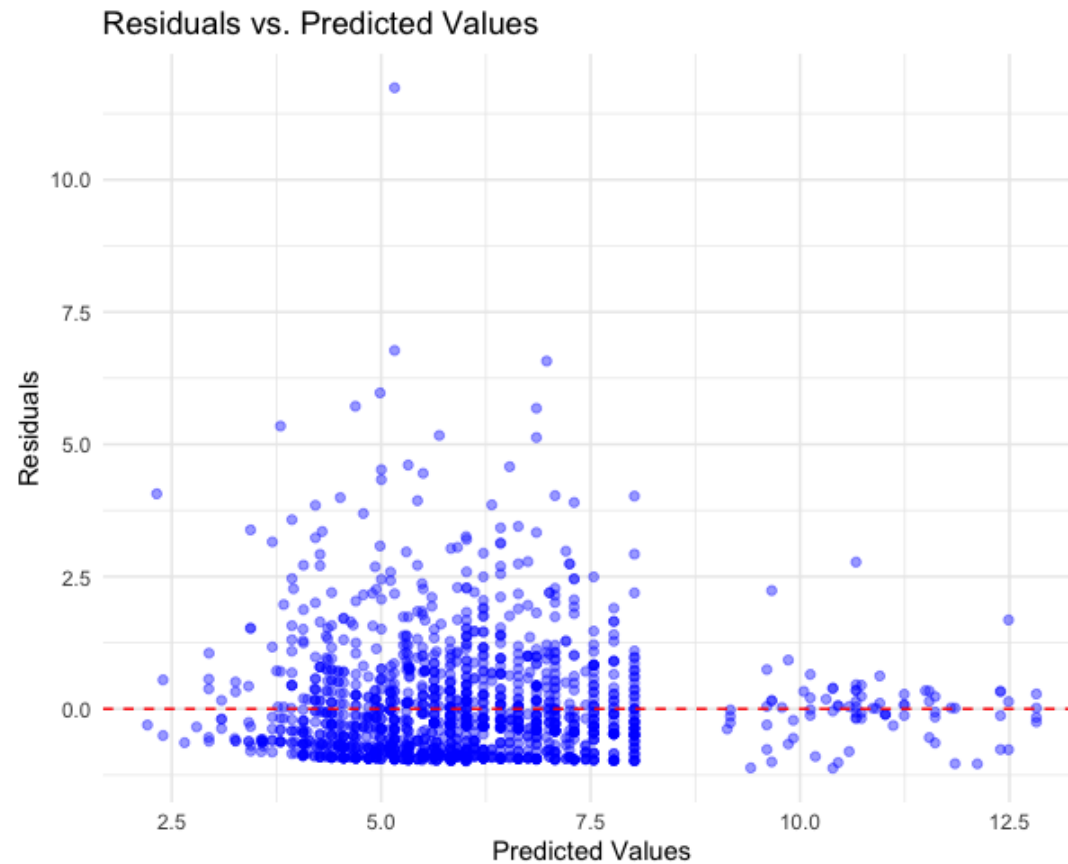


The variance of the residuals is generally in the range of -2 to +2. The negative binomial regression model basically solves the overdispersion problem.

## 2. Residual Plot

**Model3 :**  
Zero-flated\_model

The residual plot highlights useful insights, confirming the presence of dispersion and validating the need for a more flexible model.





## 3.1 Key Findings Summary

### Distribution

This report examines animal shelter stays, showing a right-skewed distribution with most animals staying under 10 days.

### Predictors

Using Poisson and Negative Binomial models with predictors like intake\_type, animal\_type, chip\_status, and month, we found that owner-surrendered or stray animals stay shorter, dogs stay slightly longer, and chip status and intake month have weaker effects.

### Modeling

Given many zero-day stays, a Zero-Inflated Negative Binomial model provided the best fit, distinguishing structural zeros from regular counts for deeper insight.

## 3.2 Limitations & Future Work

### Problem statement

- The dataset contains a large number of zero values.
- Multiple explanatory variables are present.
- GLM does not fit well.

### Proposed solution

- Explore ANN as an alternative predictive model.
- Investigate methods for transforming non-numerical variables into numerical ones.
- Consider whether the frequency of observer occurrences can serve as a conversion criterion.





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# Thank you!

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