

# Data Analysis 3

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## Introduction

In this paper we will be looking at data related to calves. The data comes from an experiment designed to study the impact dietary treatments given to pregnant heifers had on the development of the calves. The study was conducted over a three year period and involved three different dietary treatments given to select groups of heifers in the final trimester. In total the data has 22 variables for 120 entries, though some data points are missing.

For more information on the experiment, the data, or any other files used in this paper see our [Github page](https://github.com/RyanLalicker/Data-Analysis-2-STAT-325-825) which can be found at <https://github.com/RyanLalicker/Data-Analysis-2-STAT-325-825>. The coding languages used in the paper are R and SAS. The corresponding code can be found in *Appendix A - R Code* and *Appendix B - SAS Code* respectively.

## Exploring the Data

### Variables

As mentioned above the experiment used three different dietary treatments. These were DDG, CON, and MET. For the first two trimesters the heifers were given one of seven developmental treatments, found in `Development.Treatment`, and then in the final trimester the each was given one of the three treatments mentioned above. This is recorded in the `Calan.Treatment` column of the data set.

The heifers were placed into one of four pens by weight, which can be seen in the column `Pen #`. They were then artificially inseminated from an assigned sire, which we will assume was done randomly since the client says weight was not a factor. The sire is represented by the column of the same name and has six unique entries.

Upon the birth of the calves, several measurements were taken. These include the sex of the calf, weights taken at both birth and slaughter, and scores of both the calf's vigor and the ease of birth. The variable names line up with these descriptions.

Other variables, such as the id of the calf, length of gestation for the heifer, and postmortem scoring such as hot carcass weight (HCW) are included as well. (Saner (2024)). Note two birthdays are included in the data, `Birth.date` and `Birth.date.1`. These variables will not be used in the models below so no further investigation was done on our part to determine the differences.

The client's main focus is the effect the third trimester treatment and the sex of a calf have on the calf's vigor score, ease of birth score, and final body weight. Therefore, these are the variables we will place more of an emphasis on, while exploring the effect some of the other variables may have.

## Missing Values

### UPDATE THIS AFTER SEEING WHAT VARIABLES ARE NEEDED FOR THE MODEL

The data contains some missing values. In regards to the five variables the client is most interested in, 19 entries are missing one or more values. Since that still leaves 101 entries with all five variables

Initially the dataset has missing values in the following columns: Pen #: 4 missing values DMI: 36 missing values SEX: 1 missing value

The heatmap above visualizes the distribution of missing data in the dataset. Columns with missing values (e.g., "Pen #", "DMI", and "SEX") are marked with red line gaps.

```
# Might want this later.  
# May need to add more later.  
neededvars <- c("Calan.Treatment", "SEX", "Final.Calf.BW", "Calving.Ease", "Calf.Vigor")  
data <- data[complete.cases(data[, neededvars]), ]
```

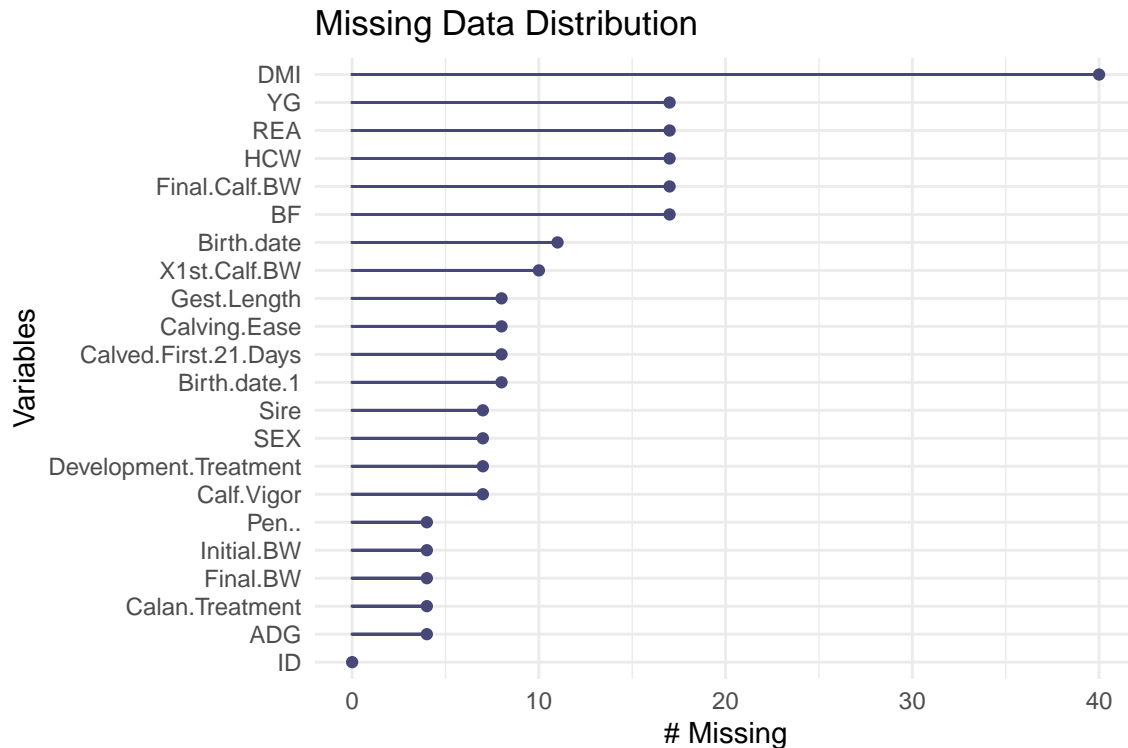


Figure 1: Chart counting the number of missing values for each variable within the data.

## Cleaning the dataset

This code cleans a data set by replacing all occurrences of . with NA to standardize missing values. It ensures columns are assigned the correct data types, converting numeric-like columns to numeric and others to factor. Missing values are handled by imputing the median for numeric columns and the mode for factor columns. After cleaning, the code verifies that no missing values remain in the dataset.

```
[1] "Remaining missing values: 0"
```

## Summary Statistics

```
[1] "Summary Statistics for Numerical Variables"
```

	ID_mean	ID_sd	ID_min	ID_max	ID_median	Birth.date_mean	Birth.date_sd
1	744.2583	80.9808	600	898	744	22.925	14.97795



```
[1] "Summary Statistics for Categorical Variables"
```

data frame with 0 columns and 1 row

## Exploring the Data

### Relationships among variables

## Potential models

### Model 1

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
calan_treatment	2	1840	920	2.347	0.100267
sex	1	4802	4802	12.254	0.000666 ***
initial_bw	1	800	800	2.041	0.155884
calan_treatment:sex	2	1472	736	1.879	0.157560
Residuals	113	44279	392		

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Tukey multiple comparisons of means  
95% family-wise confidence level

Fit: aov(formula = final\_calf\_bw ~ calan\_treatment \* sex + initial\_bw, data = data\_cleaned)

```
$calan_treatment
      diff      lwr      upr      p adj
2-1  5.210526 -5.575086 15.9961383 0.4871443
3-1 -4.287081 -14.698511  6.1243487 0.5922780
3-2 -9.497608 -19.909038  0.9138223 0.0814415
```

```
$sex
      diff      lwr      upr      p adj
2-1 12.63931  5.443081 19.83554 0.0007141
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
calan_treatment	2	1840	920	2.347	0.100267
sex	1	4802	4802	12.254	0.000666 ***
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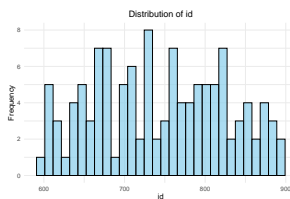


Figure 2

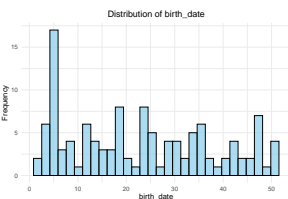


Figure 3

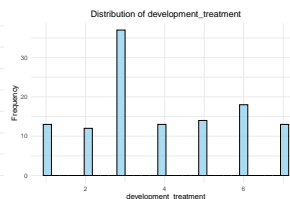


Figure 4

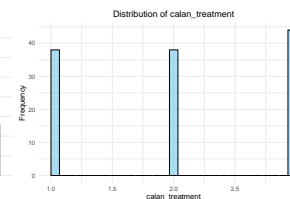


Figure 5

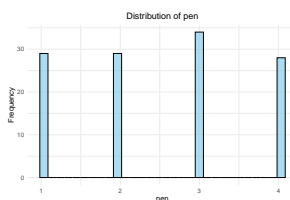


Figure 6

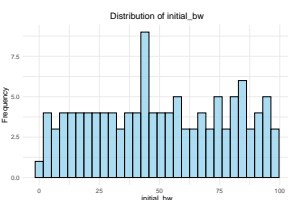


Figure 7

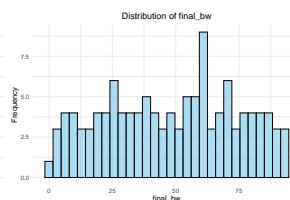


Figure 8

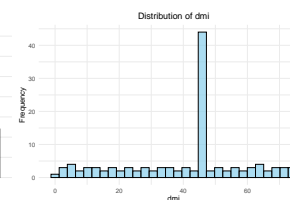


Figure 9

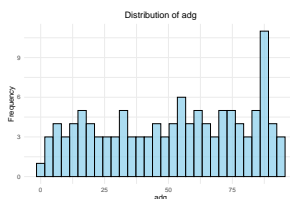


Figure 10

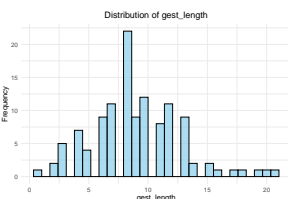


Figure 11

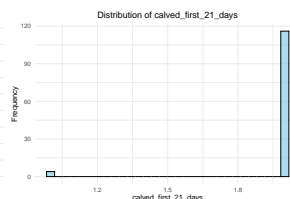


Figure 12

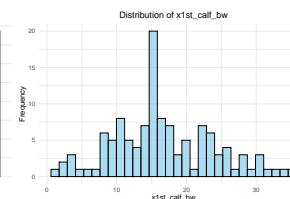


Figure 13

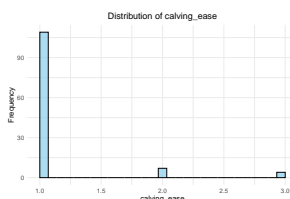


Figure 14

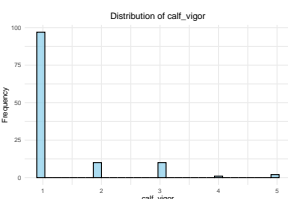


Figure 15

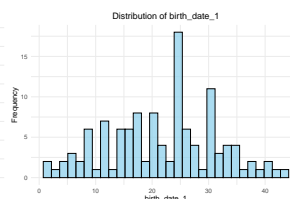


Figure 16

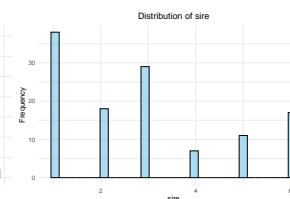


Figure 17

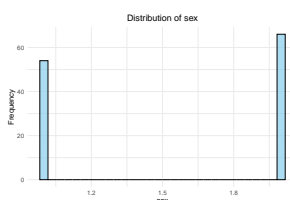


Figure 18

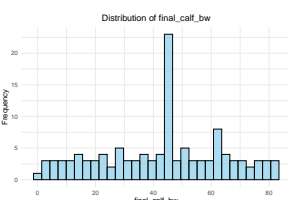


Figure 19

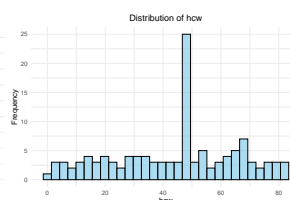


Figure 20

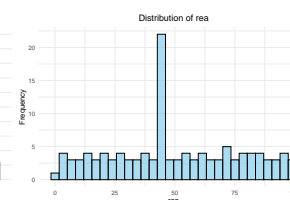


Figure 21

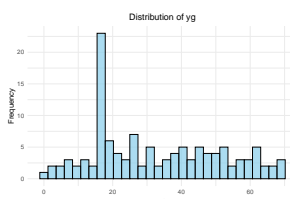


Figure 22

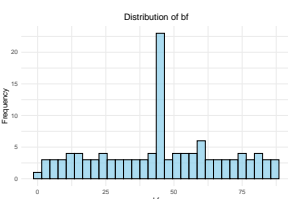


Figure 23

calan_treatment:sex	2	1472	736	1.879	0.157560
Residuals	113	44279	392		

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Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

$$y_{ijklmn} = ENTER - MODEL - HERE$$

where  $y_{ijklm}$  represents the *dependent variable*, ...

![Picture of SAS Output](filename.png){width="3in"}

## Conclusion

## Recomendation

## References

Saner, Brianna, Randy & Buseman. 2024. “How Many Pounds of Meat Can We Expect from a Beef Animal?” 2024. <https://beef.unl.edu/beefwatch/2020/how-many-pounds-meat-can-we-expect-beef-animal>.



## **Appendix A - R Code**

## Appendix B - SAS Code

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## Appendix C - Additional SAS Output

