# Analysis of MechaCar Data

## **Technical Report**

# Background

The results of an mpg testing dataset of 50 potential prototype MechaCars. The MechaCar prototypes were produced using multiple design specifications to identify ideal vehicle performance. Multiple metrics such as vehicle length, vehicle weight, spoiler angle, drivetrain, and ground clearance were collected for each vehicle.

MechaCar suspension coil test results from multiple production lots. In this dataset, the weight capacity of multiple suspension coils was tested to determine if the manufacturing process is consistent across lots.

## Observations

### Multi Linear Regression

Multi Linear Regression observations focused on below questions:

- 1. Which variables/coefficients provided a non-random amount of variance to the mpg values in the dataset?
- 2. Is the slope of the linear model considered to be zero? Why or why not?
- 3. Does this linear model predict mpg of MechaCar prototypes effectively? Why or why not?

In the summary output, each Pr(>|t|) value represents the probability that each coefficient contributes a random amount of variance to the linear model. According to our results, vehicle length and ground clearance (as well as intercept) are statistically unlikely to provide random amounts of variance to the linear model. In other words, the vehicle length and ground clearance have a significant impact on mpg. When an intercept is statistically significant, it means there are other variables and factors that contribute to the variation in mpg that have not been included in our model. These variables may or may not be within our dataset and may still need to be collected or observed.

Despite the number of significant variables, the multiple linear regression model outperformed the simple linear regression. According to the summary output, the r-squared value has increased from 0.37 in the simple linear regression model to 0.71 in our multiple linear regression model and the p-value changed from 2.636e-06 to 5.35e-11.

The slope of the linear model is not considered to be zero. Through this data, we fail to reject the null hypothesis based on the coefficient values, the dependent variable mpg has a linear relationship to few independent variables.

```
lm(formula = mpg ~ vehicle_length, data = MechaCar_table)
Residuals:
            1Q Median
                           3Q
   Min
                                  Max
-26.303 -7.160 -1.231 9.374 26.670
Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)
              -25.0622
                       13.2960 -1.885 0.0655 .
vehicle_length
               4.6733
                          0.8774
                                  5.326 2.63e-06 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 12.47 on 48 degrees of freedom
Multiple R-squared: 0.3715, Adjusted R-squared: 0.3584
F-statistic: 28.37 on 1 and 48 DF, p-value: 2.632e-06
```

Fig1: Linear Regression Summary

Fig2: Multiple Linear Regression

### Suspension Coil Summary

Suspension Coil manufacturing data summary:

Mean: 1499.531Median: 1499.747Variance: 76.234

Standard Deviation: 8.731

```
> summary(suspension_data)
VehicleID Manufacturing_Lot PSI
Length:150 Length:150 Min. :1463
Class :character Class :character 1st Qu.:1497
Mode :character Mode :character Median :1500
Mean :1500
3rd Qu.:1501
Max. :1536
```

Fig3: Summary on full suspension data

```
> # Perform Summary Analysis on suspension coil dataset
> # Since we have to perform two sample t-test to compare Mean values, group the data.
> lot_summary <- suspension_data %>% group_by(Manufacturing_Lot) %>%
+ summarise(Mean=mean(PSI), Median=median(PSI), Variance=var(PSI), SD=sd(PSI))
`summarise()` ungrouping output (override with `.groups` argument)
> lot_summary
# A tibble: 3 x 5
 Manufacturing_Lot Mean Median Variance
                                         <db1> <db1>
                       <db1> <db1>
                      1500. 1500.
                                         1.15 1.07
1 Lot1
                      <u>1</u>500. <u>1</u>499.
2 Lot2
                                                3.18
                                        10.1
                      <u>1</u>499. <u>1</u>498.
3 Lot3
                                       220.
                                               14.8
> # Total Summary
> total_summary <- suspension_data %>%
  summarise(Mean=mean(PSI), Median=median(PSI), Variance=var(PSI), SD=sd(PSI))
> total_summary
      Mean
             Median Variance
1 1499.531 1499.747 76.23459 8.731242
```

Fig4: Summary of the Total suspension data by grouping

#### **Summary Interpretation:**

The design specifications for the MechaCar suspension coils dictate that the variance of the suspension coils must not exceed 100 pounds per inch. Since suspension data variance (76.23) is under 100PSI, we can predict that the current manufacturing data meets the design specifications.

#### T-Test

Performing the t-test on the suspension coil data determines the suspension coil's pound-per-inch results are statistically different from the mean population results of 1,500 pounds per inch.

Fig5: T-test on suspension data

Here true mean is not equal to 1500PSI and there is 95% confidence interval.

### Design Study:

Upper management is looking for your expertise and wants you to design a study that compares the performance of the MechaCar prototype vehicle to other comparable vehicles on the market. In the MechaCarWriteUp.txt text file, write a short description of a statistical study that can quantify how the MechaCar outperforms the competition. In your study design, be sure to write about the following considerations:

For further design study on MechaCar performance with competitors, we can perform statistical analysis on other customer focused aspects such as cost feature of the car and its significance relations with other variables: mpg, color/specific finish, hp. With more data, we can analyze if cost is predictive of the variables by performing Regression. We can even perform correlation to identify how variables are related to each other (positively/negatively). If you have diff makes of cars, we could perform t-test based on cost analysis between different makes to observe significance difference between variables. We may also focus on the sales data to identify MechaCar outperforms with competitors.