

Air Under the Microscope: Modelling Pollutant Effects on Air Quality

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DATA 603: Statistical Modelling With Data



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A background image showing a hazy, smoggy city street. Several pedestrians are walking in the foreground, wearing face masks and winter coats. The street is lined with utility poles and wires, and the overall atmosphere is grey and overcast.

Rationale:

- Clean air is essential for **healthy living**, and air quality directly influences both health and **quality of life**;
- Air pollution significantly impacts health, causing an estimated **4.2 million premature deaths in 2019** due to respiratory diseases, cardiovascular conditions, and cancers;
- **Environmental damage** from pollutants affects ecosystems, depleting soil nutrients and harming biodiversity; and
- Air quality can vary significantly, even within hours, having a **reliable monitoring** and **forecasting** system is crucial.

Aim:

- Deliver comprehensive **insights into Calgary's air quality**; and
- Build a **model that can measure and forecast air quality**, with the ultimate goal of **saving** and **improving lives**.

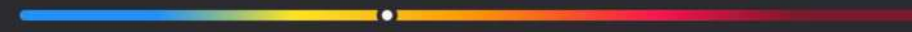
Air Quality



Moderate Health Risk

Scale: Canada (AQHI)

Air quality index is 5, which is worse than yesterday at about this time.



Health Information

No need to modify your usual outdoor activities unless you experience symptoms such as coughing and throat irritation. At-risk populations should consider reducing or rescheduling strenuous activities outdoors if experiencing symptoms.

Primary Pollutant

Nitrogen Oxides (NO₂)

Nitrogen oxides typically come from traffic, fires and power stations.

Research Questions

- 1. Which months have the best and worst air quality?**
- 2. Which communities have the best and worst levels of air quality?**
- 3. What factors contribute most to poor air quality in Calgary or specific communities?**
- 4. Can the model predict future Air Quality?**



Dataset

Dependent Variable

Air Quality Index (AQI)

- Quantitative
- Standardized scale from 1 to 10

=

Seasons

- Categorical
- Summer
- Winter
- Fall
- Spring

+

\

Regions

- Categorical
- North West
- South East
- North East

+

Nine Pollutants

- Quantitative
- Parts per billion, et cetera.
- PM2.5 Mass
- Carbon Monoxide
- Nitrogen Dioxide
- Nitric Oxide
- Ozone
- Methane
- Total Hydrocarbons
- Total Oxides of Nitrogen
- Non-methane Hydrocarbons

Source: City of Calgary

Additional Information:

- 2015–2018: Most consistent pollutant data (3 years).
- ~15% missing data. Imputed with median across regions and seasons.

Approach & Main Results



Approach (Model Building)

1. **Best Additive Model:** Use All-Possible-Regressions Selection Procedure and Stepwise Selection Procedure to find the optimal model.
2. **Multicollinearity Check:** Test for Multicollinearity to identify and remove redundant predictors.
3. **Assumptions Check:** Linearity, Independence, Normality, and Heteroscedasticity.
4. **Addressing Assumption Violations (Normality and Heteroscedasticity):** Transform model and remove outliers.
5. **Transformation and Model Updates:** For each transformation or addition of a new feature, re-check assumptions to ensure they still hold.



Model Selection

| | | | | |
|--|------------------------|--|--------------------------------|------------------|
| | Full Model | Station Name + Season + Methane + Nitric Oxide + Nitrogen Dioxide + Ozone + Non-methane Hydrocarbons + PM2.5 Mass + Total Hydrocarbons + Total Oxides of Nitrogen | | |
| All-Possible-Regressions Selection Procedure <i>(Adjusted R-squared)</i> | All | | ANOVA: <i>(0.05)</i> | Pr(>F) 0.1024 |
| | Without Seasons | | | |
| Stepwise Selection Procedure <i>(0.05 - 0.1)</i> | | | | |
| | Reduced Model: | Station Name + Season + Methane + Nitric Oxide + Nitrogen Dioxide + Ozone + Non-methane Hydrocarbons + PM2.5 Mass + Total Hydrocarbons + Total Oxides of Nitrogen | | |



Multicollinearity

| | VIF | detection |
|-------------------------------|------------|-----------|
| Station.NameCalgary Northwest | 1.6912 | 0 |
| Station.NameCalgary Southeast | 2.4372 | 0 |
| Carbon.Monoxide | 4.2789 | 0 |
| Methane | 30886.5359 | 1 |
| Nitric.Oxide | 159.2768 | 1 |
| Nitrogen.Dioxide | 71.9613 | 1 |
| Ozone | 1.7358 | 0 |
| Non.methane.Hydrocarbons | 2470.4887 | 1 |
| PM2.5.Mass | 1.7077 | 0 |
| Total.Hydrocarbons | 40544.7424 | 1 |
| Total.Oxides.Of.Nitrogen | 395.8184 | 1 |

Total Oxides of Nitrogen includes
Nitric Oxide and **Nitrogen Dioxide**

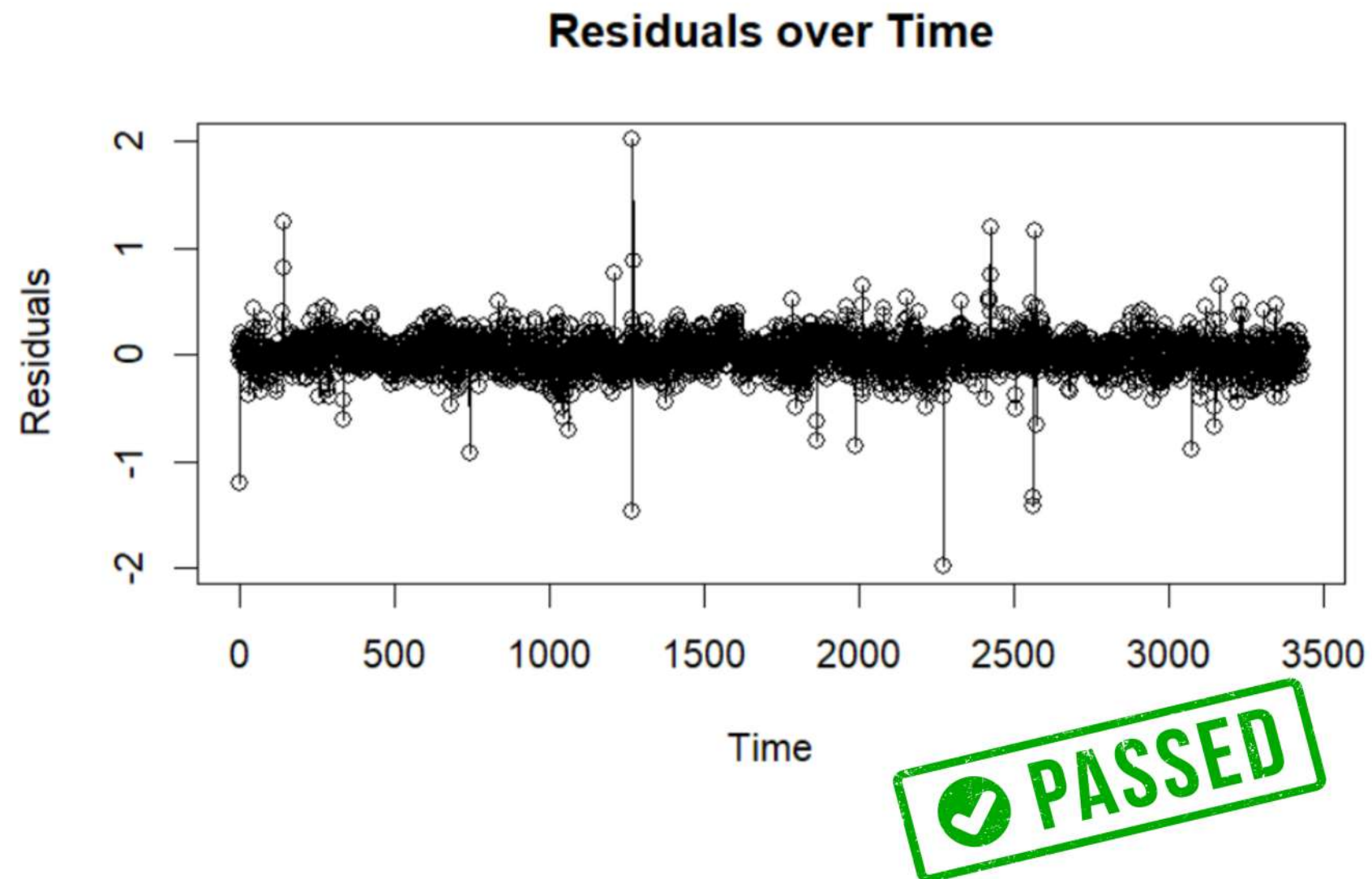
Total Hydrocarbon includes
Methane and **Non-methane**
Hydrocarbons

New Best Additive Model:

Station Name + **Season** + Methane + Nitric Oxide + Nitrogen
Dioxide + Ozone + Non-methane Hydrocarbons + PM2.5
Mass + **Total Hydrocarbons** + **Total Oxides of Nitrogen**

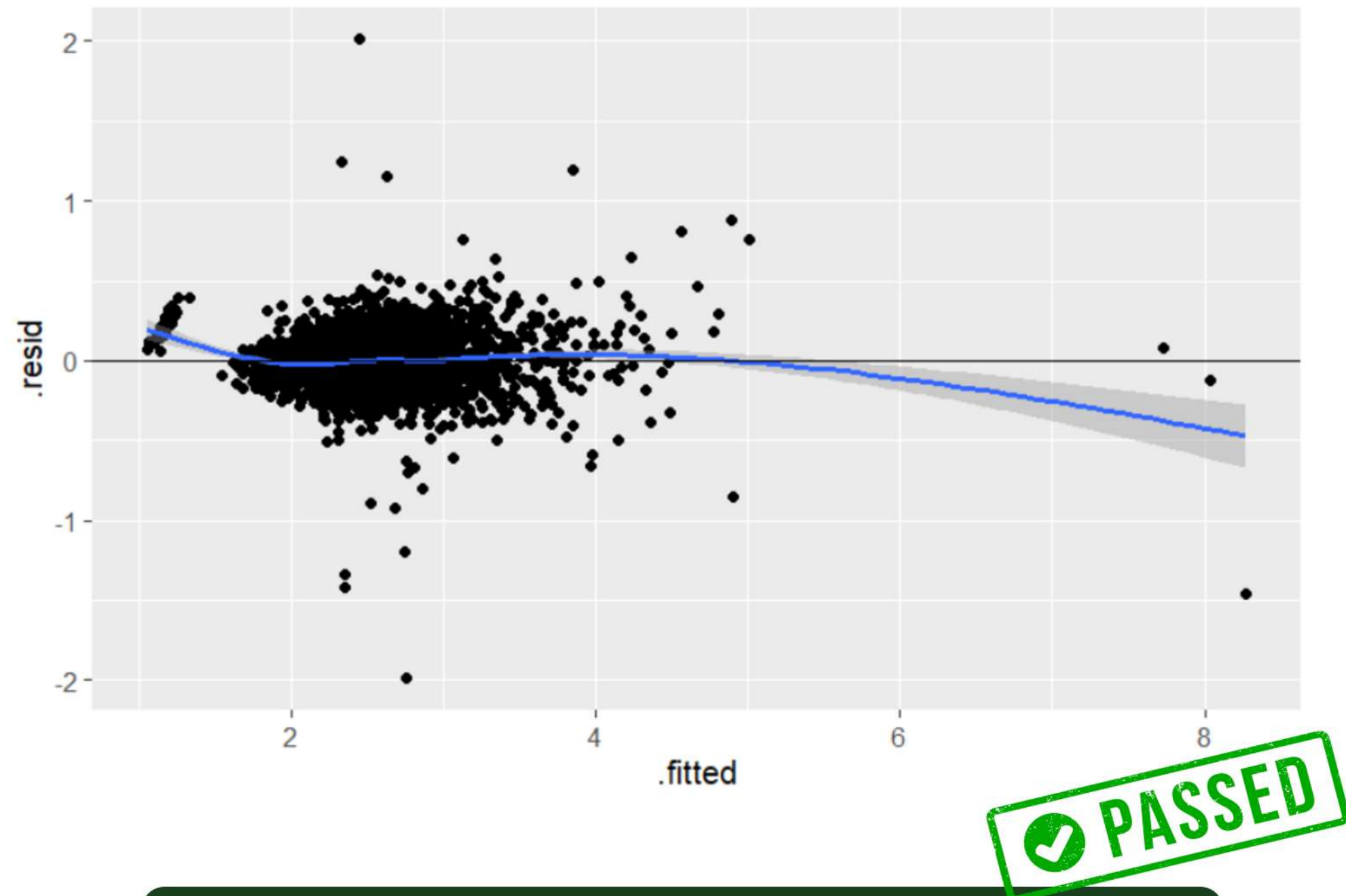
Checking Assumptions

Independence



Residuals are randomly scattered around zero with no pattern

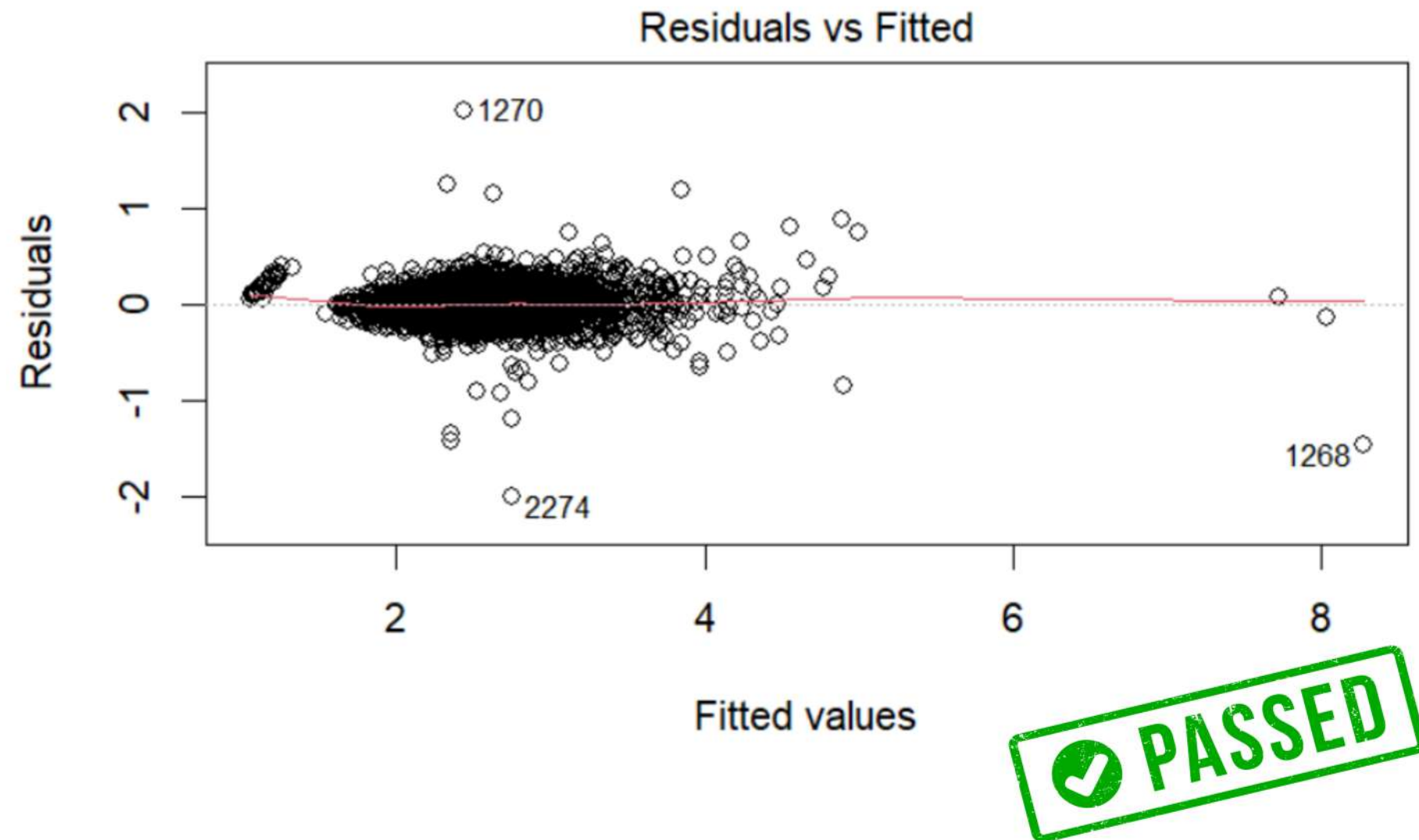
Linearity



No discernable pattern, slight curvature.

Checking Assumptions

Heteroscedasticity



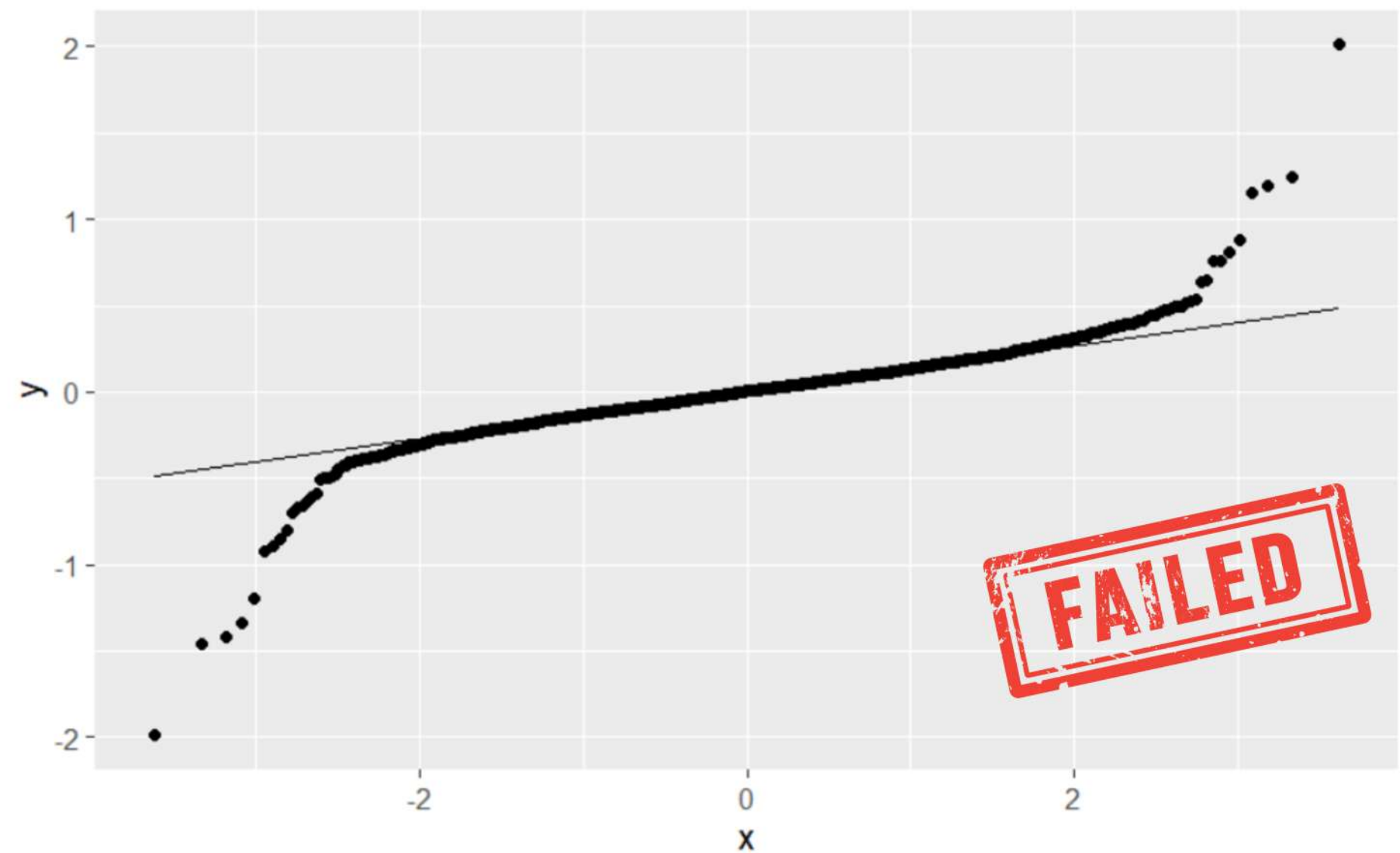
Quite horizontal, No funneling

Breush-Pagen Test:

p-value $< 2.2e-16$



Normality



Does not follow the line completely.

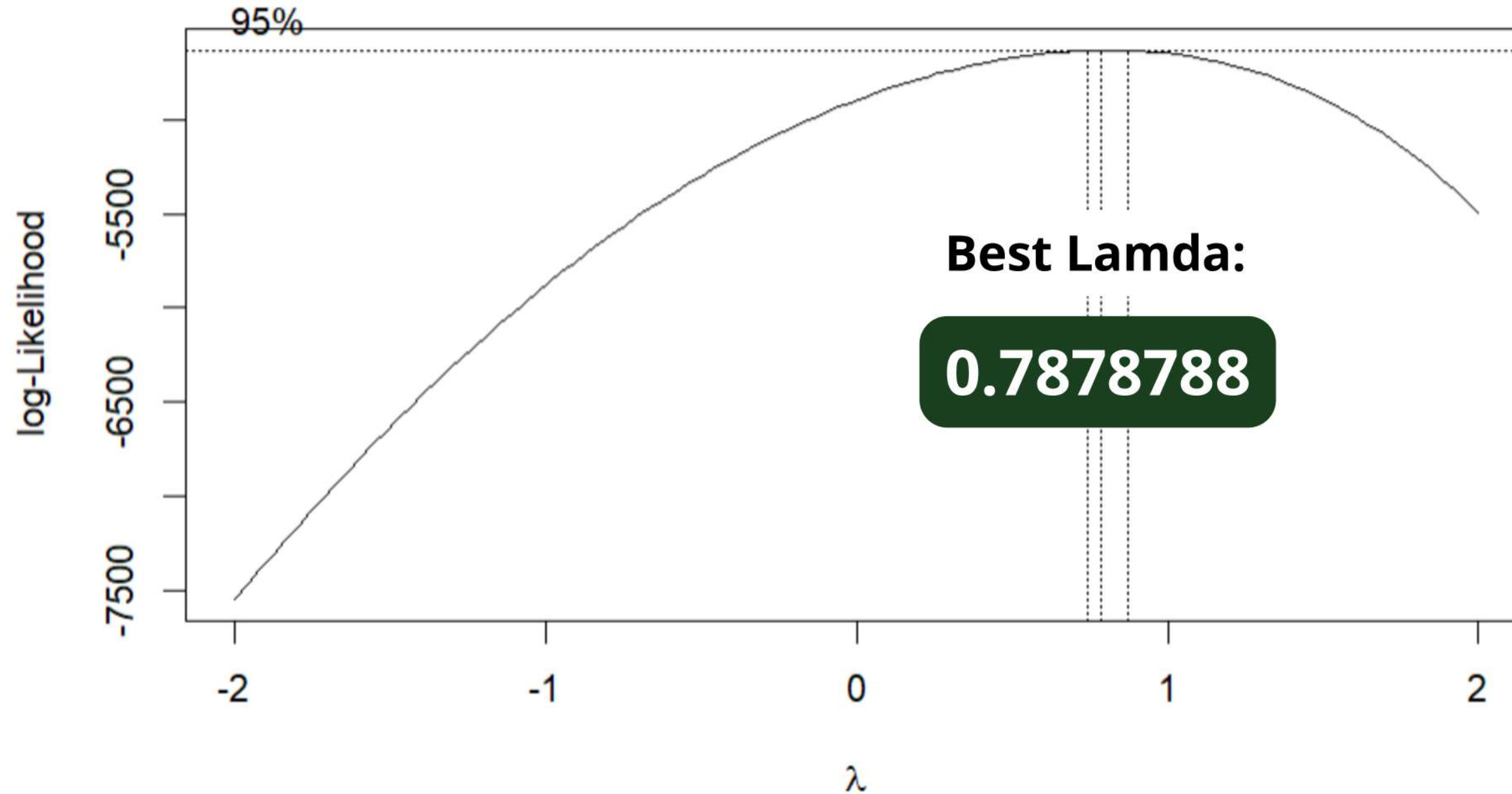
Sharpio-Wilk Test:

p-value $< 2.2e-16$



Addressing Assumption Violations:

BOXCOX TRANSFORMATION:



Breush-Pagen Test:

p-value $< 2.2e-16$



Sharpio-Wilk Test:

p-value $< 2.2e-16$



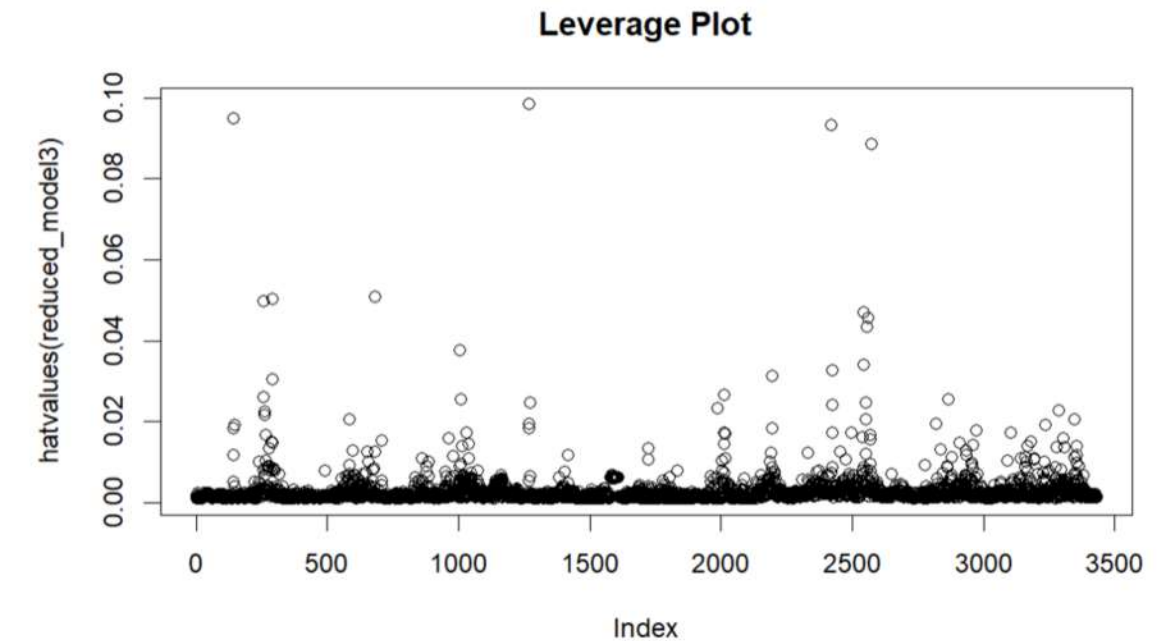
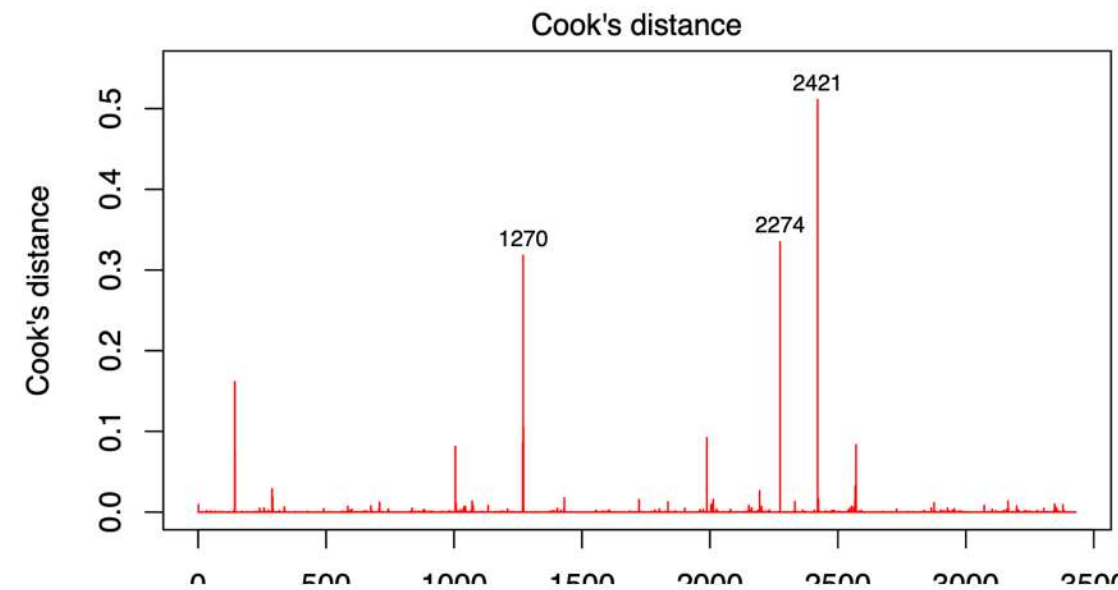
Addressing Assumption Violations and Model Updates

Influential Outliers:

Created a Cook's Distance Plot and Leverage Plot to identify the influential outliers and removed them, then we re-ran our assumption tests.

Cook's Distance Plot

Leverage Plot



Interaction Terms:

Made a model starting with all possible interaction terms. Removed insignificant interaction terms, re-ran our assumption tests

Higher Order Terms:

Made a model with higher order terms based on the gg-pairs plots and re-ran our assumption tests.

ALL FAILED NORMALITY AND HETEROSCEDASTICITY



Solution: Robust Regression Model

- What is it?**
- Use when assumptions are **violated—outliers or heteroscedasticity or Normality.**
 - Provide **reliable estimates** in the presence of assumption violations
 - Flexible approach that remains **effective when traditional regression methods would fail.**

- What it does?**
- **Minimizes the influence of outliers and violations of assumptions to provide reliable model estimates** in less-than-ideal data conditions.
 - It achieves this by using **weighting schemes that reduce the impact of extreme values** during the model fitting process.



Robust Sub-Models (Final)

Submodel for Station.NameCalgary Northwest:

$$\begin{aligned}\hat{Y}_{Air.Quality.Index} = & 1.1939 - 0.6793X_{Carbon.Monoxide} - 0.1380X_{Methane} + 0.07381X_{I(Methane^2)} \\ & + 7.583X_{Nitric.Oxide} + 34.16X_{Nitrogen.Dioxide} + 31.123X_{Ozone} + 315.9X_{I(Ozone^2)} + \\ & 1.5185X_{Non.methane.Hydrocarbons} + 0.0069X_{PM2.5.Mass} - 27.94X_{Carbon.Monoxide}X_{Nitric.Oxide} + \\ & 31.91X_{Carbon.Monoxide}X_{Nitrogen.Dioxide} + 0.07115X_{Carbon.Monoxide}X_{PM2.5.Mass} - 7.715X_{Methane}X_{Nitrogen.Dioxide} + \\ & 417.0X_{Nitric.Oxide}X_{Ozone} - 508.1X_{Nitrogen.Dioxide}X_{Ozone}\end{aligned}$$

Submodel for Station.NameCalgary Northeast:

$$\begin{aligned}\hat{Y}_{Air.Quality.Index} = & 1.067 + 0.5202X_{Carbon.Monoxide} - 0.1380X_{Methane} + 0.07381X_{I(Methane^2)} \\ & + 7.671X_{Nitric.Oxide} + 44.78X_{Nitrogen.Dioxide} + 33.51X_{Ozone} + 315.9X_{I(Ozone^2)} \\ & + 0.02846X_{Non.methane.Hydrocarbons} + 0.01117X_{PM2.5.Mass} - 27.94X_{Carbon.Monoxide}X_{Nitric.Oxide} + \\ & 31.91X_{Carbon.Monoxide}X_{Nitrogen.Dioxide} + 0.07115X_{Carbon.Monoxide}X_{PM2.5.Mass} - 7.715X_{Methane}X_{Nitrogen.Dioxide} + \\ & 417.0X_{Nitric.Oxide}X_{Ozone} - 508.1X_{Nitrogen.Dioxide}X_{Ozone}\end{aligned}$$

Submodel for Station.NameCalgary Southeast

$$\begin{aligned}\hat{Y}_{Air.Quality.Index} = & 1.1250 - 0.0674X_{Carbon.Monoxide} - 0.1380X_{Methane} + \\ & 0.07381X_{I(Methane^2)} + 11.093X_{Nitric.Oxide} + 33.73X_{Nitrogen.Dioxide} + 30.206X_{Ozone} + \\ & 315.9X_{I(Ozone^2)} + 0.21416X_{Non.methane.Hydrocarbons} + 0.01105X_{PM2.5.Mass} - \\ & 27.94X_{Carbon.Monoxide}X_{Nitric.Oxide} + 31.91X_{Carbon.Monoxide}X_{Nitrogen.Dioxide} + 0.07115X_{Carbon.Monoxide}X_{PM2.5.Mass} - \\ & 7.715X_{Methane}X_{Nitrogen.Dioxide} + 417.0X_{Nitric.Oxide}X_{Ozone} - 508.1X_{Nitrogen.Dioxide}X_{Ozone}\end{aligned}$$



Key Findings/Conclusion

1. Which months have the best and worst air quality?

Seasonality did not appear significant in predicting air quality, as it was excluded from our best additive model.

2. Which communities have the best and worst levels of air quality?

Challenging to determine which communities have the best and worst levels of air quality in general - due to the **complexity of the model**. Generally, Northwest Calgary had lower coefficients, and Northeast Calgary had higher coefficients.

3. What factors contribute most to poor air quality in Calgary or specific communities?

Carbon Monoxide, Methane, Nitric Oxide, Nitrogen Dioxide, Ozone, Non-methane Hydrocarbons, and PM2.5 Mass.

4. Can the model predict future Air Quality?

The **robust model is a strong choice for forecasting (Adjusted R-squared ~94%)**, especially if outliers are a concern or if similar irregularities are expected in future data.

Additional insights



```
# Print the confusion matrix  
print(confusion_matrix)
```

```
##  
## Actual  
## Predicted Low Risk Moderate Risk High Risk  
## Low Risk 2847 78 0  
## Moderate Risk 74 428 0  
## High Risk 0 0 3
```

```
# Print the confusion matrix  
print(confusion_matrix)
```

```
##  
## Actual  
## Predicted Low Risk Moderate Risk High Risk  
## Low Risk 2784 75 0  
## Moderate Risk 62 339 0  
## High Risk 0 0 0
```




Thank You!

Any Questions?



Let's clear the air, one insight at a time, for healthier
lives and a greener tomorrow!