

Project Report: Airfoil self-noise data analysis

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Dataset:- <https://archive.ics.uci.edu/dataset/291/airfoil+self+noise>

Research Questions

The study investigates the **factors influencing airfoil self-noise** using an experimental dataset derived from wind tunnel tests. The main research questions are:

1. Which features significantly affect the **Scaled Sound Pressure Level (SPL)** of an airfoil?
2. Can **multiple linear regression (MLR)** accurately predict SPL using aerodynamic and geometric parameters?
3. Do **feature interactions** and **non-linear effects** improve predictive performance compared to linear models?
4. Are there **dependencies (multicollinearity)** among predictors that could affect model stability?

Significance of the Problem

Understanding airfoil noise is crucial for designing quieter and more efficient aircraft. Aerodynamic noise contributes significantly to environmental sound pollution, particularly near airports. By modeling SPL as a function of key airfoil and flow parameters, this study provides statistical insights that can help in optimizing airfoil geometry and operating conditions to minimize noise.

Data Description

The dataset comprises **1,503 experimental observations** collected in a **wind tunnel** under controlled conditions, making it an **experimental dataset** rather than an observational one

Features

1. **Frequency (Hz)**: Number of oscillations of airflow over the airfoil per second.
2. **Angle of Attack (°)**: Angle between the oncoming flow and the chord line.
3. **Chord Length (m)**: Distance from the leading edge to the trailing edge.
4. **Free-Stream Velocity (m/s)**: Velocity of undisturbed airflow upstream of the airfoil.
5. **Suction Side Displacement Thickness (m)**: Thickness of the boundary layer on the suction side, affecting flow separation.

Response Variable

- **Scaled Sound Pressure Level (SPL)** measured in decibels (dB), representing the acoustic energy radiated from the airfoil surface.

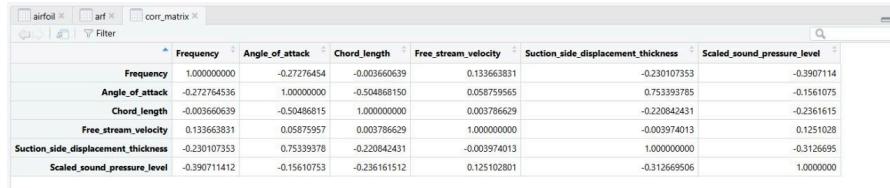
Preliminary Studies

Descriptive Statistics:

Frequency	Angle_of_attack	Chord_length	Free_stream_velocity	Suction_side_displacement_thickness	Scaled_sound_pressure_level
Min. : 200	Min. : 0.000	Min. : 0.0254	Min. :31.70	Min. :0.0004007	Min. :103.4
1st Qu.: 800	1st Qu.: 2.000	1st Qu.:0.0508	1st Qu.:39.60	1st Qu.:0.0025351	1st Qu.:120.2
Median : 1600	Median : 5.400	Median :0.1016	Median :39.60	Median :0.0049574	Median :125.7
Mean : 2886	Mean : 6.782	Mean :0.1365	Mean :50.86	Mean :0.0111399	Mean :124.8
3rd Qu.: 4000	3rd Qu.: 9.900	3rd Qu.:0.2286	3rd Qu.:71.30	3rd Qu.:0.0155759	3rd Qu.:130.0
Max. :20000	Max. :22.200	Max. :0.3048	Max. :71.30	Max. :0.0584113	Max. :141.0

Data Quality Checks

- Missing Values:** None found.
- Correlation Analysis:**



- Angle of Attack and Suction Side Displacement Thickness:** strong positive correlation (0.753).
- Angle of Attack and Chord Length:** moderate negative correlation (-0.505). These correlations indicate potential multicollinearity among predictors.

Statistical Analysis

Methods

Several regression techniques were used to model SPL:

- Multiple Linear Regression (MLR)** – baseline model to establish linear relationships.
- Model Selection** – forward, backward, and stepwise approaches using **AIC** and **BIC**.
- Polynomial Regression** – added second-order and interaction terms to capture non-linearity.
- Robust Regression** – to handle deviations from normality.
- Generalized Least Squares (GLS)** – to address heteroscedasticity and correlated errors.

Multiple Linear Regression:

$$SPL = \beta_0 + \beta_1 \times \text{Frequency} + \beta_2 \times \text{Angle of Attack} + \beta_3 \times \text{Chord Length} + \beta_4 \times \text{Free-Stream Velocity} + \beta_5 \times \text{Suction Side Displacement Thickness}$$

R-squared=0.5157, p-value<2e-16
(statistically significant)

```
lm(formula = Scaled_sound_pressure_level ~ Frequency + Angle_of_attack +
   Chord_length + Free_stream_velocity + Suction_side_displacement_thickness,
   data = arf)
```

Residuals:

Min	1Q	Median	3Q	Max
-17.480	-2.882	-0.209	3.152	16.064

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.328e+02	5.447e-01	243.87	<2e-16 ***
Frequency	-1.282e-03	4.211e-05	-30.45	<2e-16 ***
Angle_of_attack	-4.219e-01	3.890e-02	-10.85	<2e-16 ***
Chord_length	-3.569e+01	1.630e+00	-21.89	<2e-16 ***
Free_stream_velocity	9.985e-02	8.132e-03	12.28	<2e-16 ***
Suction_side_displacement_thickness	-1.473e+02	1.501e+01	-9.81	<2e-16 ***

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

Residual standard error: 4.809 on 1497 degrees of freedom
Multiple R-squared: 0.5157, Adjusted R-squared: 0.5141
F-statistic: 318.8 on 5 and 1497 DF, p-value: < 2.2e-16

Summary of the model:

Diagnostic Analysis:

1) Multicollinearity Check:

As discussed in correlation part the, VIF (Variance Inflation Factor) was calculated

Frequency 1.144444	Angle_of_attack 3.441658	Chord_length 1.510754	Free_stream_velocity 1.041698	Suction_side_displacement_thickness 2.532127
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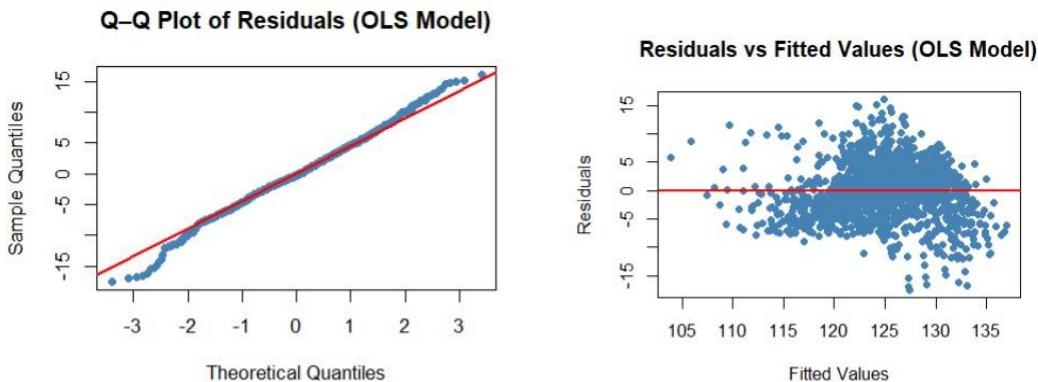
Although the values were not extremely high, there was some indication of multicollinearity. The highest VIF value was associated with Angle of Attack (VIF = 3.44).

After removing AoA:

Frequency 1.079606	Chord_length 1.054873	Free_stream_velocity 1.019099	Suction_side_displacement_thickness 1.114675
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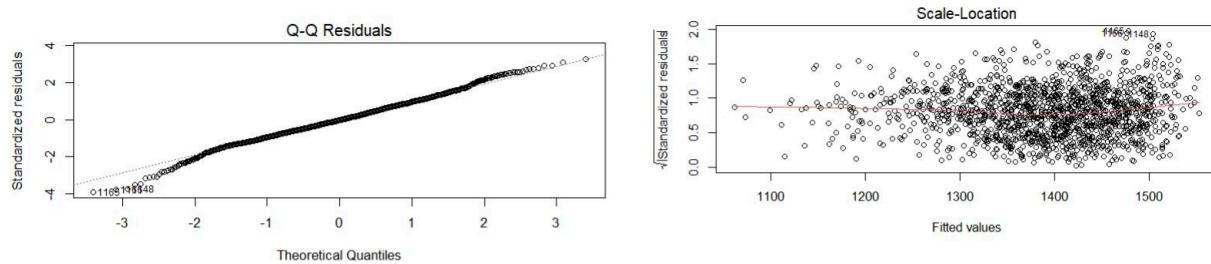
```
Call:  
lm(formula = Scaled_sound_pressure_level ~ Frequency + Chord_length +  
    Free_stream_velocity + Suction_side_displacement_thickness,  
    data = arf)  
  
Residuals:  
    Min      1Q  Median      3Q      Max  
-19.866 -3.109 -0.018  3.332 15.932  
  
Coefficients:  
            Estimate Std. Error t value Pr(>|t|)  
(Intercept) 1.304e+02 5.131e-01 254.03 <2e-16 ***  
Frequency   -1.174e-03 4.246e-05 -27.64 <2e-16 ***  
Chord_length -2.597e+01 1.414e+00 -18.36 <2e-16 ***  
Free_stream_velocity 8.686e-02 8.351e-03 10.40 <2e-16 ***  
Suction_side_displacement_thickness -2.692e+02 1.034e+01 -26.02 <2e-16 ***  
---  
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
  
Residual standard error: 4.993 on 1498 degrees of freedom  
Multiple R-squared:  0.4776, Adjusted R-squared:  0.4763  
F-statistic: 342.4 on 4 and 1498 DF,  p-value: < 2.2e-16
```

Q-Q Plot of Residuals and Residual vs Fitted::



There are noticeable deviations at the tails, suggesting that the residuals depart from normality at both ends. The plot reveals that as the fitted values increase, the spread of the residuals also increases, which violates the assumption of homoscedasticity in the OLS model.

After using box-cox transformation:



Polynomial Regression

Second-order polynomial and interaction terms improved model fit:

$$SPL = \beta_0 + \beta_1 \times \text{Free_Stream_Velocity} \\ + \beta_2 \times (\text{Angle_of_Attack})^2 \\ + \beta_3 \times (\text{Chord_Length})^2 \\ + \beta_4 \times (\text{Frequency} \times \text{Chord_Length}) \\ + \beta_5 \times (\text{Frequency} \times \text{Suction_Side_Displacement_Thickness}) \\ + \beta_6 \times (\text{Chord_Length} \times \text{Angle_of_Attack})$$

R² = 0.6192, RSS = 4.266

This model captured mild non-linear effects observed in residual plot

```
Residuals:
    Min      1Q   Median      3Q     Max 
-17.4767 -2.5365 -0.2127  2.9646 16.1672 

Coefficients:
                                         Estimate Std. Error t value Pr(>|t|)    
(Intercept)                         1.261e+02  4.097e-01 307.915 < 2e-16 ***
Free_stream_velocity                 9.575e-02  7.192e-03 13.314 < 2e-16 ***
I(Angle_of_attack^2)                -1.074e-02 1.268e-03 -8.468 < 2e-16 ***
I(Chord_Length^2)                   -2.141e+01  4.791e+00 -4.469 8.44e-06 ***
Frequency:Chord_length              -5.818e-03  2.329e-04 -24.976 < 2e-16 ***
Frequency:Suction_side_displacement_thickness -7.638e-02 3.579e-03 -21.340 < 2e-16 ***
Chord_length:Angle_of_attack       -1.092e+00  2.412e-01 -4.528 6.43e-06 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 4.266 on 1496 degrees of freedom
Multiple R-squared:  0.6192, Adjusted R-squared:  0.6177 
F-statistic: 405.4 on 6 and 1496 DF,  p-value: < 2.2e-16
```

Robust Regression

Applied to mitigate the influence of outliers:

- Improved residual distribution symmetry.
- Reduced sensitivity to heteroscedasticity, though R² slightly lower than polynomial regression.

```

Call: rlm(formula = Scaled_sound_pressure_level ~ Frequency + Chord_length +
  Free_stream_velocity + Suction_side_displacement_thickness,
  data = arf)
Residuals:
    Min      1Q  Median      3Q     Max 
-20.37126 -3.02783 -0.02526  3.21111 15.86943 

Coefficients:
                Value Std. Error t value
(Intercept) 131.1444   0.4891 268.1483
Frequency   -0.0013   0.0000 -32.5707
Chord_length -27.7069   1.3482 -20.5516
Free_stream_velocity 0.0879   0.0080 11.0445
Suction_side_displacement_thickness -285.6688   9.8579 -28.9788

Residual standard error: 4.634 on 1498 degrees of freedom

```

Generalized Least Squares (GLS)

Used to correct for non-constant variance and correlated residuals:

- Provided stable coefficient estimates.
- Improved model diagnostics with more homoscedastic residuals

```

Generalized least squares fit by REML
Model: Scaled_sound_pressure_level ~ Frequency + Chord_length + Free_stream_vel-
  ocity + Suction_side_displacement_thickness
Data: arf
AIC      BIC    logLik 
7533.388 7575.883 -3758.694

Correlation Structure: AR(1)
Formula: ~1
Parameter estimate(s):
0.8214131

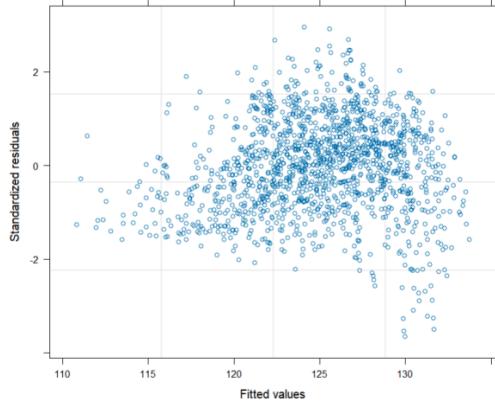
Variance function:
Structure: Power of variance covariate
Formula: ~Suction_side_displacement_thickness
Parameter estimates:
power
0.05803285

Coefficients:
            Value Std. Error t-value p-value
(Intercept) 128.61172 0.985538 130.49898 0
Frequency   -0.00086 0.000030 -28.85955 0
Chord_length -27.59020 1.348202 -20.55162 0
Free_stream_velocity -0.0178 0.011397 -8.93017 0
Suction_side_displacement_thickness -203.14591 26.688630 -7.61170 0

Correlation:
          (Intr) Frqncy Chrd_l Fr_st_
Frequency -0.044
Chord_length -0.631  0.001
Free_stream_velocity -0.604 -0.118  0.026
Suction_side_displacement_thickness -0.313  0.074  0.059  0.039

Standardized residuals:
    Min      Q1       Med      Q3      Max 
-3.648500486 -0.696654735  0.006126964  0.665812342  2.947649013

```



Conclusion

- **Best Model:** The **Polynomial Regression Model** demonstrated the best accuracy, capturing non-linear relationships with improved R^2 (0.6192).
- **Interpretability:** The **transformed MLR model** (with multicollinearity correction and Box–Cox transformation) remains better for **interpretation** and **statistical inference**.
- **Broader Implications:**
The results suggest that statistical regression techniques can effectively model aerodynamic noise. This approach can inform future airfoil design and noise prediction frameworks in aeronautical engineering.
- **Future Work:** In future we can explore the use of **Ridge** and **Lasso regression** for regularization, along with **tree-based models** like **Random Forests** and **XGBoost** to capture complex relationships and improve predictive accuracy.

- 1. Frequency:** Higher frequencies cause acoustic energy to spread over a wider range, reducing intensity per band. Additionally, viscous damping is stronger at high frequencies, lowering overall SPL.
- 2. Angle of Attack:** A moderate increase in angle of attack smoothens airflow and reduces turbulence near the surface. This minimizes flow separation and weakens noise-producing pressure fluctuations.
- 3. Chord Length:** A longer chord length promotes smoother airflow and reduces vortex shedding at the trailing edge. This decreases broadband noise generation, leading to lower SPL.
- 4. Free-Stream Velocity:** As velocity increases, aerodynamic forces and unsteady pressure fluctuations intensify. Since noise power scales sharply with velocity, SPL rises significantly at higher speeds.
- 5. Suction Side Displacement Thickness:** A thicker boundary layer reduces the velocity gradient and suppresses surface turbulence. This lowers high-frequency noise production, thereby decreasing SPL.

Answers to Research Questions:

- 1) Almost all features are significantly important.
- 2) Yes, MLR can accurately predict SPL, after removing multicollinearity and using transformations.
- 3) Yes, the Polynomial regression model gives the best result as it captures the interactions better..
- 4) Yes, there was a multicollinearity issue due to the feature Angle of Attack which was highly correlated with Suction Side Displacement Thickness and moderately correlated with Chord Length. This variable was removed to address the issue.