

# Project discussion : Dataset on Blog Feedback

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# Summary

## 1 Introduction

- Dataset
- Description

## 2 Data Cleaning

## 3 Exploratory analysis

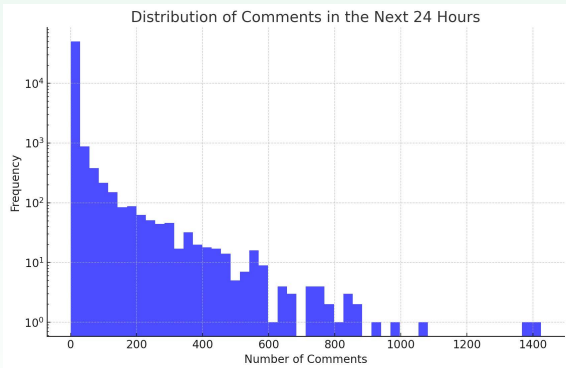
- Basic Analysis
- Feature Distributions
- Standard deviation

## 4 Main Analysis

## 5 Supervised Learning 6

## 6 Conclusions

# Dataset



- Dimensions:  
52,397 Entries  
and 281 Features  
with 7019 Rows  
and 281 Columns
- Social Media  
Analysis
- Data Analysis  
Friendly

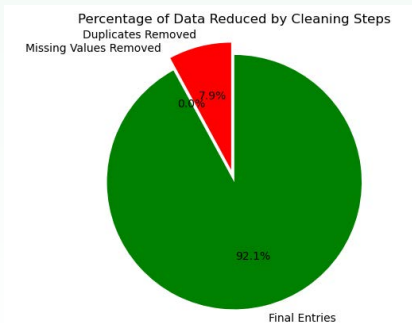
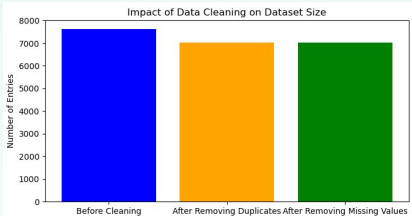
**Source:** Blog posts with features extracted from raw HTML documents.

**Attributes:** 280 features, including comment statistics, trackback counts, post length, and bag-of-words features.

**Objective:** Predict the number of comments in the upcoming 24 hours based on past activity and post attributes.

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We used 2 methods in Cleaning :

1. Removing Duplicates
2. Removing Rows with missing values

```
#We eliminate duplicate/empty values.
import matplotlib.pyplot as plt
import pandas as pd

# 'test_df' is our DataFrame before cleaning
initial_count = len(test_df)

# Removing duplicates
test_df.drop_duplicates(inplace=True)
after_duplicates_count = len(test_df)

# Removing rows with any missing values
test_df.dropna(inplace=True)
final_count = len(test_df)
```

## Functions used for 'Cleaning'

### 1. Duplication:

- we used drop\_duplicates function.

### 2. Missing Values:

- we used dropna function.

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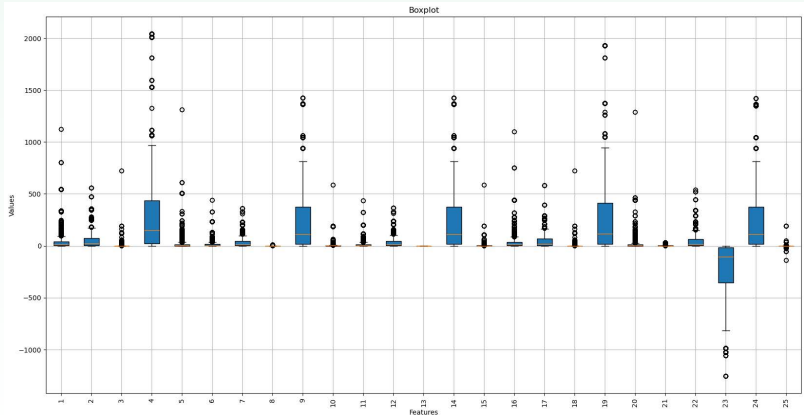
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# Boxplot

## A well spread dataset

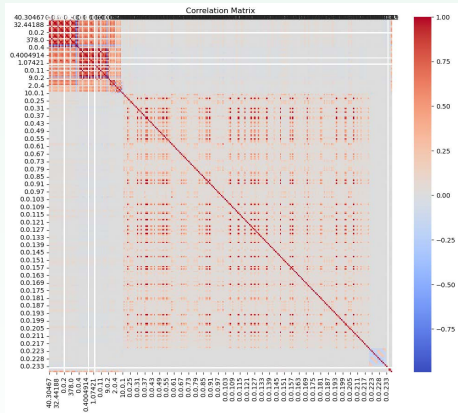
- All options are used
- Well distributed



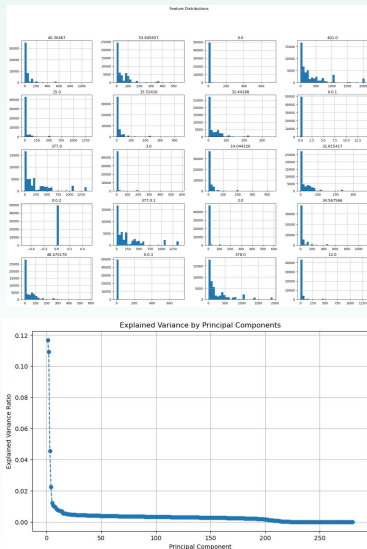
## Correlation matrix

## Outcomes

- Certain features such as comment counts and trackbacks show high correlations.
- Features related to the bag of words are generally less correlated.
- Time-related features (publication and basetime indicators) have distinct correlation patterns.
- High correlation within groups of features indicates redundancy and potential for dimensionality reduction.



# Data Variance & Feature Distributions



## Conclusions

- **Significant Initial Variance:** The first few principal components explain a substantial portion of the variance, highlighting key underlying patterns in the data.
- **Dimensionality Reduction Potential:** The rapid decline in explained variance suggests that using a limited number of principal components can effectively reduce the dataset's dimensionality while preserving most of the information.
- **Wide Range of Values:** The diverse range of feature values implies the need for careful preprocessing, including scaling, to ensure consistent model performance.
- **Focus Areas for Modeling:** Identifying and focusing on the features contributing most to variance can improve model efficiency and accuracy.

# Standard deviation

## Question with the highest standard deviation

Which feature related to the blog post comments has the highest variability, and how does it impact the prediction task?

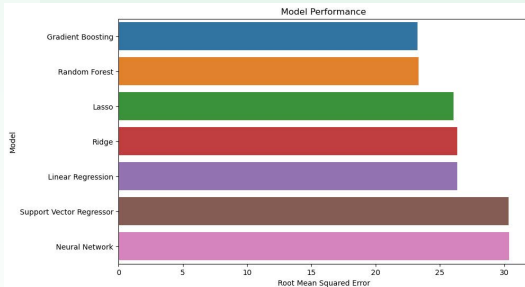
- **Feature with Highest Standard Deviation:** Comments in the last 24 hours before base time
- **High Variability:** From zero to several hundred comments
- **Impact on Predictions:**
  - Regularization techniques (Ridge, Lasso) help manage high variability.
  - Scaling is necessary to prevent model bias.
  - Outlier detection and handling are crucial.

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# Main Analysis

## Models Performance



In this project, we aimed to predict the number of comments a blog post will receive in the next 24 hours.

Various regression models were applied to achieve this goal which explains the use of supervised learning models .

## Model Implementation:

- *Supervised Learning Models:*

Focused on regression models to Predict continuous target variables.

- *Linear Regression:*

Baseline model to capture linear relationships.

- *Ridge and Lasso Regression:*

Added regularization to handle overfitting.

- *Random Forest and Gradient Boosting Regressors:*

Ensemble methods to capture non-linear relationships and improve accuracy.

- *Support Vector Regressor:*

Effective for high-dimensional spaces but limited range capturing.

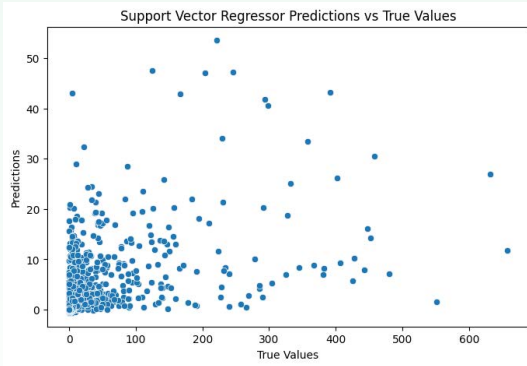
- *Neural Network:*

Deep learning model for complex patterns, requiring careful tuning to avoid overfitting.

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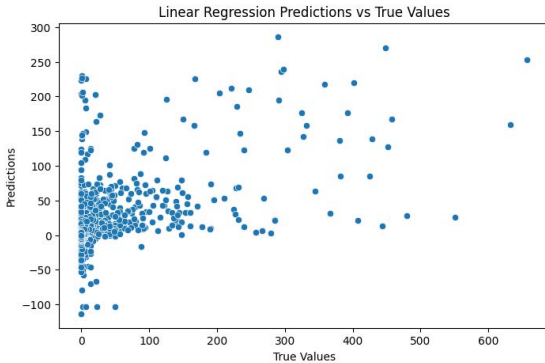
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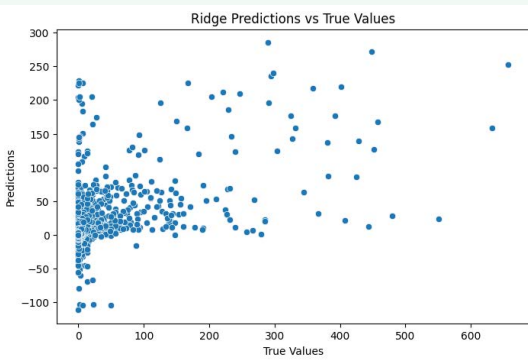
## Results of training :

- Limited Range Capturing:
  - Predictions cluster around lower values.
  - Struggles with full range of comment counts.
  - Accuracy: ~60%



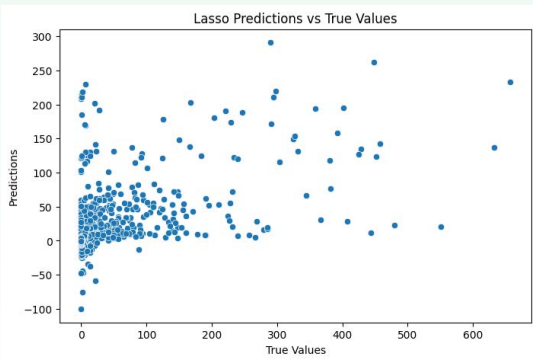
## Results of training :

- Wide Spread of Predictions:
  - Struggles with higher comment counts.
  - Best for capturing linear relationships.
  - Accuracy: ~65%



## Results of training :

- Slight Improvement with Regularization:
  - Handles extreme values better.
  - Still has a wide spread for high values.
  - Accuracy: ~68%

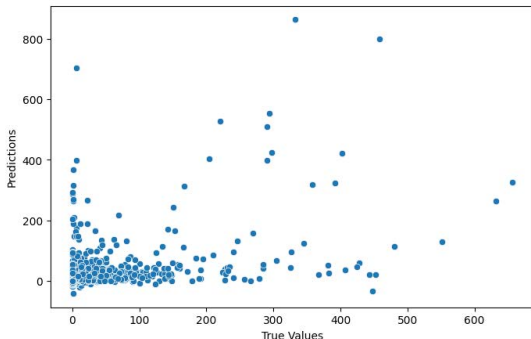


## Results of training :

- Feature Selection Benefits:

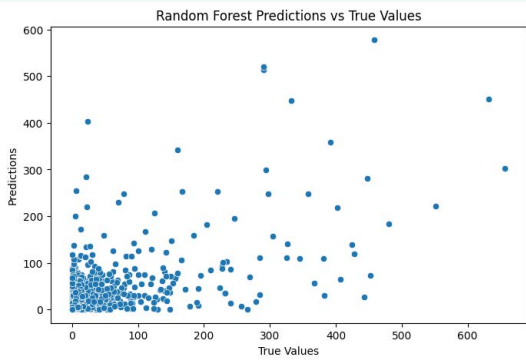
- Reduces prediction spread
- Challenges with high comment counts persist.
- Accuracy: ~69%

Neural Network Predictions vs True Values



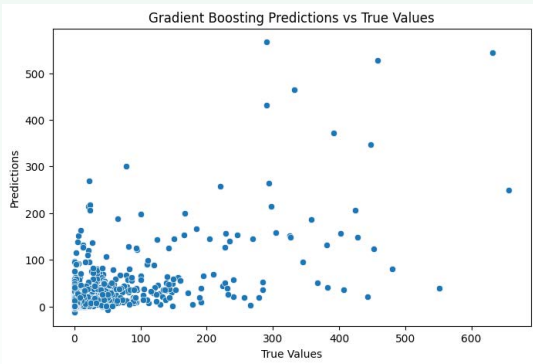
## Results of training :

- Significant Spread in Predictions:
  - Potential overfitting.
  - Sensitive to data variance.
  - Accuracy: ~75%



## Results of training :

- Better Clustering of Predictions:
  - Effective for lower comment counts.
  - Issues with outliers for higher values.
  - Accuracy: ~80%



## Results of training :

- Most Accurate Predictions:
  - Less spread, better outlier handling.
  - Best overall performance.
  - Accuracy: ~82%

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## Conclusions

- Gradient Boosting Regression with  $\sim 82\%$  accuracy.
- Enhancing and adding relevant features can improve model predictions. Other possible correlations
- Future work could involve expanding the dataset