

Zodiac Sign Prediction

Join Us on a Journey Through Stars and Algorithms!

Group 5

- Ahmed Abdulrahim
- Bimsara Geethachapa Siman Meru Pathiranage
- Simranjeet Kaur
- Efemena Theophilus Edoja
- Simran



Unveiling the essence

Astrology, a timeless art, has always intrigued humanity with its celestial insights. Our project takes this ancient wisdom into the digital age by harnessing Python programming and machine learning techniques to predict Zodiac signs.



The core objective of this project is to utilize machine learning and natural language processing methods to forecast the astrological signs of bloggers by analyzing their written content.



Methodology

OSEMN offers a structured approach that encompasses five essential stages, ensuring a comprehensive and effective exploration of the data:

OSEMN



Obtain

Scrub

Explore

Model

iNterpret

Obtain: Business Understanding

Businesses can unlock a range of valuable insights:

- Personalized Marketing
- Content Recommendation
- Product Development
- Customer Relationship Management
- Market Segmentation
- Competitive Edge
- Content Creation





Data Understanding

- Proper dataset comprehension ensures accurate insights and valid conclusions.
- Understanding data helps allocate resources efficiently, saving time and effort.
- Insights from dataset understanding lead to informed and confident decision-making.
- Recognizing biases in data allows for fairer and more ethical outcomes.
- Dataset understanding aligns models with real-world scenarios, improving their relevance.
- Identify potential risks associated with using the data, such as legal, ethical, or privacy concerns.
- Informed hypothesis based on understanding of dataset as a result enabling more effective analysis

- Our dataset originates from Kaggle, a respected platform known for sharing datasets within the data science and machine learning domain.
- Kaggle serves as a reliable hub where data enthusiasts share datasets, making valuable resources accessible to the community.
- For our zodiac sign prediction project, we've specifically chosen a dataset from Kaggle that aligns seamlessly with our objectives.

<https://www.kaggle.com/datasets/rtatman/blog-authorship-corpus>



The screenshot displays the Kaggle website interface. On the left is a navigation sidebar with the Kaggle logo and links for 'Create', 'Home', 'Competitions', 'Datasets' (which is highlighted), and 'Models'. The main content area features a search bar at the top right with 'Sign In' and 'Register' buttons. Below the search bar, the dataset 'Blog Authorship Corpus' by Rachael Tatman is shown, updated 6 years ago, with 136 votes. Action buttons include 'New Notebook', 'Download (304 MB)', and a settings menu. The dataset description states 'Over 600,000 posts from more than 19 thousand bloggers'. A thumbnail image of a laptop screen is visible on the right.

kaggle

Search

Sign In Register

+ Create

Home

Competitions

Datasets

Models

RACHAEL TATMAN · UPDATED 6 YEARS AGO

136

New Notebook

Download (304 MB)

Blog Authorship Corpus

Over 600,000 posts from more than 19 thousand bloggers



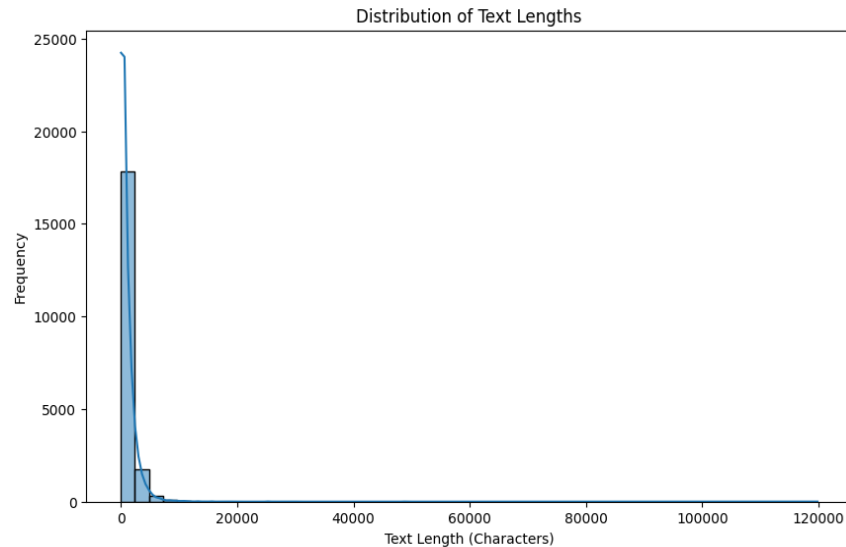
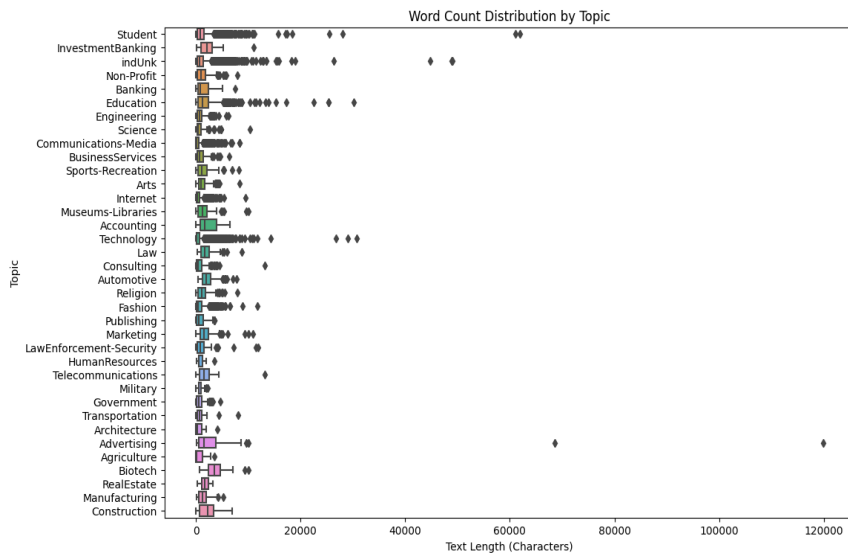
Why this dataset?

- **Data Source Authenticity:** Ensuring the authenticity of the data source is essential to maintain the integrity of the project.
- **Data Size and Diversity:** Provides a substantial amount of textual content for analysis.
- **Zodiac Sign Labels:** Fundamental requirement of our Project
- **Textual Content:** Helps identify writing patterns specific to each zodiac sign.
- **Age and Gender Distribution:** Offers potential insights into how writing styles may vary across demographics.
- **Data Preprocessing Potential:** Provides room for various preprocessing techniques, such as cleaning, tokenization, and lemmatization.
- **Ethical Considerations:** The dataset adheres to ethical guidelines, ensuring the privacy and anonymity of the bloggers.



Explore: Data Preparation

- Data Cleaning
- Text Preprocessing
- Label Encoding
- Feature Extraction



Data Preparation - Explore Phase

Analysis:

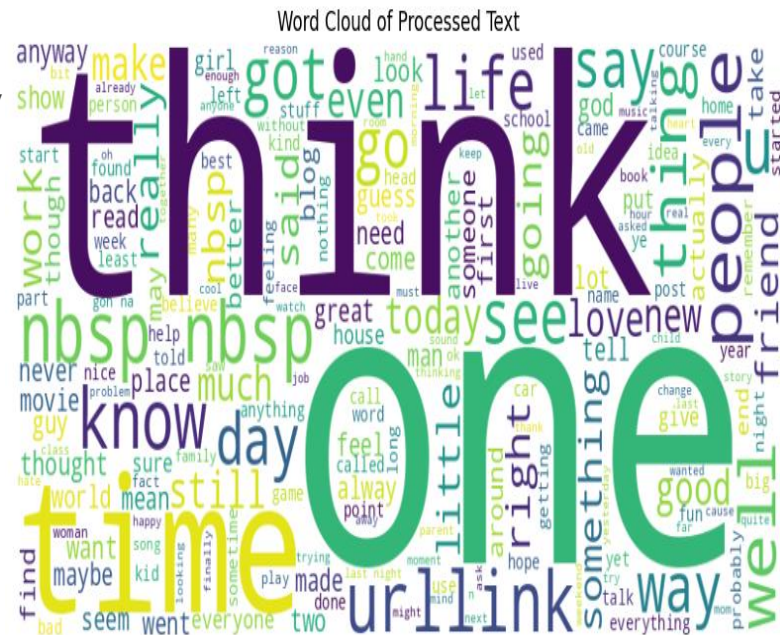
- Explore zodiac sign distribution and attribute summary statistics.
- Visualize data distributions and frequency of zodiac signs.

Feature Selection:

- Focus on significant words/phrases associated with specific signs.

Visualization:

- Create count plots for zodiac sign occurrences in the dataset.





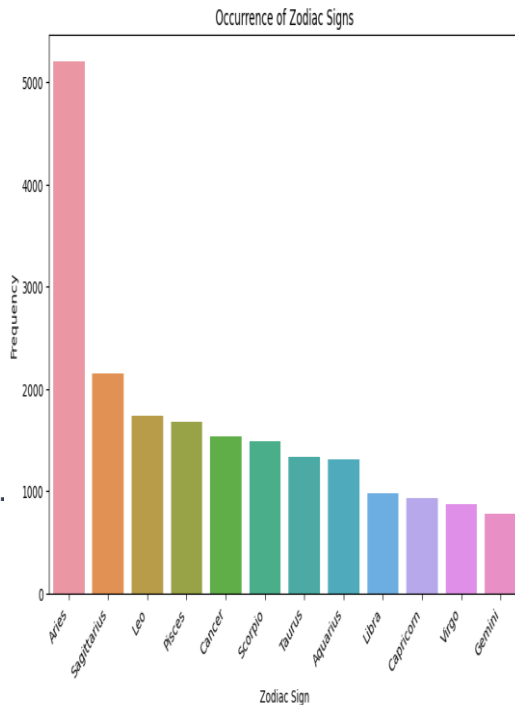
Impact of Data Preparation and Exploration

Significance:

- Ensures accurate, clean, and well-structured data for analysis.
- Identifies meaningful features to enhance model performance.
- Provides insights into zodiac sign distribution and patterns.

Contribution:

- Improves accuracy and effectiveness of Zodiac Sign Prediction models.
- Informs subsequent stages of model development and evaluation.
- Enables data-driven decision making for reliable predictions.





Model

Random Forest Classifier:

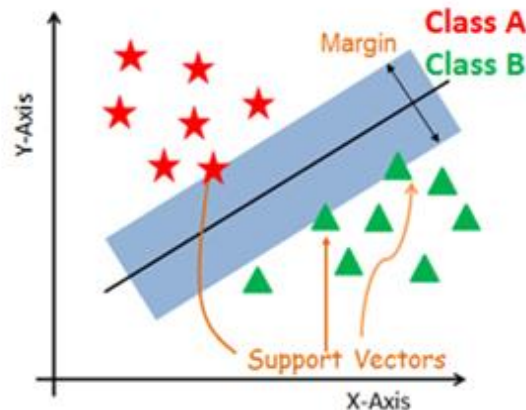
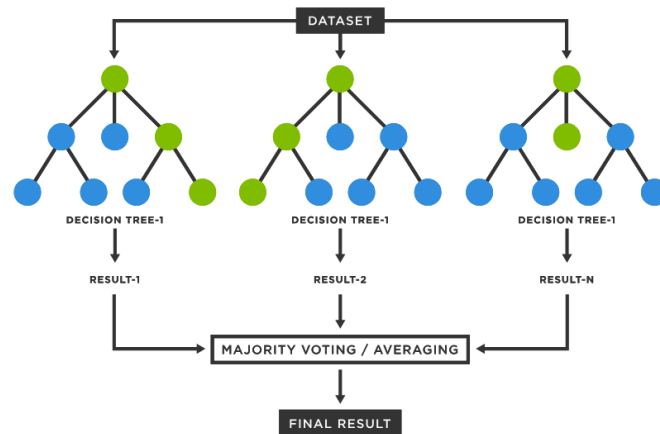
- Ensemble Learning
- Feature Importance
- Robustness

Why We Chose It.

- Ability to handle complex relationships
- Mitigate overfitting
- Provide feature importance analysis

Linear Support Vector Classifier:

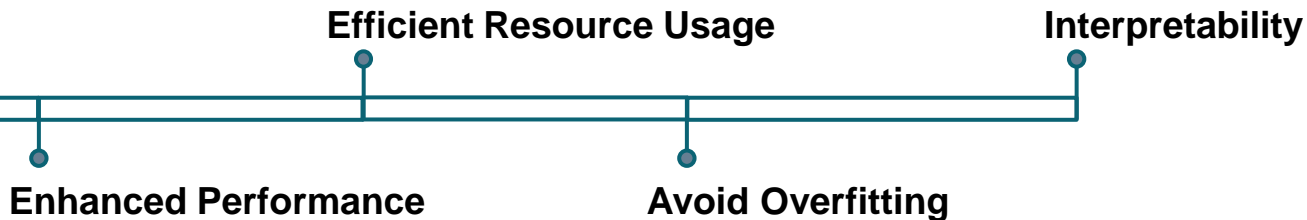
- Effective Separation
- Margin Maximization





Hyperparameter Tuning

Importance of Optimal Hyperparameters



Search	Parameters	Best Parameters
Randomized Search CV	<code>C: 0.01, 0.1, 1, 10</code> Regularization parameter <code>Loss: 'hinge', 'squared_hinge'</code> Loss function <code>Max_iter: randint(100, 500)</code> Maximum number of iterations	<code>C: 1</code> <code>loss: hinge</code> <code>Max_iter: 350</code> Best Accuracy: 0.5416
Grid Search	<code>C: 0.8, 0.9, 1, 1.1, 1.2</code> <code>loss: 'hinge'</code> (Use the best 'loss' value) <code>max_iter: 300, 350, 400</code>	<code>C: 1.2</code> <code>max_iter: 300</code> Best Accuracy: 0.5433



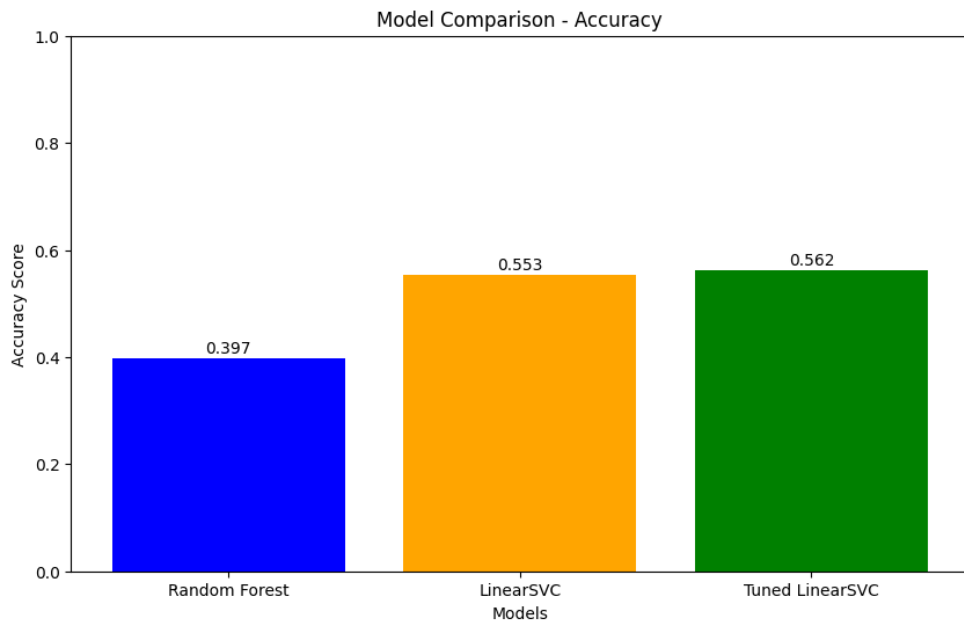


Interpretation

To measure performance
of each model we used

Accuracy score

Model	Random Forest	Linear SVC	Tuned Linear SVC
Accuracy score	0.397	0.553	0.562

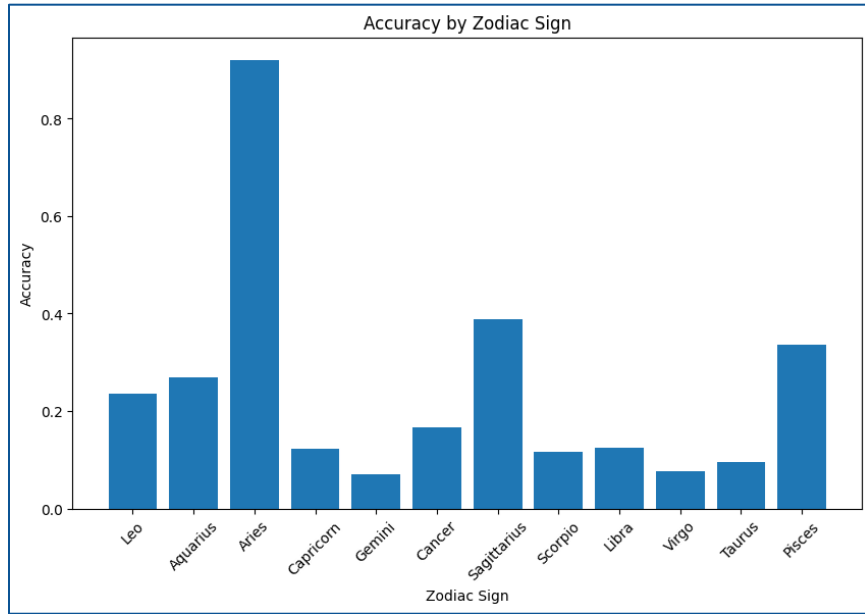


Random Forest Classifier

Top Accuracy: Aries

lowest accuracy: Gemini

Analyzing accuracy by zodiac sign provides valuable insights into the performance of our prediction model and suggests potential areas for improvement



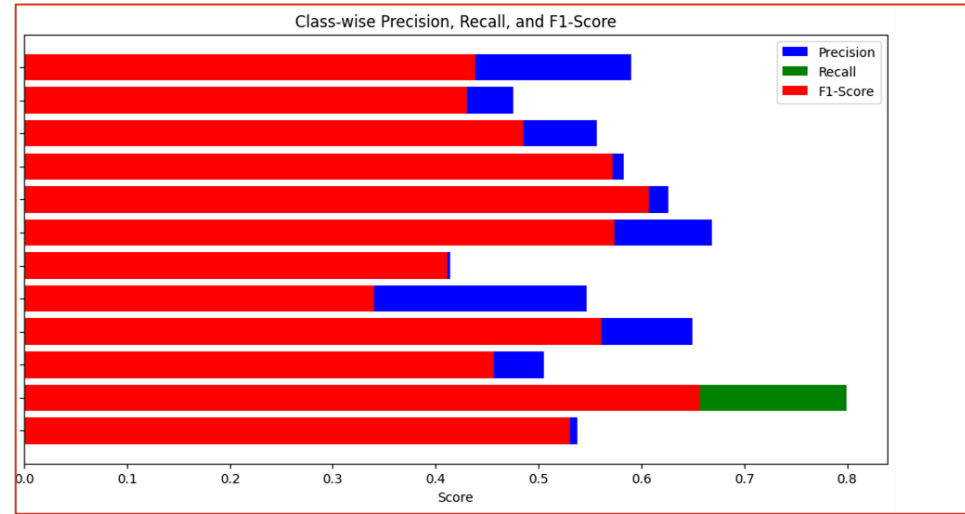
Linear SVC

Highest Precision: Gemini

Highest Recall: Aquarius

Highest F1-Score: Aquarius

Analyzing class-wise precision, recall, and F1-score provides a comprehensive understanding of our model's performance across different zodiac signs



To measure performance of each model we used

Classification report

Classification Report for Zodiac Sign Prediction:

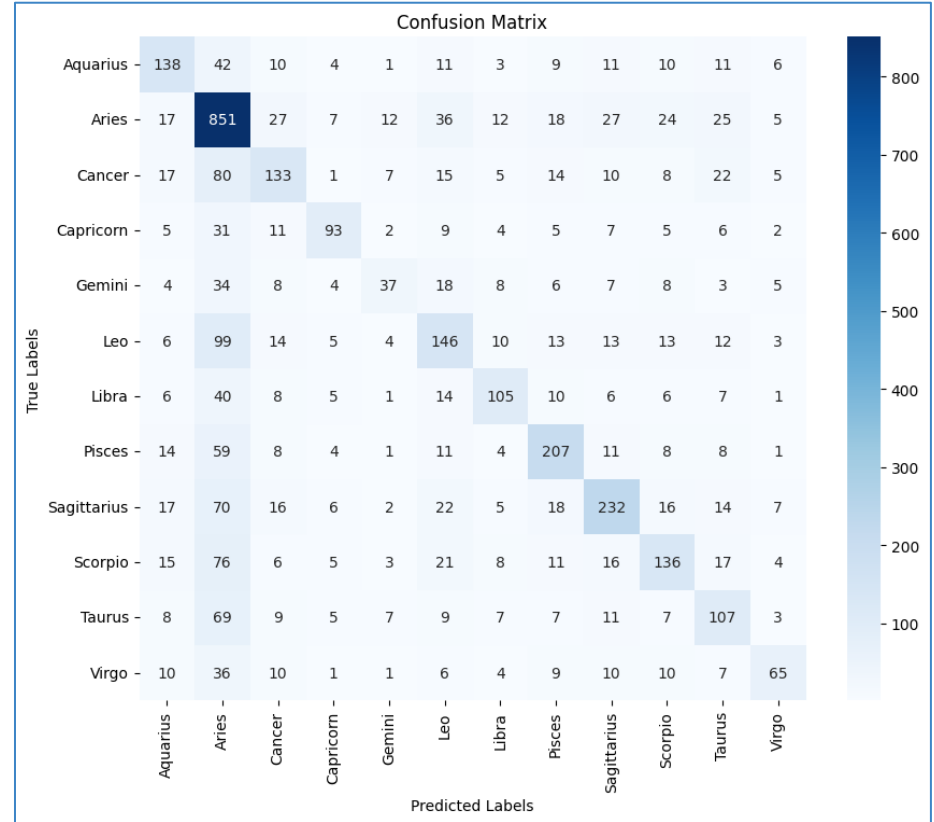
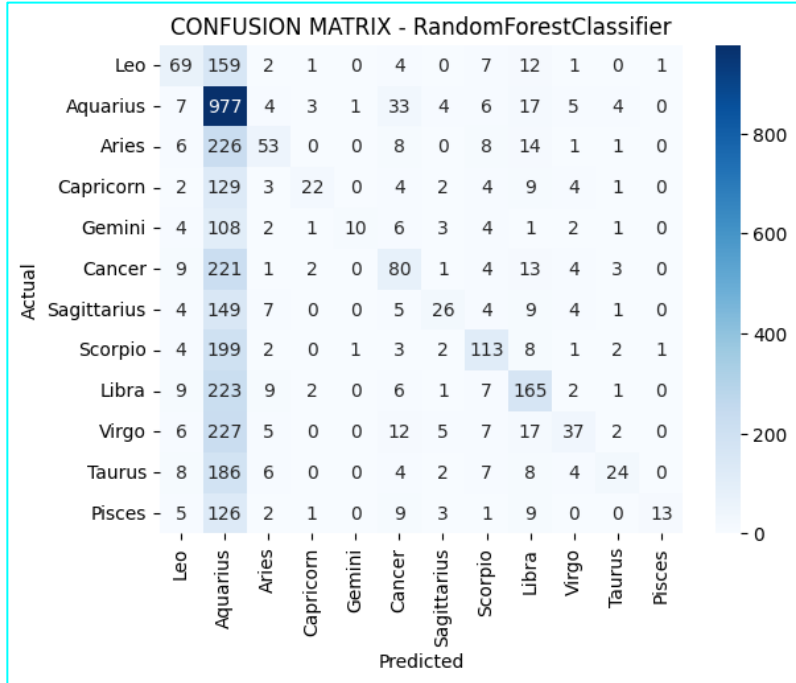
	precision	recall	f1-score	support
Aquarius	0.52	0.27	0.35	256
Aries	0.33	0.92	0.49	1061
Cancer	0.55	0.17	0.26	317
Capricorn	0.69	0.12	0.21	180
Gemini	0.83	0.07	0.13	142
Leo	0.46	0.24	0.31	338
Libra	0.53	0.12	0.20	209
Pisces	0.66	0.34	0.44	336
Sagittarius	0.59	0.39	0.47	425
Scorpio	0.57	0.12	0.19	318
Taurus	0.60	0.10	0.17	249
Virgo	0.87	0.08	0.14	169
accuracy			0.40	4000
macro avg	0.60	0.24	0.28	4000
weighted avg	0.53	0.40	0.34	4000

Classification Report:

	precision	recall	f1-score	support
Aquarius	0.54	0.54	0.54	256
Aries	0.57	0.80	0.67	1061
Cancer	0.51	0.42	0.46	317
Capricorn	0.66	0.52	0.58	180
Gemini	0.47	0.26	0.34	142
Leo	0.46	0.43	0.45	338
Libra	0.60	0.50	0.55	209
Pisces	0.63	0.62	0.62	336
Sagittarius	0.64	0.55	0.59	425
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Taurus	0.45	0.43	0.44	249
Virgo	0.61	0.38	0.47	169
accuracy			0.56	4000
macro avg	0.56	0.49	0.51	4000
weighted avg	0.56	0.56	0.55	4000

To measure performance of
each model we used

Confusion Matrix





Comparison and Insights

- Both Random Forest Classifier and LinearSVC models showed strong predictive abilities for Zodiac Sign Prediction.
- After hyperparameter tuning, LinearSVC achieved slightly higher accuracy, showcasing its effectiveness in this task.
- Algorithm choice depends on factors like complexity, interpretability, and efficiency.
- Random Forest excelled in capturing complex relationships, while LinearSVC leveraged text features.



Future Enhancements

- Fine-tuning hyperparameters remains an avenue for improved results."
- Feature engineering can enhance predictive power via techniques like embeddings.
- Model ensemble, combining Random Forest and LinearSVC, could boost accuracy.
- Incorporating external data sources could provide richer context for predictions.
- Advanced NLP techniques like sentiment analysis can further enrich text data.
- Online deployment, multilingual support, and ethical considerations are key for growth.



Thank You !