Resume classification and search

Automated Resume Parsing, Skill Extraction, and



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Introduction:

Purpose:

Simplify resume screening with automated tools.

Features:

Extract text from resumes (.docx, .pdf).
Identify technical skills and experience.
Classify resumes using machine learning.
Search and rank resumes based on criteria.

```
# Load the file
file_path = "Master Resume.docx"

[4]: # Load the .docx file
document = Document(file_path)

[5]: # Extract text from the .docx file
content = [paragraph.text.strip() for paragraph in document.paragraphs if paragraph.text.st

[6]: # Create a DataFrame
df = pd.DataFrame({'content': content, 'profile': ['Master Resume'] * len(content)})
df

[6]: content profile

O Chinna Subbarayudu M Master Resume

1 DOB: 06th March 1994 Master Resume
```

Key Components & Technical Stack:

Technical Stack:

Python libraries: Streamlit, scikit-learn, TfidfVectorizer, PyPDF2.

Predefined technical skills for matching.

Machine Learning:

Gradient Boosting Classifier for classification.

Data Processing:

TfidfVectorizer for text similarity.

Regex for experience extraction.

Features and Functionality:

Key Functionalities:

Resume Extraction: Supports .docx and .pdf formats.

Text Processing: Extracts skills and experience.

Search and Filter: Allows criteria-based filtering.

Model Training: Uses Gradient Boosting Classifier.

EDA and Visualizations:

Exploratory Data Analysis (EDA):

Analyzing the distribution of key features such as skills, experience years, and resume categories.

Visualizations:

Bar Charts: Display skill frequency in resumes.

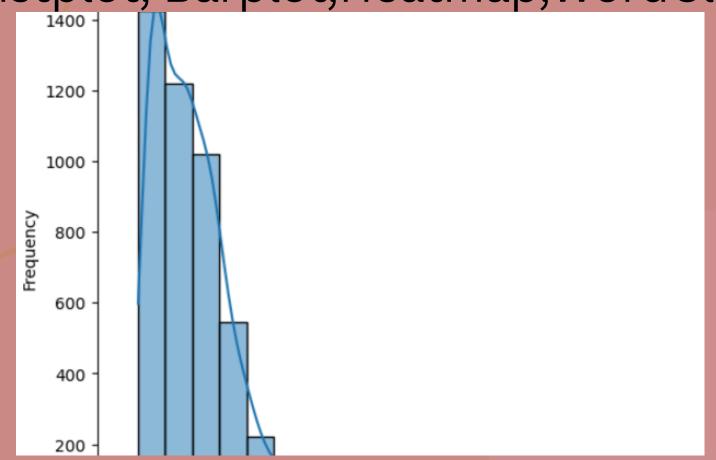
Word Clouds: Highlight most common skills across resumes.

Experience Distribution: Visualize the years of experience among

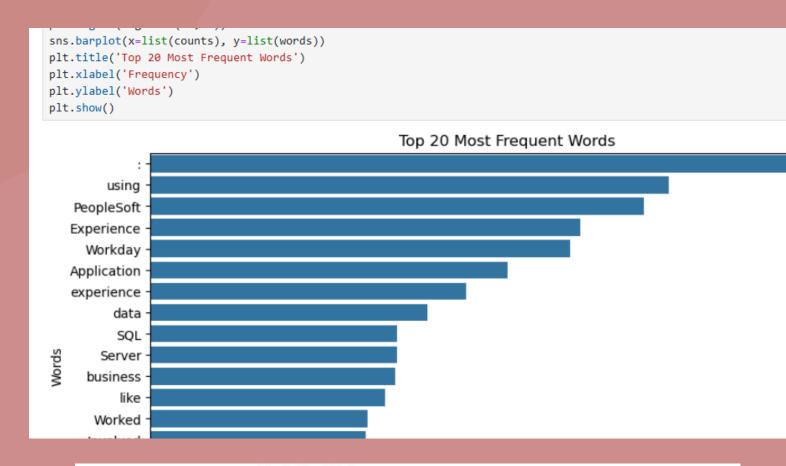
resumes.

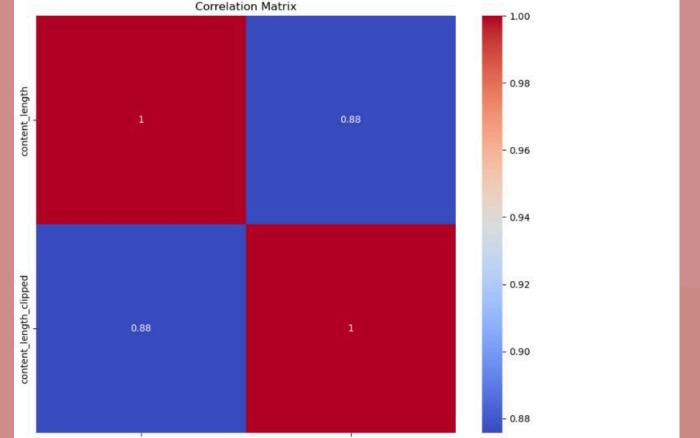
EDA & Visualizations part:

Plots:Histplot, Barplot, Heatmap, Word Cloud, etc.









Model Evaluation & Improvements:

Model Performance:

Accuracy, Precision, Recall, and F1-score.

Improvement Techniques:

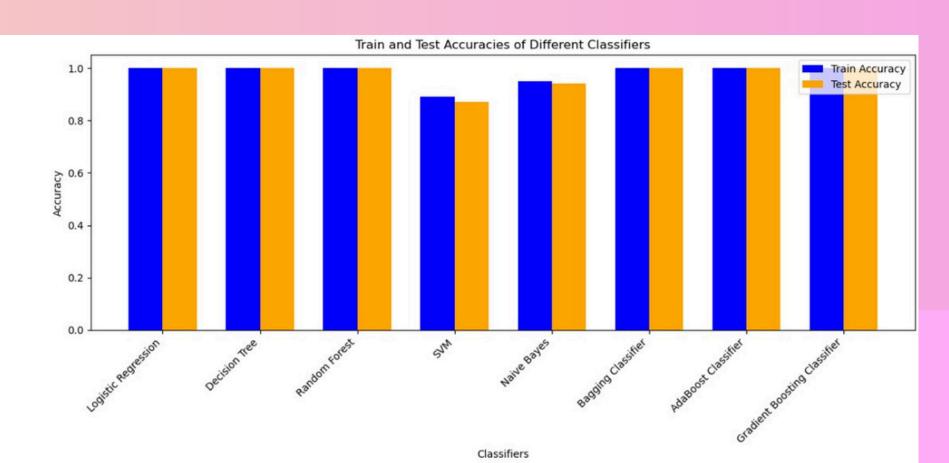
Hyperparameter Tuning.

Cross-validation to ensure model generalization.

```
train_accuracy = accuracy_score(y_train, y_train
test_accuracy = accuracy_score(y_test, y_test_pr

print(f"Training Accuracy: {train_accuracy:.2f}"
print(f"Testing Accuracy: {test_accuracy:.2f}")
```

Training Accuracy: 1.00 Testing Accuracy: 0.96



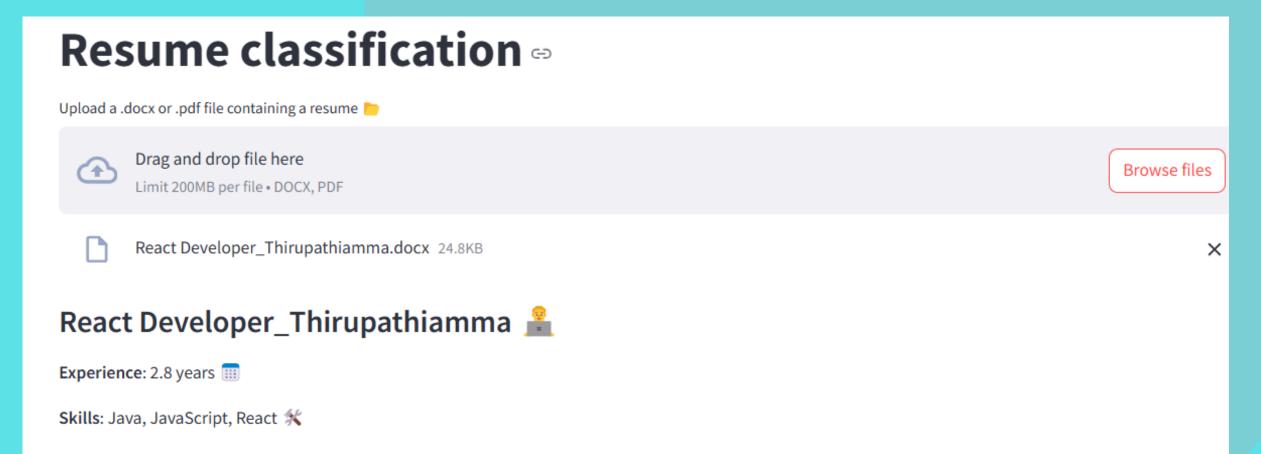
Deployment & Scalability:

Deployment Process:

Deploy the Streamlit app on cloud platforms (AWS, Heroku, Google Cloud).

Scalability:

Ensure scalability for handling large datasets of resumes. Use parallel processing to speed up processing time.



Challenges and Considerations:

Key Challenges:

Handling varied resume formats.

Extracting structured data from unstructured text.

Considerations:

Balancing search specificity and broad applicability.

Future Enhancements and Conclusion:

Future Enhancements:

Integration with ATS systems.

Multilingual resume support.

Advanced classification techniques.

Conclusion:

Streamlines recruitment processes, saves time, and improves productivity.

Efficient solution for resume parsing and classification using advanced ML techniques.

THANKYOU