Resume Analysis Tool

DATS 6312: Natural Language Processing

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Introduction

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Objective:

Elevate Recruitment efficiency through elaborate Resume Analysis

Key Features:

- Job Role Matching:
 - Multilabel Classification of Resumes to match relevant Job roles
- Smart Resume Analysis:
 - Resume Summarisation
 - Resume Skill Extraction
- Resume Job Description Similarity
 - Similarity Score for resume-job description
- Custom Criteria:
 - Tailor evaluations to match specific confidence thresholds.



Data Description

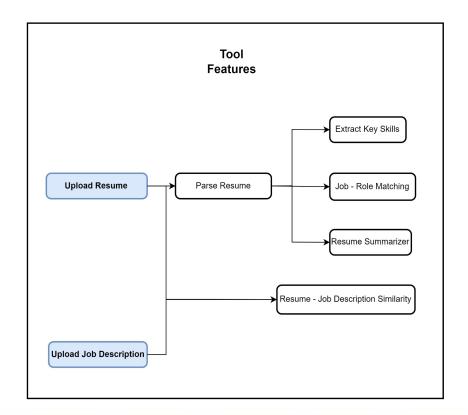
Multi-labeled dataset of resumes Labeled with occupations

- 1. Resumes_corpus.zip:
 - Contains individual resumes files in ".txt" format.
 - Corresponding labels are in a ".lab" file.
- 2. Resumes_sample.zip:
 - Aggregated dataset in a single text file.
 - Structure: Reference ID; Occupations; Text Resume.
 - Occupations separated by ";" in the second field.
- 3. Normalized_classes: Associations between original occupations and normalized forms.

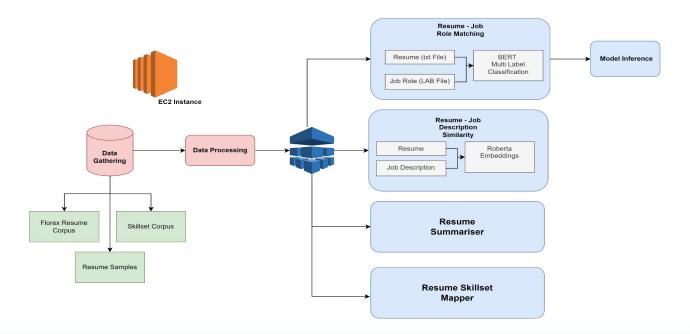


Features

- 1. Skillset Extraction
- 2. Job Role Matching
- 3. Resume Summariser
- 4. Resume Job Description Similarity



Experimental Setup



Resume - Job Role Matching

- Objective: The code aims to map resumes to job labels using a Machine Learning Model.
- Components: Utilizes the Traditional ML and Transformers library for BERT, PyTorch for model training.
- Data: Custom dataset loaded from a CSV file containing cleaned resumes and corresponding job labels.



Resume - Job Role Matching (continued)

Base Model: Multinomial NB

• Probabilistic algorithm suitable for text classification tasks.



Resume - Job Role Matching (continued)

Data Preprocessing

- Tokenization: Resumes are tokenized using the BERT tokenizer with a specified maximum length.
- Label Encoding: Job labels are processed using MultiLabelBinarizer, converting them into a binary matrix.
- Train-Validation Split: Data is split into training and validation sets (80% training, 20% validation).

Model Architecture

- BERT-based Model: Utilizes the pre-trained BERT model for sequence classification (BertForSequenceClassification).
- Labels: Number of output labels determined by the number of unique job classes in the dataset.
- Optimizer & Loss Function: Adam optimizer with BCEWithLogitsLoss as the loss function.



Results

Model	# Epochs	# Parameters	Precision	Recall	F1
BASE MODEL - MultinomialNB	-	-	54.70%	85.29%	66.65%
BERT Base - Linear Head	10	109.5M	85.15%	78.84%	81.87%
BERT Base - CNN Head	10	109.7M	88.29%	88.29%	85.18%
BERT Base - LSTM Head	10	111.6M	87.09%	82.80%	84.89%



Resume - Job Description Similarity

1. TF-IDF Cosine Similarity:

- Utilizes TF-IDF to represent document content.
- Measures cosine similarity to quantify the similarity between a resume and a job description.

2. BERT and RoBERTa Embeddings:

- Applies pre-trained NLP models (BERT and RoBERTa).
- Tokenizes and computes embeddings for both the resume and job description.
- Uses mean pooling for simplicity in obtaining document-level embeddings.



Resume - Job Description Similarity (continued)

- 3. Cosine Similarity Using BERT and RoBERTa:
 - Calculates cosine similarity between the embeddings of the resume and job description.
 - Provides similarity percentages for each model.

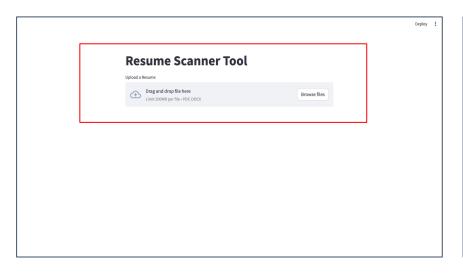


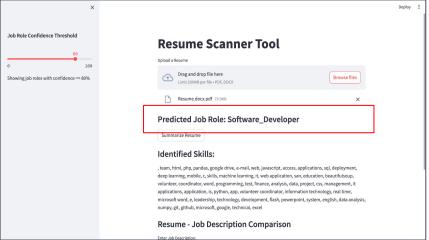
Skillset Mapping

- Step 1: Skills Extraction
 - Extracted skills from the resume_samples dataset.
- Step 2: Master List Creation
 - o Processed and compiled a master list of 4000 unique skills.
- Step 3: Skill Matching
 - Matched these skills with the input to determine if the resume contains relevant skills using regular expressions.

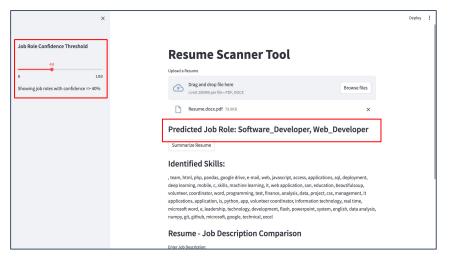


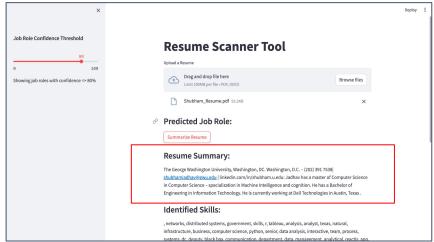
UI and Results





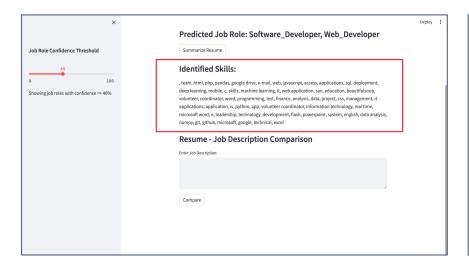
Resume Summarizer

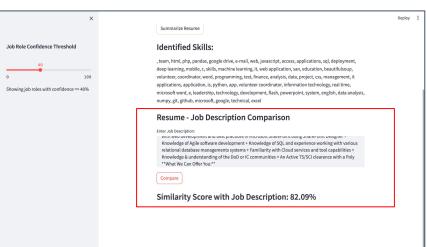




Method: Utilized transformer-based models for automated summarization of resume content.







Conclusion

- Job Role Matching: BERT Base with CNN classifier head has given the best F1 score of 85.18%.
- Smart Resume Analysis: A Skillset with 400 skills is curated and considered for Skill set Matching.
- Resume Job Description Similarity : Pretrained Roberta Tokenizer and Model is used to extract embeddings for Resume and Job Description followed by cosine similarity.

