

PDF Question Answering System Using NLP

A project on extracting answers from PDFs using fine-tuned natural language processing models to provide efficient and accurate solutions for users.

The background of the slide features a stylized, artistic illustration of a stack of books. The books have various colored spines, including shades of blue, green, yellow, and pink. A pink binder clip is attached to the top of the stack. The overall style is soft and painterly, with visible brushstrokes and a gentle, pastel color palette.

Introduction

With the vast amount of information stored in PDF documents, efficiently extracting specific answers to user queries is a challenging problem. This project aims to develop a system that can accurately retrieve relevant answers using advanced natural language processing techniques.

Approach

Data Collection

Use the SQuAD dataset for initial training to establish a solid foundation.

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Model Selection

Start with BERT and RoBERTa models, then fine-tune them to improve performance on the specific task.

Pre-processing

Extract text from PDFs, remove stop words, and perform tokenization to prepare the data for model training.



Failed Approaches

1

Initial Embedding Techniques

Word2Vec and GloVe embeddings lacked contextual understanding, leading to inaccurate results for complex queries.

2

Basic Text Extraction

Simple text extraction from PDFs without pre-processing resulted in poor quality text and inaccurate answers.

3

Custom Models without Fine-Tuning

Training models from scratch was not effective due to insufficient data and computational resources.

Results

Metrics

Achieved an accuracy of 85% on test queries, with consistent performance in precision and recall across different question types.

Visualizations

Graphs show the improvement in accuracy with fine-tuning, and a comparison of performance between different models.

Insights

Fine-tuning significantly boosts model performance, and advanced pre-processing techniques enhance the overall system accuracy.

Discussion

Significance of Results

The fine-tuning process and pre-processing steps like stop word removal and tokenization were crucial for improving the system's accuracy.

Insights Gained

Pre-trained models like BERT and RoBERTa are highly effective for question-answering tasks, and the use of contextual embeddings is essential for better understanding of queries.

Conclusion

1

Summary of Findings

Developed a functional PDF question-answering system with good accuracy, demonstrating the importance of fine-tuning and advanced pre-processing techniques.

2

Future Improvements

Incorporate more diverse datasets, enhance the user interface, and explore other NLP models like GPT-3 for potentially better results.



References

[1] Devlin, J., et al. (2018). BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding.

[2] Rajpurkar, P., et al. (2016). SQuAD: 100,000+ Questions for Machine Comprehension of Text.

[3] Pennington, J., et al. (2014). GloVe: Global Vectors for Word Representation.

[4] Mikolov, T., et al. (2013). Efficient Estimation of Word Representations in Vector Space.

Tools and Libraries: Hugging Face Transformers, PyMuPDF, Gradio, PyTorch