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OCR BASED IMAGE TEXT TO SPEECH CONVERSION

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ABSTRACT

This paper emphasizes the crucial role of customer retention in the contemporary banking sector, recognizing the substantial expenses associated with acquiring new customers. Images may contain important textual data that the user may need to edit or store digitally. Manual entry of data is time taking and may contain errors. There are millions of blind people in the world who are visually impaired. Disability to read has a large impact on the life of visually impaired people. The proposed system helps the visually impaired person to hear the text. The main idea of this project is optical character recognition which is used to convert text character into the audio signal. Tkinter for GUI, pytesseract for image processing and character recognition and pytsx3 for audio conversion are used.

Keywords:

OCR, Tesseract, video text, Image Capturing, Image To Text Converter, Python TSX3 Text To Speech Synthesizer, Translation and Speech Output.

INTRODUCTION

In the recent era, visual text in natural or man-made scenes might carry very important and useful information. Therefore, the scientists have started to digitize these images, extract and interpret the data by using specific techniques.

Optical character recognition is employed to recognize and extract the words and finally the extracted text is converted to appropriate speech using text-to-speech synthesizer. Text-to-speech (TTS) conversion is the process of converting written text

into spoken words using computer software. The resulting audio output can be customized by adjusting the speed, pitch, and other parameters to match the user's preferences.

LITERATURE SURVEY

Survey of major areas of project Deep Learning. Deep learning is a subset of machine learning that involves the use of artificial neural networks to learn from large volumes of data. This technology has been around for decades, but only in recent years has it seen significant growth and development. With the rise of big data and advancements in computing power, deep learning has become increasingly popular in solving complex problems across a wide range of industries. Deep learning is modeled after the structure and function of the human brain, using artificial neural networks to process and analyze data. These neural networks consist of layers of interconnected nodes, each of which performs a specific function. The first layer of nodes receives the input data, which is then processed through subsequent layers to produce a final output. The more layers a network has, the deeper it is, hence the name "deep learning." One of the main advantages of deep learning is its ability to automatically learn and improve from experience. This is accomplished through a process called backpropagation, which allows the network to adjust its parameters based on the error between its predictions and the actual outputs. By iteratively updating the weights and biases of the network, it can gradually improve its accuracy over time.

Deep learning has seen tremendous success in a wide range of applications, from image recognition to natural language processing and speech recognition. For example, image recognition models can be trained to accurately identify objects within photos,

allowing them to be used for tasks such as self-driving cars, security surveillance, and medical imaging. Natural language processing models can be trained to understand and interpret human language, enabling applications such as chatbots and voice assistants. One of the key advantages of deep learning is its ability to handle unstructured data, such as images, text, and audio. Traditional machine learning algorithms rely on structured data, which is typically represented in a tabular format. Deep learning, on the other hand, can learn directly from raw data, allowing it to extract patterns and insights that may not be immediately apparent in structured data. 4 Despite these challenges, deep learning has seen significant growth and development in recent years. Advances in hardware, such as the development of specialized deep learning chips and cloud-based computing resources, have made it more accessible to a wider range of users. Additionally, the development of new algorithms and techniques, such as transfer learning and reinforcement learning, have opened up new possibilities for deep learning in a wide range of applications. In conclusion, deep learning is a powerful subset of machine learning that has the ability to learn and improve from experience. Its ability to handle unstructured data has made it particularly valuable in fields such as image recognition and natural language processing. While there are challenges associated with deep learning, such as the need for large volumes of data and computing resources, advances in technology and algorithms have made it more accessible and applicable to a wider range of users and applications. PyTesseract PyTesseract uses a combination of several algorithms to perform optical character recognition (OCR) on images. The main algorithm used by PyTesseract is the Tesseract OCR engine, which was originally developed at Hewlett-Packard Laboratories in the 1980s and has since been developed and maintained by Google. The Tesseract OCR engine uses a combination of image processing techniques, statistical models, and machine learning algorithms to recognize characters in an image. The engine works by first preprocessing the image to enhance its contrast and remove noise. It then segments the image into individual characters or words and applies a set of rules to recognize each character or word. To improve the accuracy of OCR, PyTesseract also uses several other algorithms and techniques. For example, PyTesseract can perform adaptive thresholding to improve the contrast of the

image and reduce noise. It can also apply language-specific dictionaries and models to improve the recognition of words and phrases. In addition to these algorithms, PyTesseract also provides several options and parameters that can be used to fine-tune the OCR process. For example, you can specify the language of the text to be recognized, adjust the contrast and brightness of the image, and control the amount of noise reduction applied to the image. Overall, PyTesseract uses a combination of algorithms and techniques to perform OCR on images with high accuracy and reliability. 5

The Tesseract OCR engine is the backbone of the PyTesseract library, but other algorithms and techniques are used to enhance the OCR process and improve the accuracy of the results. OCR stands for Optical Character Recognition, which is the technology used to extract text from scanned documents, images, or other digital files. It involves analyzing the text characters in an image and converting them into machine-readable text that can be edited, searched, and processed by computers. OCR can handle various types of fonts, including standard fonts like Times New Roman and Arial, as well as less common fonts like handwriting or cursive. OCR can be used with various file formats, such as PDFs, JPEGs, TIFFs, and PNGs. OCR technology can be trained to recognize specific languages or scripts, such as Latin, Chinese, or Arabic. OCR accuracy can vary depending on various factors, such as image quality, font type, and character complexity. However, modern OCR systems can achieve high levels of accuracy, ranging from 95% to 99%. OCR technology can be enhanced with additional features, such as natural language processing, document classification, and data extraction. These features can help improve the accuracy and efficiency of OCR applications. OCR technology has advanced significantly in recent years, with the development of deep learning algorithms and neural networks. These technologies can improve the accuracy and speed of OCR applications, especially for complex documents or handwriting. OCR is not foolproof and can still make errors, especially with poorly scanned or low-quality images. Therefore, it is always a good idea to review and verify the OCR output to ensure its accuracy. Overall, OCR is a versatile and powerful technology that can have significant benefits in various industries, including document management, data entry, and accessibility. With continued advancement in OCR technology, we

can expect to see even more innovative applications and use cases in the future. OCR technologies.

Simultaneously, industries are increasingly adopting data-driven decision-making, driven by the growth of big data and analytics capabilities. In the airline industry, big data analytics holds promise for optimizing aviation operations, enhancing customer service, and revenue management. Airfare prediction, a complex task, has advanced with machine learning algorithms such as Linear Regression, Support Vector Machines, and Random Forests. Recent studies explore more sophisticated models like Deep Regressor Stacking to improve airfare price prediction accuracy, showcasing the continuous evolution of predictive modeling in the airline industry.

However, challenges persist in data availability and organization, particularly concerning airline ticket data. While datasets like T100 and DB1A/1B exist, the limited association between prices and specific flight information poses obstacles to comprehensive analysis. Researchers often rely on web crawling or private data sources, making replication and comparison challenging.

1. PROBLEM DEFINITION

Implement the solution using appropriate programming languages, libraries, and frameworks for image processing, machine learning, and speech synthesis. Consider deploying the system on platforms suitable for the intended use case, such as desktop applications, mobile devices, or cloud services.

By addressing the outlined components, challenges, and solution approach, you can develop an effective OCR-based image text-to-speech conversion system.

Limitations of existing system

Accuracy: Measure the accuracy of text extraction and speech synthesis against ground truth data. **Efficiency:** Evaluate the system's speed and resource usage, especially for real-time applications. **User Experience:** Gather feedback on the naturalness and understandability of the synthesized speech. **Robustness:** Test the system's

performance under various conditions, including different types of images and text.

Proposed system

The proposed system uses pytesseract for Optical Character Recognition which internally uses Long Short-Term Memory which is a part of Recurrent Neural Network which is more suitable for handwritten texts and to recognize large portions of textual data. Not only from the image but it also extracts the text from the videos by splitting the video into frames using OpenCV. Further, the extracted text from images and videos is converted into speech by using pyttsx3.

FIGURES

Figure 1: SYSTEM ARCHITECTURE

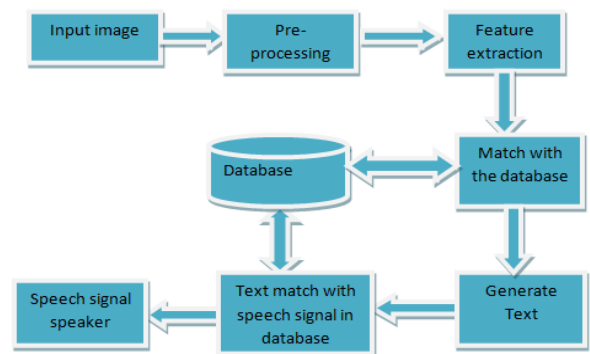


Figure 2

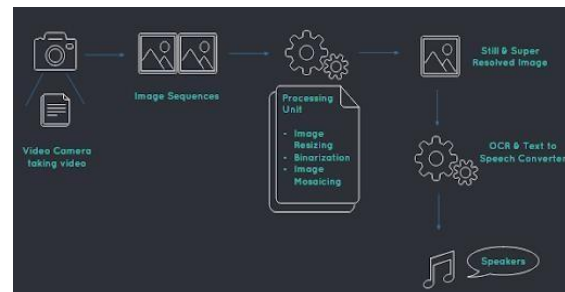
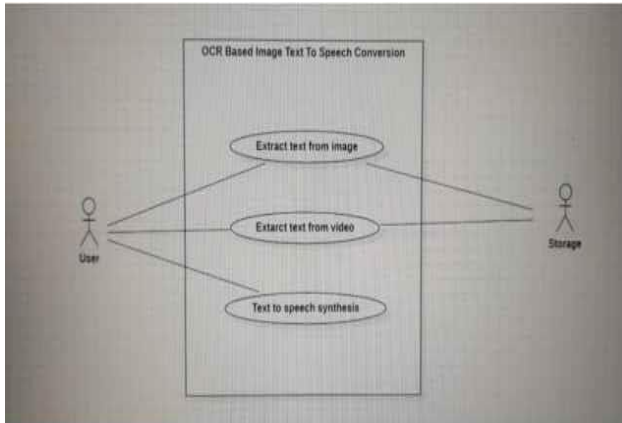


Image working model of the images and videos

Use Case Diagram



MODULES

1. Image Input Module:

Responsible for accepting input images containing text. Handles image loading, whether from files, cameras, or other sources.

2. Image Preprocessing Module:

Performs preprocessing techniques to enhance OCR accuracy. Includes operations like resizing, noise reduction, contrast enhancement, and binarization.

3. Optical Character Recognition (OCR) Module:

Text Detection: Identifies regions of text within the input image. Text

Recognition: Recognizes and extracts text from the detected regions.

Post-processing: Refines the OCR output to improve accuracy and formatting.

4. Text Processing Module:

Text Cleaning: Removes noise, artifacts, and

unwanted characters from the extracted text.

Language Identification: Determines the language of the text for accurate pronunciation during speech synthesis.

Text Normalization: Standardizes the format and structure of the text for better TTS performance.

5. Text To Speech Conversion:

Text Analysis: Analyzes the processed text for prosody, intonation, and emphasis.

Voice Selection: Optionally allows users to choose different voices or accents for speech synthesis.

Speech Synthesis: Converts the processed text into audible speech using TTS engines or models.

6. Output Module:

Handles the output of synthesized speech. Plays the synthesized speech through speakers or headphones. Provides a graphical user interface (GUI) for user interaction and feedback.

7. Integration and Control Module:

Orchestrates the flow of data and control between different modules. Manages error handling, logging, and feedback mechanisms for system monitoring and improvement. Provides APIs or interfaces for integration with other applications or services.

8. Performance Optimization Module:

Optimizes the system for efficiency and speed. Utilizes techniques like parallel processing, model optimization, and hardware acceleration for improved performance.

Monitors and optimizes resource usage to ensure scalability and reliability.

9. Deployment Module:

Handles deployment aspects such as platform compatibility, scalability, and security. Supports deployment on various platforms, including desktop, mobile, and cloud environments. Implements security measures to protect user data and ensure privacy during image processing and speech synthesis.

DEPLOYMENT

Choose Deployment Platform:

Determine the platform(s) on which the system will be deployed, such as desktop, mobile, or cloud. Consider factors like target audience, scalability requirements, and resource constraints.

Software Packaging:

Package the system components into deployable software artifacts suitable for the chosen platform(s). For desktop deployment, create executable files or installer packages. For mobile deployment, package the system as a mobile application for iOS or Android. For cloud deployment, containerize the system using Docker or deploy it as serverless functions.

Platform-specific Considerations:

Desktop: Ensure compatibility with various operating systems (Windows, macOS, Linux) and provide installation instructions for end-users.

Mobile: Publish the application on app stores (e.g., Apple App Store, Google Play Store) and comply with platform-specific guidelines and requirements.

Cloud: Deploy the system on cloud platforms like AWS, Azure, or Google Cloud Platform, considering factors like scalability, availability, and cost. Scalability and

Performance:

Configure the deployment environment to handle varying workloads and scale resources dynamically as needed. Optimize system performance for responsiveness and reliability, especially during peak usage periods.

Testing and Quality Assurance:

Conduct thorough testing of the deployed system to ensure functionality, performance, and user experience meet

expectations. Perform integration testing, compatibility testing, and user acceptance testing across different deployment platforms.

Documentation and Support:

Provide comprehensive documentation, including installation instructions, user guides, and troubleshooting tips. Offer support channels (e.g., email, forums, helpdesk) for users to seek assistance and report issues.

Monitoring and Maintenance:

Implement monitoring tools and processes to track system performance, usage metrics, and user feedback. Regularly update the system with bug fixes, enhancements, and security patches to maintain reliability and security.

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