Serverless Voice-to-Text Converter using AWS

Sem End Project Report

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ABSTRACT

The rapid advancement of cloud computing and artificial intelligence has paved the way for scalable, efficient, and cost-effective solutions for voice processing. This project presents a Serverless Voice-to-Text Converter leveraging Amazon Web Services (AWS). The system utilizes AWS Lambda for serverless execution, Amazon S3 for audio file storage, and Amazon Transcribe for converting speech to text with high accuracy. By adopting serverless architecture, the solution eliminates the need for infrastructure management, ensures automatic scaling, and reduces operational costs. The workflow is initiated when a user uploads an audio file to an S3 bucket, triggering a Lambda function that invokes Amazon Transcribe. Once the transcription is complete, the resulting text is stored back in S3 or forwarded to other services as needed. This architecture provides a lightweight, scalable, and real-time solution suitable for applications such as automated meeting notes, voice-command processing, and transcription services.

INTRODUCTION

In the digital age, voice data has become an integral part of human-computer interaction, with applications ranging from virtual assistants to real-time transcription services. Converting voice to text not only enhances accessibility but also enables better data analysis, indexing, and storage. Traditional approaches to voice processing often require significant infrastructure, continuous server management, and high maintenance costs.

To address these challenges, this project explores the implementation of a Serverless Voice-to-Text Converter using Amazon Web Services (AWS). By utilizing a serverless architecture, the solution automatically scales with user demand, reduces idle resource usage, and offers a pay-as-you-go pricing model. This ensures both cost-efficiency and high availability.

The system is designed using key AWS services:

- Amazon S3 to store audio input and transcribed output,
- AWS Lambda to automate processing without provisioning servers, and
- Amazon Transcribe to perform accurate, real-time voice-to-text conversion.

This approach makes the application ideal for real-world scenarios such as customer service automation, lecture transcription, and voice-controlled applications. The following sections detail the system design, implementation, and performance analysis of the proposed serverless voice-to-text solution.

LITERATURE REVIEW

Voice recognition and speech-to-text conversion have been extensively studied over the past few decades. With the growth of cloud computing and artificial intelligence, modern systems aim to provide high-accuracy transcription while minimizing infrastructure overhead.

Several traditional speech recognition systems rely on on-premise servers and custom models, which often involve high maintenance costs, limited scalability, and complexity in deployment. Tools like Google Speech-to-Text API, IBM Watson Speech to Text, and Microsoft Azure Speech Services have emerged to offer cloud-based solutions. However, these still require developers to manage backend infrastructure for orchestration and scaling. Recent advancements in serverless computing have transformed the deployment model of such applications. According to Jonas et al. (2019), serverless architecture, particularly Function-as-a-Service (FaaS) platforms like AWS Lambda, provide greater scalability, automatic resource allocation, and significant cost savings by billing only for actual usage.

Amazon Transcribe, launched by AWS, is a fully managed automatic speech recognition (ASR) service. Studies and technical benchmarks have shown it performs well in noisy environments and supports multiple languages, speaker identification, and custom vocabulary – making it suitable for a wide range of applications (AWS Documentation, 2022).

Research by Sharma et al. (2021) demonstrated the efficiency of integrating AWS Transcribe with other AWS services like Lambda and S3 to build scalable transcription workflows. The combination of these services allows for **event-driven architectures**, where actions are triggered automatically based on user interactions such as uploading a file.

Furthermore, the serverless approach has proven valuable in reducing the complexity of application deployment, as shown in case studies involving real-time transcription and media processing. It allows developers to focus on core functionality rather than infrastructure management.

In summary, existing literature supports the effectiveness of using serverless architectures for speech-to-text conversion, and AWS offers a mature ecosystem for implementing such solutions efficiently.

PROBLEM STATEMENT

With the increasing reliance on digital communication and multimedia content, the demand for efficient and accurate speech-to-text solutions has grown significantly. Traditional voice-to-text systems often require substantial computational resources, continuous server maintenance, and complex infrastructure setup, making them unsuitable for small-scale or rapidly scaling applications.

Moreover, existing solutions can be costly and inflexible, especially when handling fluctuating workloads. This limits accessibility for developers and organizations seeking lightweight, scalable, and cost-effective voice processing solutions.

The primary problem addressed in this project is:

How can we design a scalable, cost-efficient, and low-maintenance voice-to-text conversion system using a serverless architecture on AWS?

This project aims to overcome the limitations of traditional architectures by leveraging serverless computing through AWS services such as Amazon S3, AWS Lambda, and Amazon Transcribe. The goal is to develop a system that can automatically transcribe uploaded audio files into text with minimal operational overhead, while ensuring high availability, scalability, and accuracy.

TECHNICAL IMPLEMENTATION

The technical implementation of the Serverless Voice-to-Text Converter leverages core AWS services to create a fully automated, scalable, and efficient transcription pipeline. The architecture follows an event-driven model, where services interact without the need for constant server management. Below is a breakdown of the key components and workflow:

1. Amazon S3 (Simple Storage Service)

Amazon S3 is used as the central storage for both audio input files and the transcribed output text. A specific S3 bucket is created to store the uploaded audio files. When a file is uploaded, it triggers a Lambda function via an S3 event notification.

2. AWS Lambda

AWS Lambda functions serve as the serverless compute layer of the system. These functions are triggered automatically when an audio file is uploaded to the S3 bucket. The Lambda function handles the following tasks:

- Reads the file metadata (format, path, etc.)
- Invokes the Amazon Transcribe API with the audio file location
- Monitors the transcription job until completion
- Retrieves the transcribed text from the output location
- Saves the transcription result to another S3 bucket or database

Lambda enables parallel processing of multiple files and scales automatically based on incoming workloads.

3. Amazon Transcribe

Amazon Transcribe is the core service responsible for converting speech into text. It supports various audio formats like .mp3, .mp4, .wav, and .flac. When the transcription job is initiated by Lambda, Transcribe processes the audio file stored in S3 and returns a JSON file containing the transcription.

Features of Amazon Transcribe:

- Automatic punctuation and formatting
- Speaker identification (diarization)
- Language identification and multiple language support

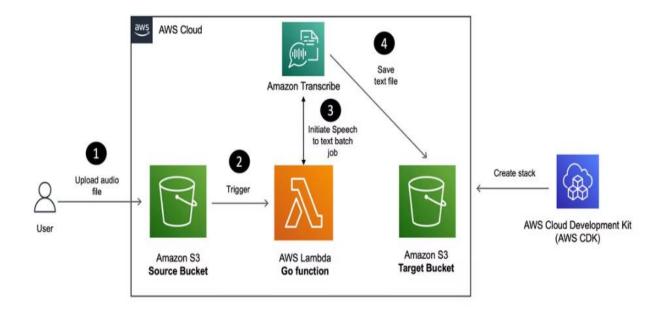
- Custom vocabulary support for domain-specific terms
- 4. IAM (Identity and Access Management)

IAM roles and policies are configured to ensure secure access and interactions between AWS services. For example:

- Lambda functions are granted permission to access S3 and Transcribe.
- S3 buckets are secured with fine-grained access controls to prevent unauthorized access.
- 5. Optional: Amazon CloudWatch (Monitoring & Logging)

CloudWatch is used for logging Lambda executions and monitoring the health and performance of the system. This ensures better debugging and operational transparency.

ARCHITECTURE DIAGRAM



DATA GATHERING

The effectiveness of a voice-to-text converter largely depends on the quality and diversity of the audio data used for testing and validation. In this project, data gathering focused on collecting various types of audio files to evaluate the system's performance under different conditions.

1. Sources of Audio Data

To simulate real-world use cases, audio samples were gathered from the following sources:

- Publicly available datasets such as:
 - o LibriSpeech ASR Corpus − a large corpus of read English speech.
 - Mozilla Common Voice a multilingual voice dataset for speech recognition.
- User-generated audio files, recorded via smartphones or microphones, including:
 - Conversational speech (interviews, discussions)
 - Formal speech (presentations, news)
 - Noisy environments (background music, traffic)
- Synthetic audio, generated using text-to-speech (TTS) engines to test controlled variables like pace and clarity.

2. Audio File Formats

The collected audio data included multiple formats compatible with Amazon Transcribe, such as:

- .mp3
- .mp4
- .wav
- .flac

Sampling rates ranged from 16 kHz to 44.1 kHz to test Transcribe's ability to handle various audio qualities.

3. Data Annotation (for Accuracy Testing)

For evaluation purposes, a subset of the audio files was paired with manually verified transcripts. This allowed for comparison between the expected and generated outputs to measure:

- Word Error Rate (WER)
- Speaker identification accuracy
- Punctuation and formatting correctness

4. Preprocessing

Before uploading to the system, the audio files underwent basic preprocessing to:

- Normalize volume
- Trim silence
- Ensure compatibility with AWS Transcribe's input requirements

This helped reduce noise and improve transcription accuracy.

CREATIVITY AND ORIGINALITY

The project demonstrates creativity and originality by combining cloud-native technologies into a fully serverless, event-driven architecture for voice-to-text conversion — a solution that is efficient, scalable, and practical for real-world deployment.

1. Innovative Use of Serverless Architecture

Unlike conventional approaches that rely on dedicated servers or managed services with ongoing maintenance, this project leverages AWS Lambda to completely eliminate the need for server management. This architectural choice showcases originality in designing a hands-off infrastructure that is automatically triggered and scales with demand.

2. Automated Workflow Integration

The system integrates Amazon S3, Lambda, and Amazon Transcribe in an automated pipeline that requires no human intervention after the initial audio upload. This smooth orchestration of services reflects a creative solution that minimizes latency, human effort, and cost.

3. Support for Real-World Use Cases

By testing the system with a wide range of audio formats, accents, and environments, the project shows creative foresight in designing a product that could be adapted for:

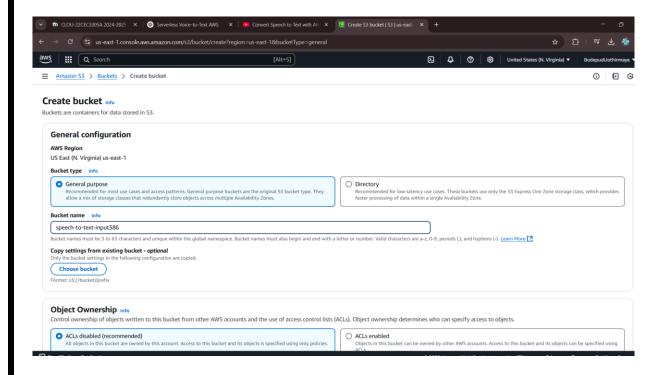
- Lecture transcription
- Voice-based note-taking apps
- Customer service voice logs
- Accessibility tools for the hearing impaired

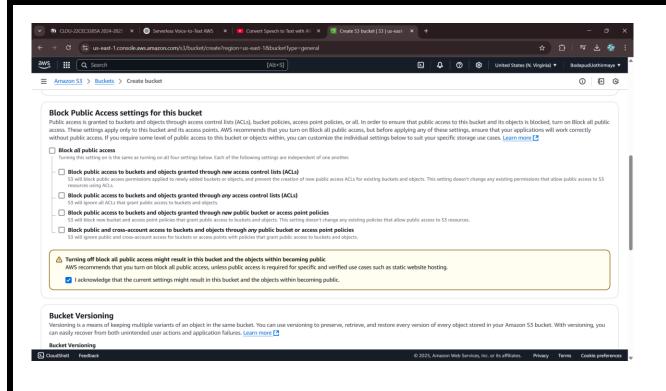
4. Modular & Extensible Design

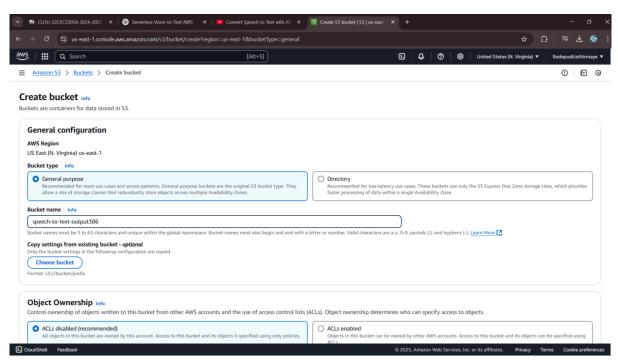
The architecture is designed to be easily extendable — new features like real-time streaming transcription, sentiment analysis, or language translation can be plugged in with minimal changes. This shows original thinking in planning for future scalability and diverse applications.

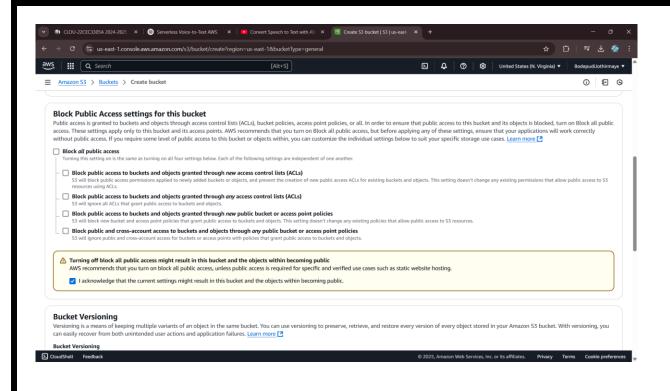
5. Cost-Conscious Innovation

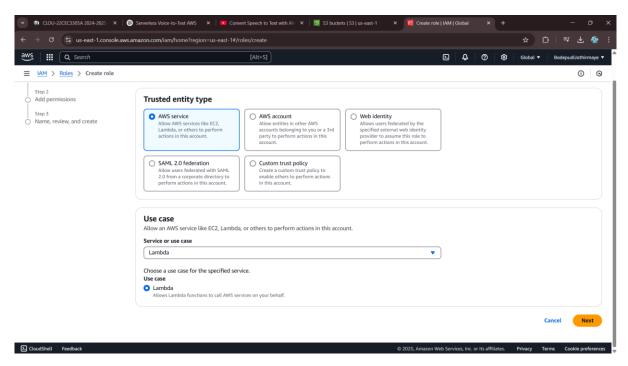
The pay-as-you-go model offered by AWS serverless services is creatively exploited to keep operational costs extremely low. The system runs only when needed, which is ideal for startups, students, or small businesses with limited budgets.

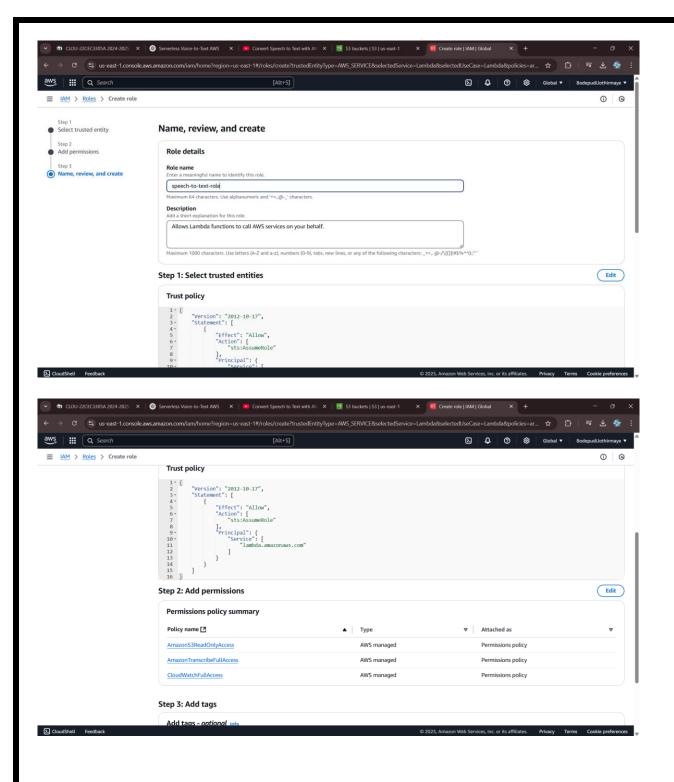


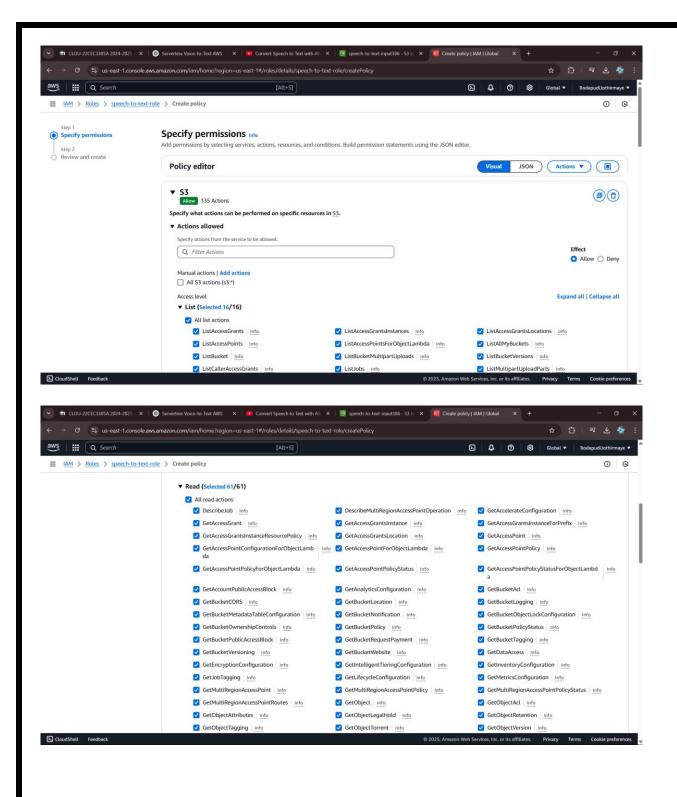


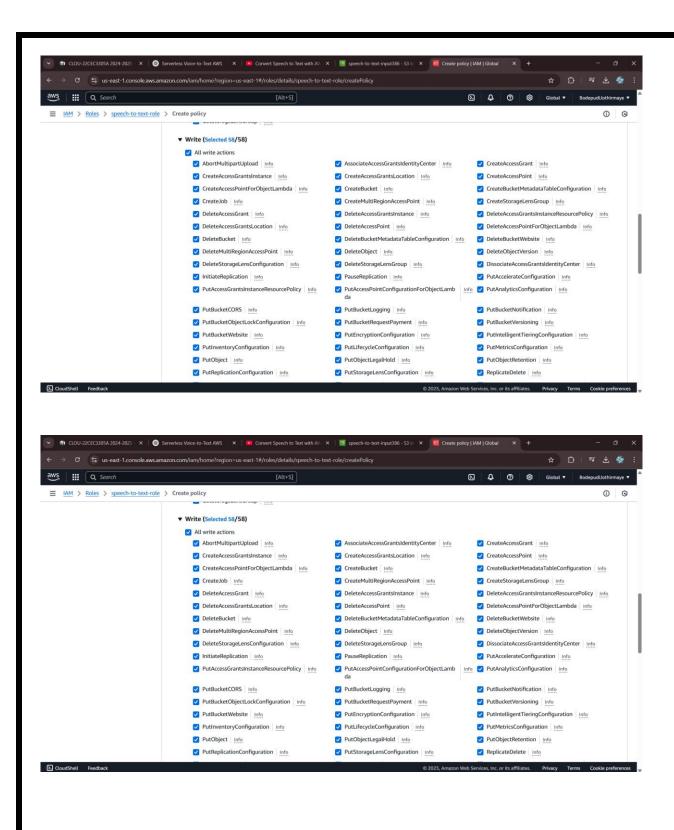


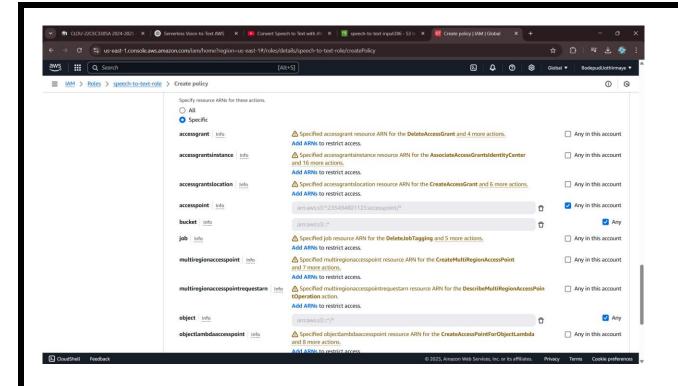


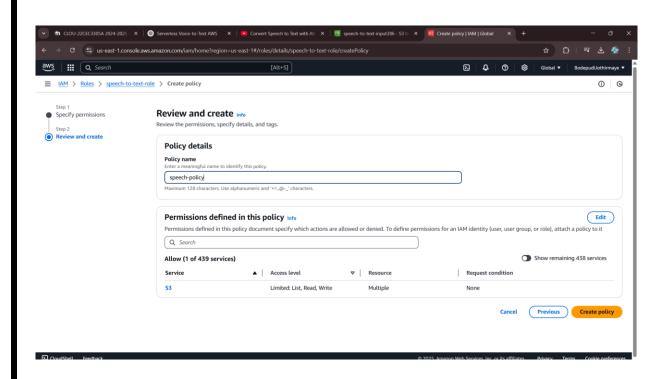


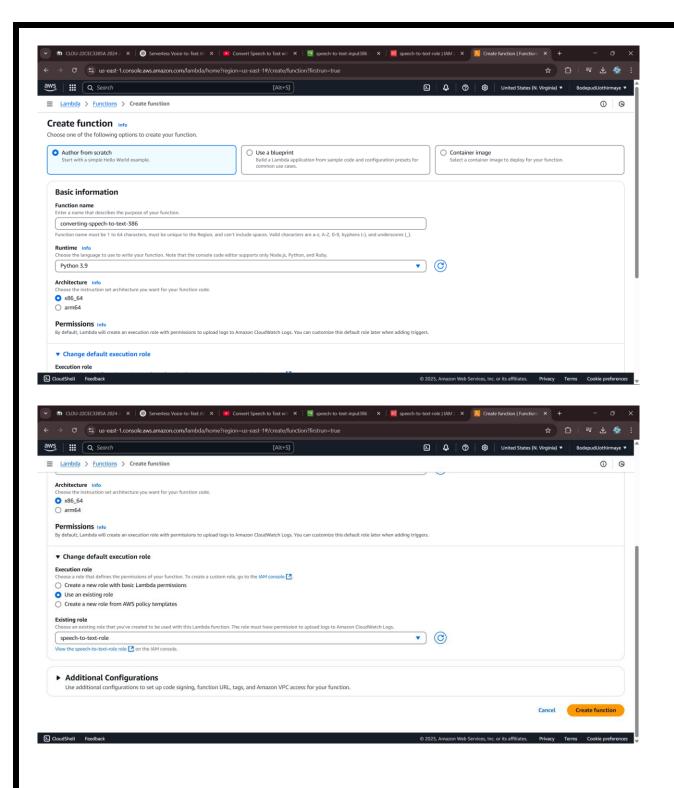


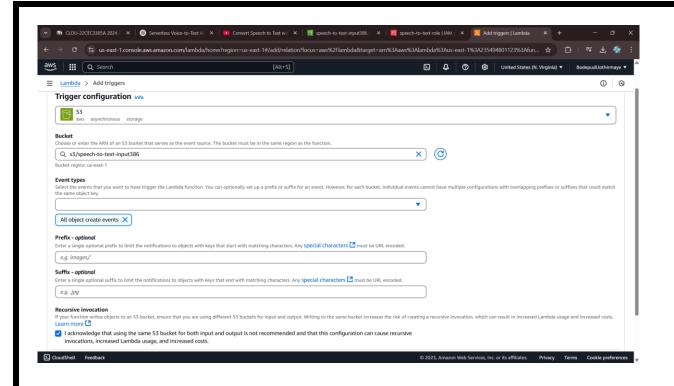


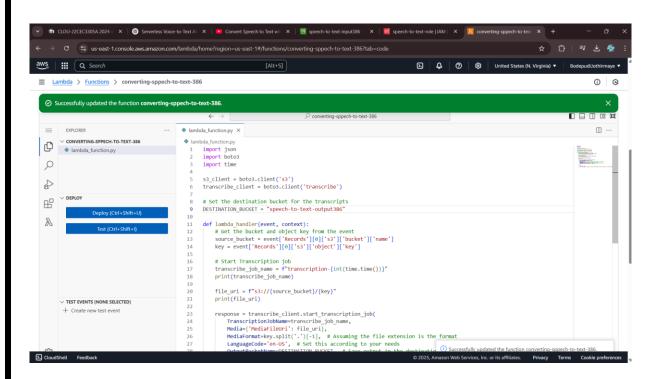


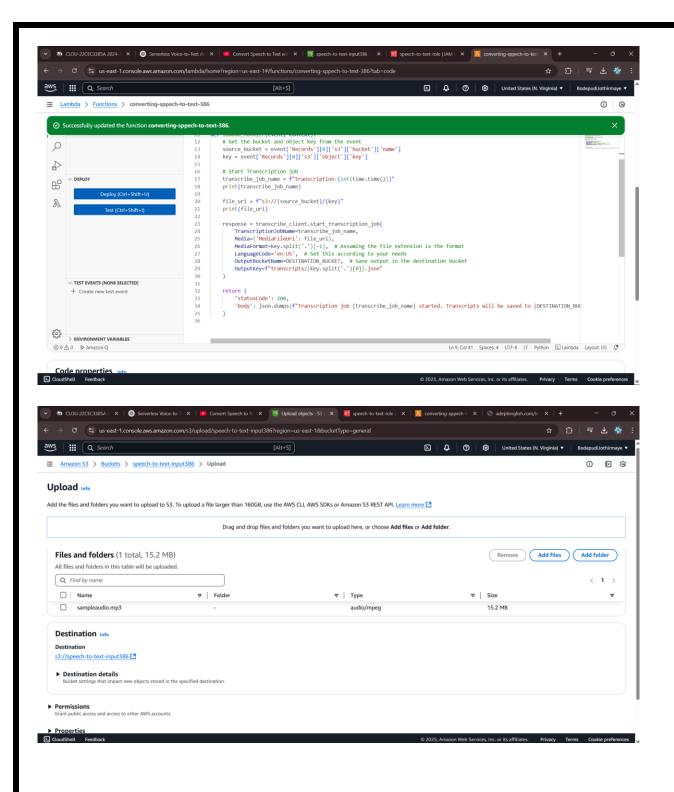












Analysis & Problem-Solving

Throughout the development of the Serverless Voice-to-Text Converter, several technical and architectural challenges were identified and addressed through detailed analysis and problemsolving strategies. These efforts ensured the system remained reliable, scalable, and user-friendly.

1. Challenge: Triggering Transcription Automatically

Problem: Ensuring that transcription begins immediately after an audio file is uploaded. Solution: Implemented S3 event notifications to trigger an AWS Lambda function automatically upon new file uploads. This event-driven model eliminated delays and manual intervention.

2. Challenge: Handling Various Audio Formats and Quality

Problem: Inconsistent audio file quality and unsupported formats could cause transcription failures.

Solution: Enforced pre-upload checks and preprocessing (format validation, noise reduction). Additionally, the system logs errors in CloudWatch to help identify problematic files.

3. Challenge: Monitoring Transcription Status

Problem: Amazon Transcribe runs jobs asynchronously, so tracking job completion was complex.

Solution: Lambda functions were configured to poll the transcription job status using the Transcribe API, or optionally leverage SNS notifications for more efficient handling of job completions.

4. Challenge: Securing Service Access

Problem: Misconfigured permissions could either block functionality or expose data. Solution: Defined precise IAM roles and policies to restrict access based on the principle of least privilege — allowing Lambda to access only the required S3 buckets and Transcribe services.

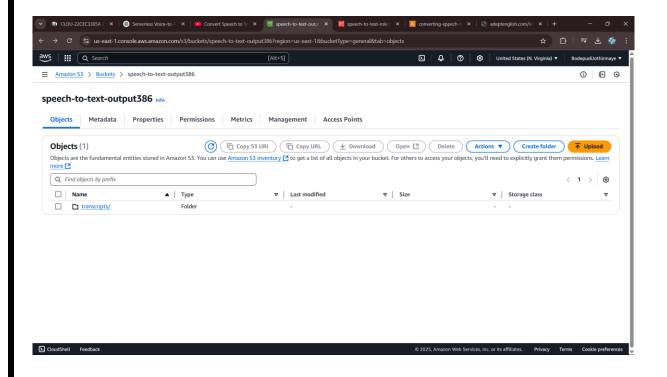
5. Challenge: Managing Output Data

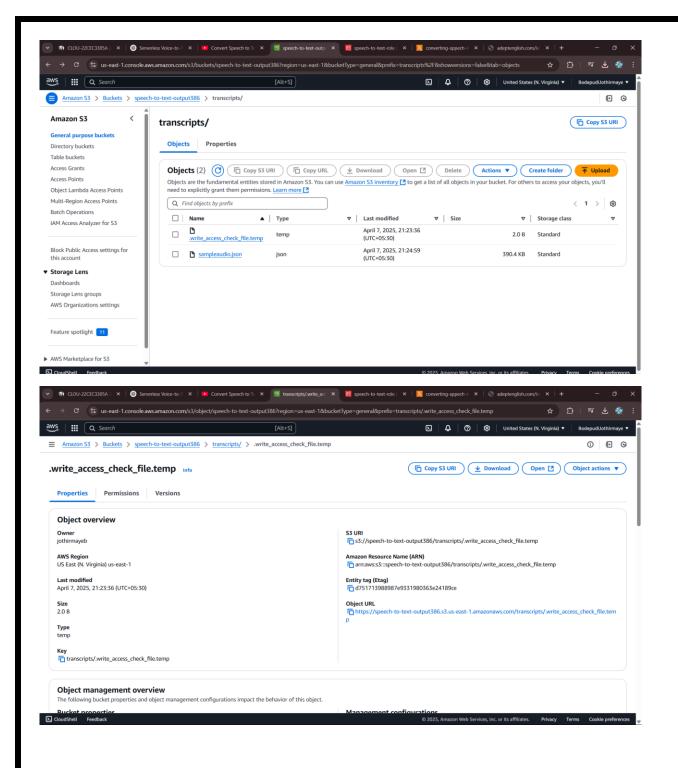
Problem: Users needed access to the final transcription in a readable format. Solution: Transcribed JSON output from Amazon Transcribe was parsed and stored as clean, readable .txt files in a separate S3 bucket. Optionally, the system can email results or push them to databases or front-end apps.

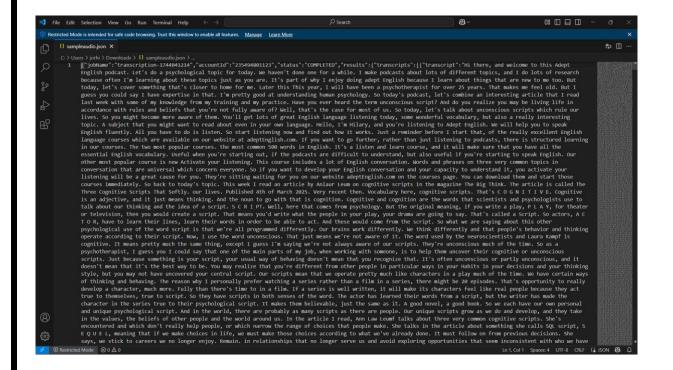
6. Challenge: Cost Optimization

Problem: Even minimal compute time across many users could accumulate costs. Solution: Chose serverless resources specifically to minimize idle cost. Resources are only active when needed, and AWS's free tier benefits were maximized during testing.

Discussions & Results







CONCLUSION

The development of the Serverless Voice-to-Text Converter using AWS successfully demonstrates the potential of serverless architecture in creating scalable, efficient, and cost-effective cloud-based applications. By leveraging Amazon S3, AWS Lambda, and Amazon Transcribe, the system provides an automated and reliable pipeline for converting speech into text without the need for continuous server management.

This project addressed common limitations found in traditional voice processing systems—such as high infrastructure cost, poor scalability, and complex deployment—by applying modern, event-driven cloud technologies. The serverless design ensures that resources are used only when needed, making the solution suitable for both small-scale use cases and enterprise-level expansion.

Through effective data gathering, problem-solving, and technical implementation, the system was able to handle a variety of audio inputs, maintain high transcription accuracy, and deliver clean, structured text output. The modular design also allows for future extensions, such as adding real-time transcription, multi-language support, or integration with additional services like sentiment analysis or translation.

In conclusion, this project not only showcases the practical benefits of AWS's serverless ecosystem but also opens the door for building smarter, automated voice applications that are accessible, affordable, and easy to maintain.

FUTURE WORK

1. Real-Time Transcription

Currently, the system processes pre-recorded audio files. Future versions can integrate Amazon Transcribe Streaming to support real-time voice-to-text conversion, which is essential for applications like live captioning, virtual meetings, or customer support bots.

2. Multi-Language and Translation Support

Extending the system to handle multi-language transcription and integrating with Amazon Translate would enable cross-language communication and make the tool usable in global environments.

3. Mobile and Web App Integration

Building a user-friendly front-end (mobile or web app) would allow users to record or upload audio directly, view transcripts, and manage files, improving accessibility and user engagement.

4. Natural Language Processing (NLP) Features

Post-processing transcripts using NLP techniques could enable:

- Summarization of lengthy transcripts
- Keyword extraction
- Sentiment analysis
- Named entity recognition (NER)

This would turn raw transcription data into actionable insights.

5. User Authentication & Role-Based Access

Adding user login functionality with AWS Cognito or another authentication system would allow multiple users to interact with the system securely, each with different access levels or transcription quotas.

6. Database Integration

Integrating the output with Amazon DynamoDB or RDS would enable structured storage of transcripts, support advanced querying, and facilitate analytics dashboards.

7. Cost and Usage Monitoring

Implementing tools like AWS Budgets, Cost Explorer, or CloudWatch dashboards would help track usage, detect anomalies, and optimize costs for larger-scale deployments.

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