### A system to generate speech to text in real-time

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Abstract— A real-time speech to text conversion system converts the spoken words into text. Speech-to-Text technology enables us to convert audio to text by applying powerful neural network models. It has a number of applications for users with and without disabilities. Speech-to-text has been used for voice search, help writers boost their productivity, and to provide alternate access to a computer for individuals with physical impairments. Other applications include speech recognition for foreign language learning, voice-activated products for the blind and many familiar mainstream technologies. It is a driving force behind the success of new age voice-controlled speakers like Amazon Echo and Google Home.

While cloud solutions like Google Cloud Speech-to-Text, Amazon Transcribe, IBM watson Speech to Text etc effectively convert audio and voice into written text, but there are some scalability challenges in effectively using them. In this project, we tried to address some of those challenges and also tested what are the configurations and parameters yield best results for such a system.

Keywords — Speech-to-Text, voice search, Amazon Echo, Google Home

#### 1. Introduction.

In this article, we are proposing a scalable system architecture to generate text from speech input in real time. We also are publishing our experimental results and performance metrics for different technologies used as the building blocks of this architecture.

# 2. Related Work, Challenges And How We Addressed the Challenges.

A lot of work is already been done in this field. Cloud services became fairly good at converting audio and voice into written text using Machine Learning Algorithm like Neural Network, but still, such a system has scalability challenges in effectively using them. Following are few challenges that drives most of the architecture decisions while designing such a system -

### A. Selection of scalable system design patterns.

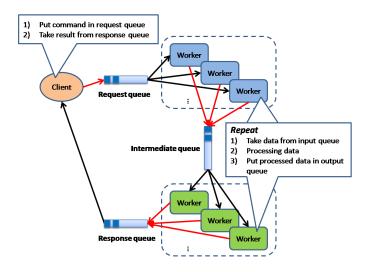
If we push streaming audio directly to web server, then such a system is bound to choke under heavy load, as CPU will be busy processing previous inputs and new inputs will be waiting to be consumed and eventually system will start throwing 5XX errors . So we need to provisionally store the audio file , before an available process can pick it up and start processing .

We researched several Scalable System Design Patterns<sup>[2]</sup> and found out " event-based architecture" model is most suitable to address this problem .Event-based architecture supports several communication styles:

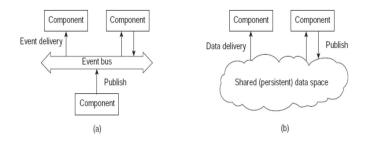
- Publish-subscribe
- Broadcast
- Point-to-point

Publish-subscribe communication style decouples sender & receiver and facilitates asynchronous communication. Event-driven architecture (EDA) promotes the production, detection, consumption of, and reaction to events[3].

The main advantage of this architecture is that they are loosely coupled.



So for this application, we are pushing the audio files to a distributed queue (Kafka ) and a multithreaded Spring Boot Application consumes those Audio files from the queue.



B. Selection of storage platform for provisional storage of audio file.

We need to store the audio file in a DataBase / Storage system before a consumer picks that up and starts processing . Audio files need to be stored as BLOB data . We looked at several DataBases and found NoSql database like MongoDB performs better at storing BLOB data compared to traditional RDBMS systems<sup>[4]</sup>.

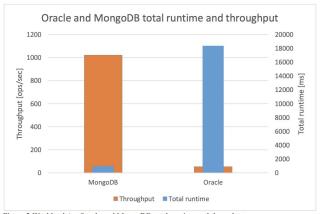


Figure 2 Workload A - Oracle and MongoDB total runtime and throughput

However, we found that storing the Audio files in MongoDB will not be a good choice, because -

- a. Maximum document size supported is 16MB and large sound files can easily exceed that range .
- b. If we store the audio in MongoDB, then consumer application need to constantly poll the DB to check if there is a new file. If we increase the polling frequency then this new event detection is instantaneous, but it increases the system workload. So there is a tradeoff between workload and response time.

So , we **decided to use Kafka as provisional storage platform** for Audio files. Kafka performs drastically faster than other point-to-point messaging platform like Activemq or Rabbitmq<sup>[5]</sup>.

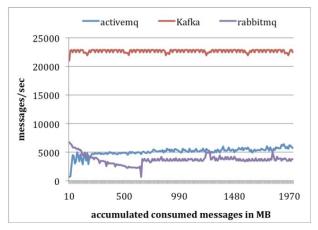


Figure 5. Consumer Performance

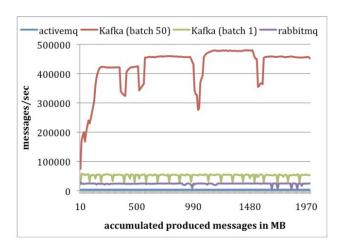


Figure 4. Producer Performance

C. Storing Large files in Kafka

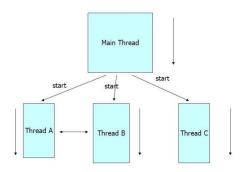
Kafka was initially designed as Distributed Messaging System for Log Processing. So default max message size  $\sim$  1MB . A > 10 Sec 1 channel .wav file easily exceeds 1MB in size. So we need to override default configs in order to support large messages. Following properties had to be changed to support this. These values are in bytes , hence out system could easily handle  $\sim\!\!40\text{MB}$  message.

message.max.bytes=41943040 | max.request.size=41943040 | replica.fetch.max.bytes=41943040 | fetch.message.max.bytes=41943040

D. Multithreaded kafka consumer

Multiple producers push to kafka in a multi-partition topic, it's obvious that we also need a multi-threaded consumer to keep the whole thing robust and balanced . Previous work has been done in this topic<sup>[6]</sup> to implement a multithreaded Kafka consumer using Kafka-Python .But kafka-python is not thread safe and does not support multi-threading.

To overcome this challenge we decided to use **Spring-kafka** along without multithreaded Spring Boot application, which is thread safe and graciously supports multithreading.



# 3. TECHNOLOGIES USED IN SYSTEM IMPLEMENTATION

A. Apache Kafka - Apache Kafka® is a distributed streaming platform. A streaming platform has three key capabilities: 1) Publish and subscribe to streams of records, similar to a message queue or enterprise messaging system. 2) Store streams of records in a fault-tolerant durable way. 3) Process streams of records as they occur.

We are using kafka as our transient storage of Audio files. Related work<sup>[6]</sup> in this area shows with default message size configuration overrides, how it can effectively be used as a storage of BLOB files. And how it's faster compared to other messaging platforms[5].

 B. MongoDB - MongoDB is an open-source, document database designed for ease of development and scaling.

We are using MongoDB to store the transcripts for future reference. Related work in this area affirms MongoDB performs better when compared to RDBMS for unstructured / semistructured data[4]. Ease of use was one of key reason of our choice as well.

C. Google cloud Speech to Text - Powerful speech recognition cloud service that enables developers to convert audio to text by applying powerful neural network models in an easy-to-use API.

For this experiment we tried other services like Amazon Transcribe, IBM watson Speech to Text and found Google Speech to Text gives more accurate results with low median latency compared to other two platforms.

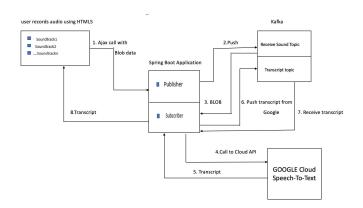
However, Amazon Transcribe supports .mp3 file format and it allows the user to customize vocabulary which is not supported in Google Speech-to-text service.

D. **Spring Boot** - Spring Boot is a Spring framework module which provides RAD (Rapid Application Development) feature to the Spring framework.

We used Spring along with Tomcat as out application server to process client requests. We are also using Spring-Kafka to that Spring components can work as kafka Producer/Consumer.

- E. Sockets a JavaScript library for real time web applications. It enables real time, bi-directional communication between web clients and servers.
- F. HTML5 Web Audio API The Web Audio API provides a powerful and versatile system for controlling audio on the Web, allowing developers to choose audio sources, add effects to audio, create audio visualizations, apply spatial effects (such as panning) and much more.

#### 4. PROPOSED SYSTEM ARCHITECTURE AND DESIGN



A. We are using HTML5 MediaStream Recording API to record the audio inside a browser environment

```
navigator.mediaDevices.getUserMedia(constraints).then(function(stream) {
    /* assign to gumStream for later use */
    gumStream = stream;

    /* use the stream */
    input = audioContext.createMediaStreamSource(stream);

    /*
    Create the Recorder object and configure to record mono sound (1 channel)
    Recording 2 channels will double the file size
    */
    rec = new Recorder(input,{numChannels:1})

    //start the recording process
    rec.record()

}).catch(function(err) {
    //enable the record button if getUserMedia() fails
    recordButton.disabled = false;
    stopButton.disabled = true;
});
```

B. Posting the Audio as BLOB data using Ajax to a Spring Boot application running on Tomcat Server

C. A Spring controller receives the Audio data and pushes it to Kafka topic received sound

```
@RequestMapping("/api/savesound")
public void sendMessage(@RequestParam MultipartFile audio_data) throws IOException{
  byte[] bytes = audio_data.getBytes();
  kafkaBlobTemplate.send("recieved_sound", bytes);
}
```

D. A kafka Consumer receives an event that new audio is available and it consumes the audio and stream that audio to Google Speech-To-Text api. And when the transcript is available pushes it to topic 'transcriptToClient'.

```
@KafkaListener(topics = "recieved_sound")
public void processSound(byte[] sounddata) {
    String transcript = cloudService.transcript(sounddata);
    logger.info(transcript);
    kafkaTemplate.send("transcriptToClient", transcript);
}
```

E. A service method calls to google Speech-To-Text api to get the transcript from google.

F. When the transcript is available in 'transcriptToClient' topic a consumer picks it up and stores it in MongoDB for future reference. it also pushes the transcript to a open socket connection initiated by the client.

```
@KafkaListener(topics = "transcriptToClient", groupId = "group-id"
public void listen(String message) throws Exception{
    logger.info("Received Messasge: " + message);
    //Saves transcript in mongo
    repository.save(new Transcript(""+message.hashCode(), message);
    webSocket.convertAndSend("/topic/backToClient", message);
}
```

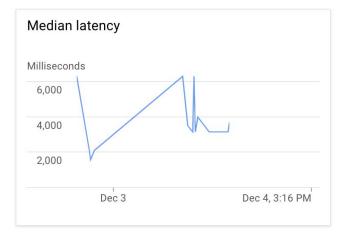
G. Client who is listening on an open socket receives the transcript and displays it in the browser

```
function connect() {
   var socket = new SockJS('/gs-guide-websocket');
   stompClient = Stomp.over(socket);
   stompClient.connect({}, function (frame) {
        setConnected(true);
        console.log('Connected: ' + frame);
        stompClient.subscribe('/topic/backToClient', function (transcript) {
            showTranscript(transcript.body);
        });
    });
}
```

#### 5. RESULTS AND THEIR ANALYSIS.

A. Comparison of cloud services to generate text from speech

We tried out Google Cloud Speech-to-Text, Amazon Transcribe, IBM watson Speech to Text for our experiment . We found Google Cloud Speech-to-Text performs the best compared to other too . Error rate and latency time is both significantly lower for Google Speech-to-Text. Avg. Latency time is 6sec for < 1MB files.



#### B. System load testing

We tested our system with multiple parallel client requests and a dual partition kafka topic with 2 brokers . And received following metrics in a macOs Mojave (MacBook Pro 2017, 2.3 GHz Intel Core i5, 8 GB 2133 MHz LPDDR3)

PRODUCER

```
start.time
                          : 2018-12-03 21:38:28:094
end.time
                          : 2018-02-03 21:38:28:449
compression
                          : 0
message.size
                          : 100
batch.size
                          : 200
total.data.sent.in.MB
                          : 0.01
MB.sec
                          : 0.0269
total.data.sent.in.nMsg
                          : 100
                          : 281.6901
nMsg.sec
```

 start.time
 : 2018-12-04 11:29:41:806

 end.time
 : 2018-12-04 11:29:46:854

 fetch.size
 : 1048576

 data.consumed.in.MB
 : 0.0954

 MB.sec
 : 1.9869

 data.consumed.in.nMs
 : 1001

 nMsg.sec
 : 20854.1667

Consumer

## C. Effects of Audio Encodings Formats on Generated Transcripts.

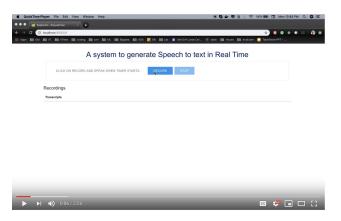
We recorded mono sound (1 channel ) and noticed recording 2 channels double the file size. We used FLAC encoding , LINEAR16 and MULAW encoding separately with .wav file format .

**FLAC and LINEAR16 performs better** with almost error free output when used for Text generation .

FLAC and LINEAR16 are lossless encoding compared to MULAW . MULAW is an 8 bit PCM encoding, where the sample's amplitude is modulated logarithmically rather than linearly. As a result, MULAW reduces the effective dynamic range of the audio thus compressed. So a higher error rate was expected for MULAW.

### 6. Demo Link.

HTTPS://www.youtube.com/watch?v=4fz2eoAz8P8&t=6s



#### 7. Source Code -

HTTPS://GITHUB.COM/SAP9433/SPEECHTOTEXT

#### REFERENCES

- [1] Yadav, Devender. "Save Large Files in MongoDB Using Kundera DZone Database." Dzone.com. July 24, 2016. Accessed December 01,
- 2018.https://dzone.com/articles/save-large-files-in-mongo db-using-kundera.
- [2] Ho, Ricky. "Scalable System Design Patterns DZone Database." Dzone.com. October 18, 2010. Accessed December 01, 2018.

https://dzone.com/articles/scalable-system-design

- [3] Distributed Systems Architecture . Accessed November 14, 2018.
- $http://cse.csusb.edu/tongyu/courses/cs660/notes/distarch.p\\ hp$
- [4] Kamil Kolonko, "Performance comparison of the most popular relational & non-relational database management systems" IEEE

http://www.diva-portal.org/smash/get/diva2:1199667/FU LLTEXT02

[5] Jay Kreps, Neha Narkhede, Jun Rao, "Kafka: a Distributed Messaging System for Log Processing". Accessed November 14, 2018. ACM

978-1-4503-0652-2/11/06

http://notes.stephenholiday.com/Kafka.pdf

[6] Guy Shilo, "Moving binary data with Kafka – a more realistic scenario". Accessed November 14, 2018. http://www.idata.co.il/2017

/12/moving-binary-data-with-kafka-a-more-realistic-scena rio/

[7] Nodar MOMTSELIDZE, "Apache Kafka - Real-time Data Processing". Accessed November 14, 2018. https://jtst.ibsu.edu.ge/jms/index.php/jtst/article/view/80 [8] Misal, Disha. "Google Speech Vs Amazon Transcribe: The War Of Speech Technology." Analytics India Magazine. October 22, 2018. Accessed December 05, 2018.

 $https://www.analytics in diamag.com/google-speech-vs-am\ azon-transcribe-the-war-of-speech-technology/.$