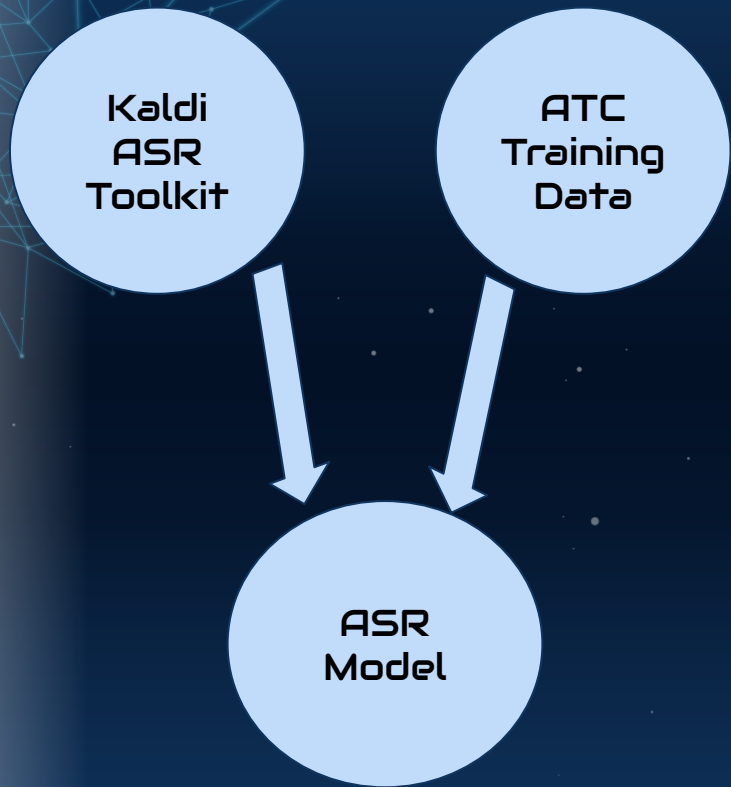


# Kaldi ASR Research Team

Team: Tabitha O'Malley, Tamina Tisha, David Serfaty, Milan Haruyama, Adam Gallub  
Customer: Dr. Jianhua Liu

# What is the Kaldi ASR Toolkit?



- **Kaldi ASR Toolkit**
  - Builds automatic speech recognition models
- **Automatic Speech Recognition**
  - Conversion of speech to text
- **Training Data**
  - 30-hour ATC dataset
- **ASR Model**
  - Speech recognition for ATC applications

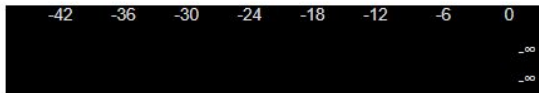
# Background

- Complexities in ATC communication requires extensive training
  - Instructor time
  - Lack of existing resources for students
- RTube Web Application
  - Transcribe live ATC transmissions in real time

You are listening to:

**KDAB App/Dep - Daytona Beach, Florida, United States**

KDAB 281353Z 32007KT 10SM BKN120 BKN200 10/03 A3022 RMK A02 SLP232 T01000033



# Customer Needs

## In Progress

1. Understanding current sample ASR model provided by Kaldi ASR Toolkit

## Future Iterations

1. Adjust the sample ASR model using the 30-hour ATC dataset
2. Compare the performance of the trained models
3. Experiment with other models
4. Apply the best trained model
5. Experiment with callsign and frequency identification

# Major Requirements

## 1. Computer

- a. Windows Subsystem for Linux (WSL)
- b. Kaldi ASR Toolkit
- c. Sufficiently powerful GPU
- d. Sufficient amount of memory

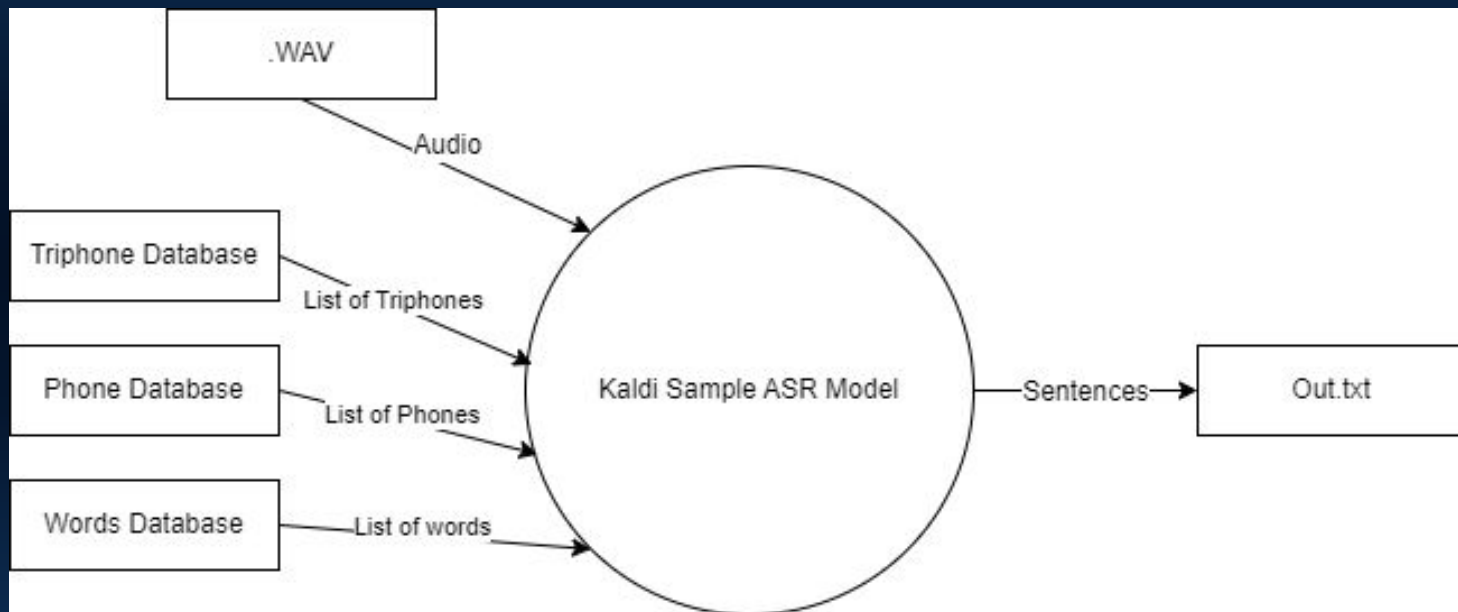
## 2. WAV file as input

- a. Contains spoken speech
- b. Convert non-WAV files using FFMPEG utility

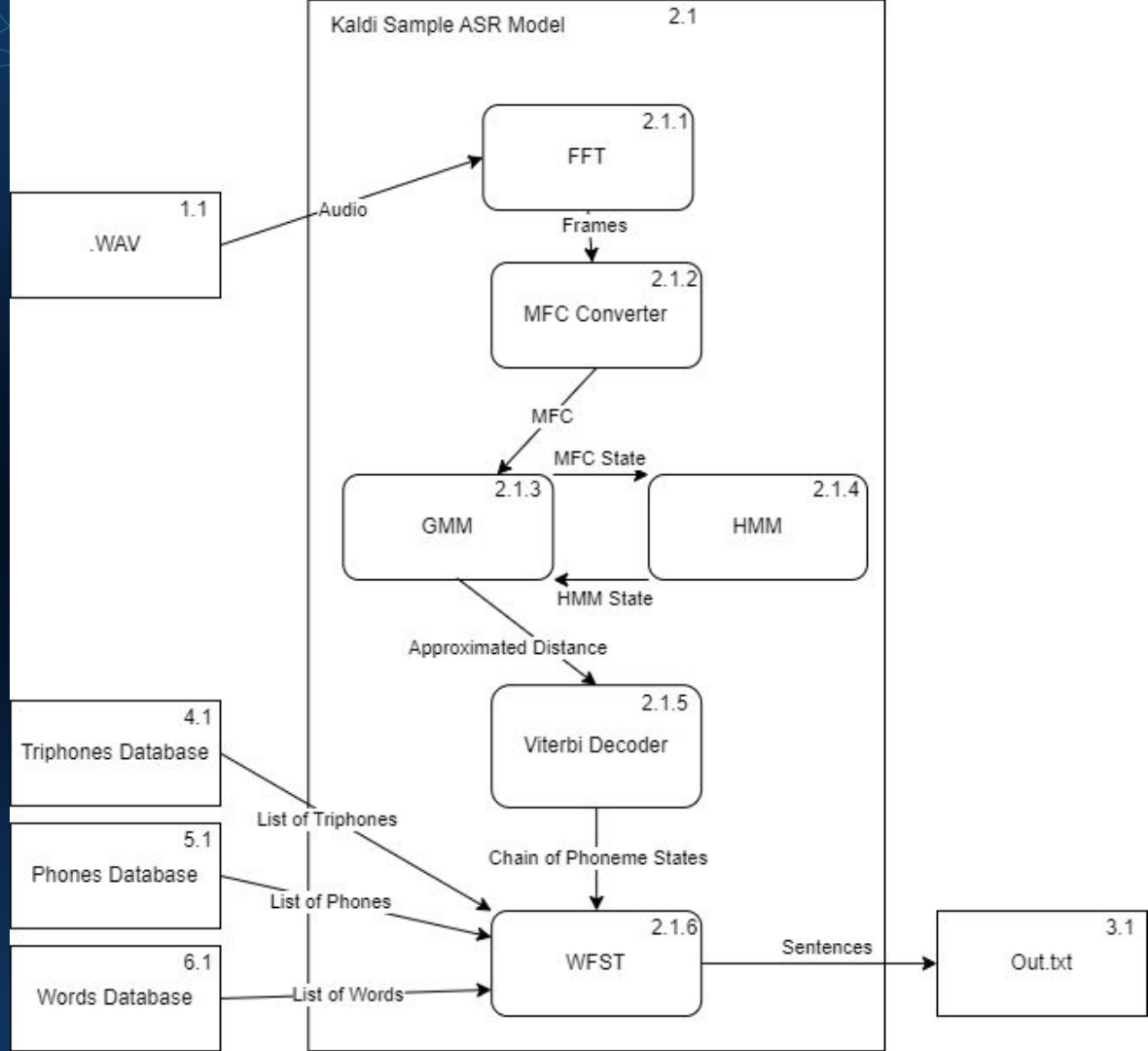
## 3. TXT file as output

- a. Contains words transcribed from the WAV file

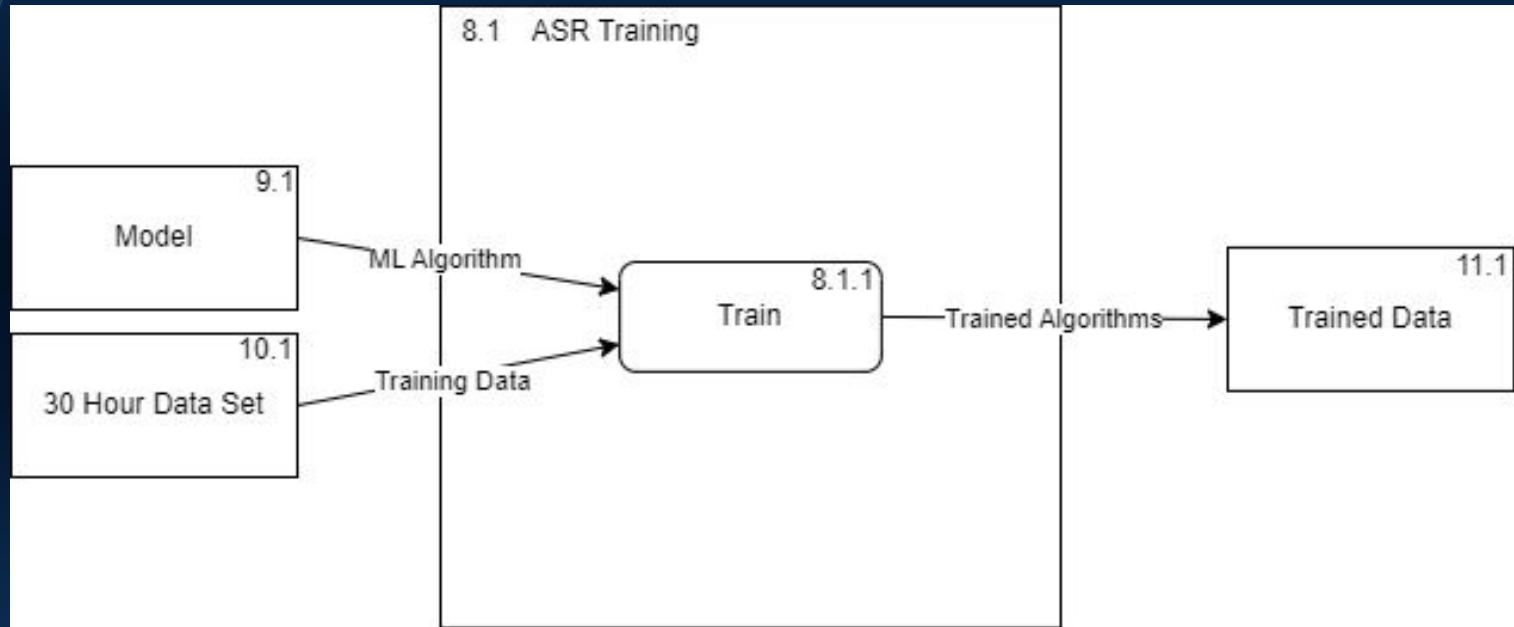
# DFD Level 0: Context Diagram



# DFD Level 1 Sample ASR Model

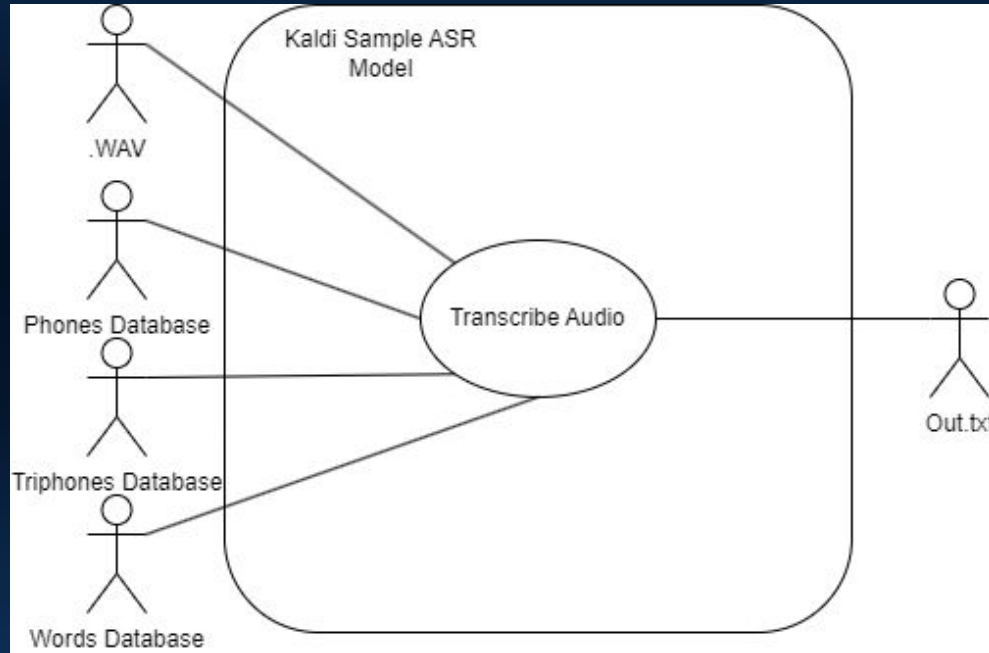


# DFD Level 1 ASR Training

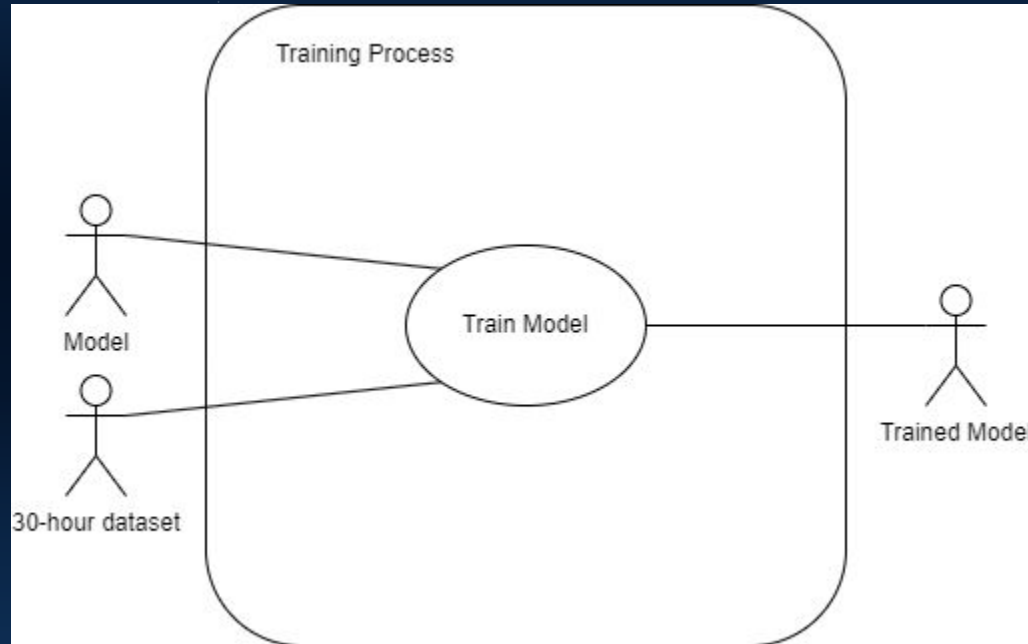


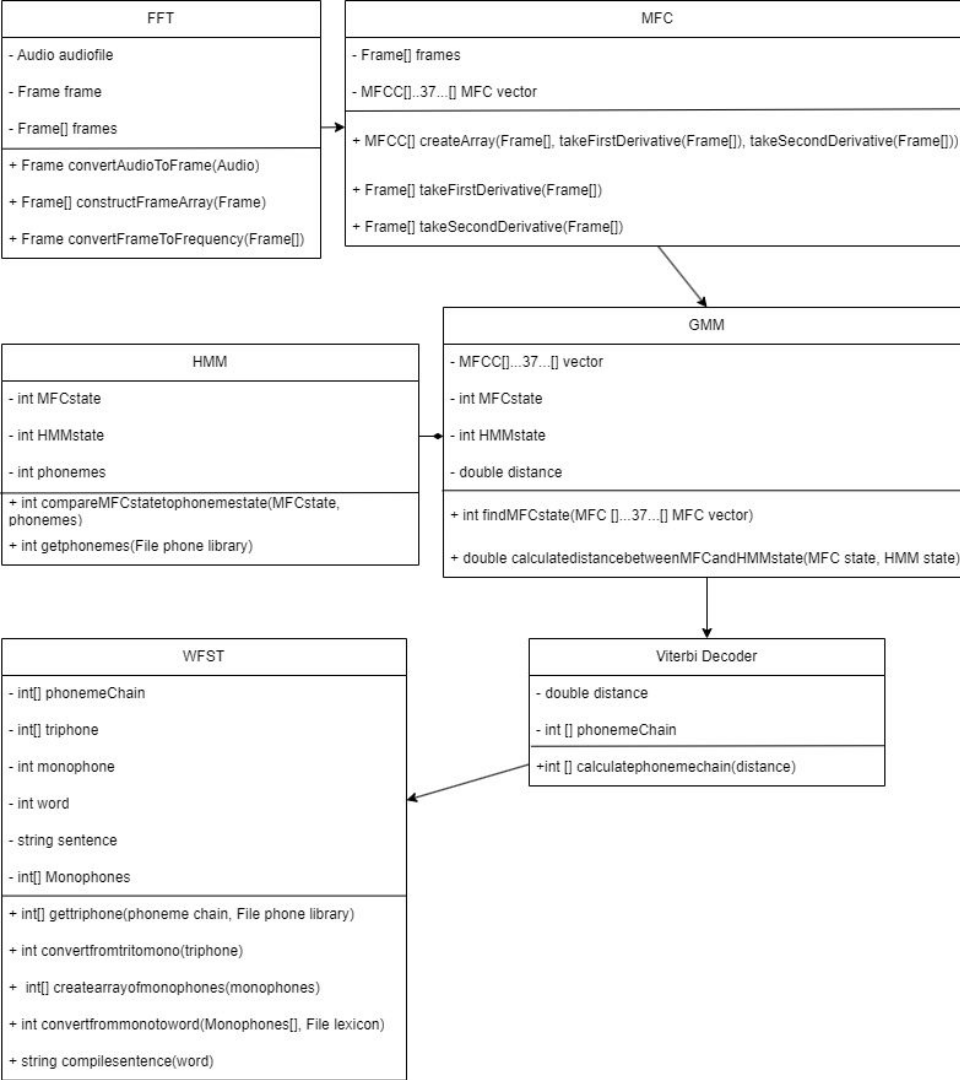


# Use Case Diagram: Sample ASR



# Use Case Diagram: Training Process





# Class Diagram

# Future Testing

## Input Testing

- “The system shall return exceptions/errors if the input is a non-WAV file.”
- “The system shall return exceptions/errors if there are multiple inputs.”

## Accuracy Testing

- Assess transcription accuracy of ASR model using predefined benchmarks (e.g., a minimum transcription accuracy of 80%)

## Runtime Testing

- Evaluate efficiency of ASR model based on processing time
- Measure time taken by ASR model to process standard WAV file and produce output text file

# Lessons Learned

## Complexity of Speech Recognition

- Black box that became slightly more translucent
  - Spaghetti code
  - Bad documentation
- Nuances in human speech
  - Pronunciation (e.g., regional accents)
  - Articulation
  - Volume
  - Pace

## Hardware Limitations

- High performance requirements
  - Training speed dependent on GPU speed

# Next steps

Fall 2023	Spring 2024	202X	202X	202X	202X
Understand current ASR model (on going process)	Adjust the model using the 30-hour ATC dataset	Compare the performance of trained models	Experiment with other models	Apply the best trained model	Experiment with callsign and frequency identification



Questions?