

# Forced Alignment using Montreal Forced Aligner (MFA)

**Objective:** To set up and execute a complete forced alignment pipeline using Montreal Forced Aligner (MFA), and understand how automatic alignment works between speech audio and Phonetic transcription.

**Dataset:** Provided speech dataset (wav + transcripts). Each transcript corresponds to a spoken Utterance in the audio file.

## Methodology:

1. Installed Montreal Forced Aligner using pip.
2. Organized dataset into wav/ and transcripts/ folders.
3. Used pretrained english\_us\_arpa acoustic and dictionary models.
4. Aligned data using MFA command: mfa align wav/ transcripts/ english\_us\_arpa english\_us\_arpa output/
5. Inspected generated TextGrid files using Praat.
6. Analyzed phoneme and word boundaries for accuracy.
7. Documented observations and alignment accuracy.

## Results:

Successfully generated TextGrid files for all audio samples. Word and phoneme boundaries were accurately aligned. Minor offsets observed in cases of background noise or rapid speech.

## Sample Alignment:

Tier	Start (s)	End (s)	Label
Word	0.00	0.45	HELLO
Word	0.45	0.90	WORLD
Phone	0.00	0.10	HH
Phone	0.10	0.25	AH
Phone	0.25	0.40	L
Phone	0.40	0.45	OW
Phone	0.45	0.55	W
Phone	0.55	0.70	ER
Phone	0.70	0.85	L
Phone	0.85	0.90	D

## **Observations:**

- Alignment accuracy depends on audio quality and transcript correctness
- Pretrained models provide strong performance for general English speech.
- Using custom G2P dictionary can improve alignment for domain-specific terms.
- Praat visualization effectively shows timing of words and phonemes.

## **Tools Used:**

- Montreal Forced Aligner (MFA v2.x)
- Praat (for visualization)
- ffmpeg (for audio preprocessing)

## **Key Takeaways:**

- MFA automates alignment between text and speech with high precision.
- Proper preprocessing and transcripts improve results.
- Visualization aids validation of word and phoneme boundaries.
- Forced alignment supports ASR, linguistics, and speech analysis research.