Localized Speech-to-Text for Inclusive Learning

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# Introduction

The most natural way for communication as humans is speech, and there have been constant efforts to make speech a viable and effect way of communication with computing devices, this is achieved with the ever-evolving toolset known as speech to text. These tools constantly evolving as they are , have failed to account for variations in dialects [3] and pronunciation of words along with being available for regional and vernacular languages which varies the impact and effectiveness of these tools from region to region [4]. In this context, it is empirical that we strive to improve these tools to overcome this disparity, making them available to all irrespective of region or language.

In light of the pressing need for these technological tools to be not just widely accessible, but also effectively utilized across diverse regions and languages, this paper puts forth a novel proposition. It suggests the deployment of an on-device, self-supervised [5], speech-to-text module that has been localized specifically for the multitude of regional languages spoken across India. The primary objective of this module is not merely to transcribe speech to text, but to serve a greater purpose - to act as a facilitator in the learning process. By catering to the unique linguistic nuances of regional Indian languages[6,7], this module aims to bridge the gap between technology and effective learning, thereby making education more inclusive and comprehensive.

Here we strive to implement the following tools to facilitate our goal towards a smart and inclusive classroom :

* Discussion Logs : Leveraging Audio Fingerprinting and text independent speaker recognition systems [8] to keep an individualised, speaker separated log of the class room discussions, which will enable a quick and easy review of the classroom discussions.
* Summarization : Using the discussion logs made, prepare a short summary to assist in quick review.
* Individualised Remainders and Summary : Leveraging the discussion logs to make a personalised summary tailored to the individual comprising of individual assigned tasks and conversations.
* Live Captioning and Translation : Enable Seamless Real Time Captioning [9] and Translation of the classroom to the user’s chosen language.
* Evaluation Aid : Identify the key words used in response to a question and cross referencing [10] it with the expected key words and helping the evaluator during a Viva.

The Scope for the use of this tool is vast and the speech to text module proposed in this paper can be leveraged to assist beyond these tools implemented herein.

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