

2. Tesseract OCR Engine Module

The binarization method for the captured image is completed in this section and is evaluated after the text form.

ords and lines have been found after blobs detection. The term or string is passed to various phases. Every letter or character is extracted from the string in these phases, which analyzes the need to attach the extracted character or to disrupt related characters. At last, extracted characters are identified with embedded fuzzy features coordinated with a specific Unicode character language training data. Then after each character is joined to make a word in the same order as extracted and matched with the words of the particular language dictionary.

3. Dictionary words Matching Module

Each group of a word sequence is sent to Dictionary software installed on the Yandex server, where it will help in recognizing the term more precise and accurate than just providing the outcome of a meaningless word. Now, the identified text will be sent to the next module that is Unicode text post processing [5, 6].

4. Unicode Text Postprocessing Module

In the final module, Unicode characters will be displayed after the recognized text is passed to this module, and the user has the privilege to convert the identified text into a particular known language by selecting from the drop-down list under the settings [7, 8].

B. Hybrid Translation Module

Hybrid machine translation is an approach in which more than one machine translation strategy is used. The main idea for developing a hybrid model is to reduce the failure of any single machine translation approach. In the proposed system, we are integrating our system with the YANDEX TRANSLATION server, which uses the CatBoost algorithm. The CatBoost algorithm is a collection of statistical machine translation and neural machine translation [9, 10].

▪ Statistical Machine Translation

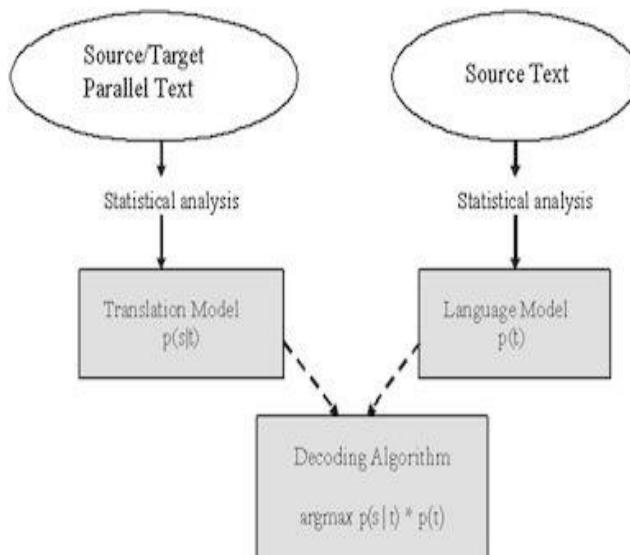


Fig.2. Block diagram of Statistical Translation

▪ Neural Machine Translation

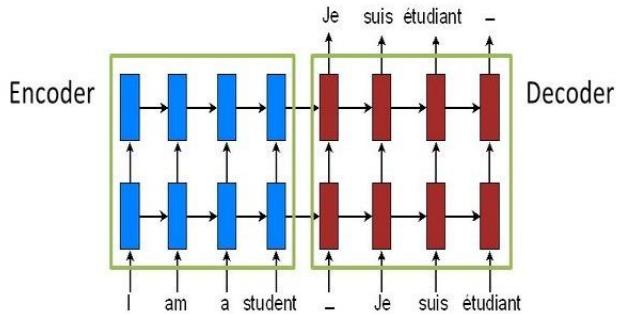


Fig.3. Neural Translation

III. PROPOSED METHOD

The system captures images of the signboard via a smartphone camera. Captured image forwarded to the main server called the Yandex server. Then it is pre-processed and recognized after extraction of the text. After that, the identified text is translated into the target language and appropriately conveyed for user convenience. This process is reflected in Figure 1. The above process contains (i) Acquisition of signboard (ii) Text pulling and recognition, (iii) Text/Voice Translation.

C. Acquisition of Signboard image

For capturing the signboard image, we use a handheld camera. To obtain good quality images, the gap between the mobile and the signboard should be up to 10 meters. The photos are taken with a smartphone camera; they are often noisy and contain complicated environments.

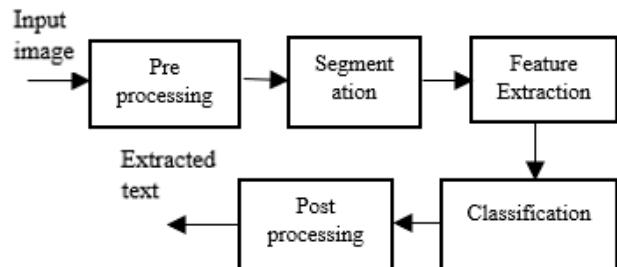


Fig.4. Optical Character Recognition

D. Text Pulling and Recognition

Captured image text pulling and identification are termed as the Optical Character Recognition (OCR). We use a Tesseract OCR engine in this article to get the text from the picture you have acquired [11, 13]. The process of optical character recognition (OCR) is broken down into the various steps: Pre-processing, Segmentation, Feature Extraction, and classification, Postprocessing.

1. Pre-processing of Image

Pre-processing consists of the enhancement of image contrast [12], removal of noise, binarization, and smoothing. For better contrast, the images are equalized according to the histogram. Adaptive histogram equalization increases contrast of image and reduces image imperfections.



In order to remove noise content, improved images are additionally processed with a central filter. A local threshold process is used to produce a binary image as a result. Binarization transforms a gray or color image into a binary image, where the background is preferably represented by one color and the text by another. Smoothing finally that implies both filling and thinning. The standardization is applied to get the uniform size, slant, and rotation characters.

2. Segmentation

Image segmentation is a common technique used in the processing and analysis of digital images to divide images into multiple parts or regions, often based on the image pixels. Image segmentation could be based on color-or shape similarities between the foreground and the background or cluster regions.

3. Feature Extraction and Classification

The characteristics that are important to classify characters at the stage of recognition are extracted at this stage. Classification is the method of defining every character and assigning the appropriate class of character to it. The approaches to classification are in two types. 1) Decision-theoretic methods: Statistical methods, Artificial neural networks and Minimum distance classifiers are the principal approaches to decision-theoretical recognition. 2) Structural Methods: Similarity actions can be formulated using grammar concepts based on the relationship of the structural component. Assume we have two distinct character classes that are produced by S1 and S2 grammars. Given an unnamed character, we say that if S1 generates it, but not S2, it is more like the first class.

4. Postprocessing of Image

This stage performs symbol clustering and error handling. The way symbols are grouped into strings is commonly called grouping. It depends on where the symbol is found in the text. Symbols that consider themselves pretty similar are grouped.

E. Text/Voice Translation

The Yandex translator (CatBoost algorithm) is used for text and voice translation, and we have used Google speech recognition to identify words or characters spoken by the user.

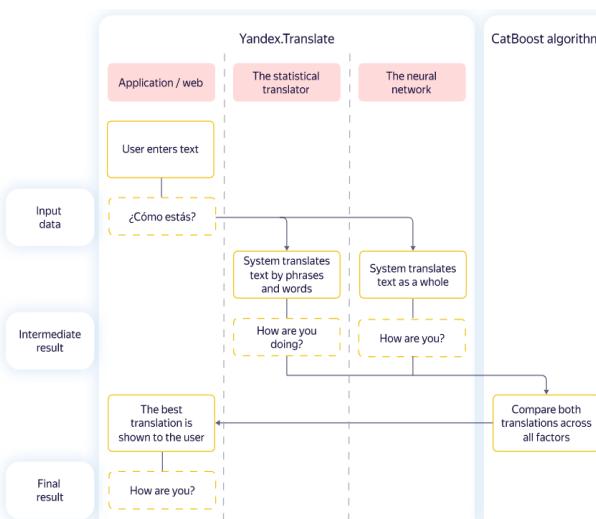


Fig.5. Shows how this translator works

You have to click on the text translator option next; you have to choose the desired language in which you want that text, you have to type text or speak from your microphone and click on the go. Take your data as an input by the app interface, and it will send it to the Yandex API Server where your data will be processed, and then it will fetch by the app interface, and then it will display the result in the selected language. This application supports about 50+ languages across the world and able to recognize nearly all possible standards symbols.

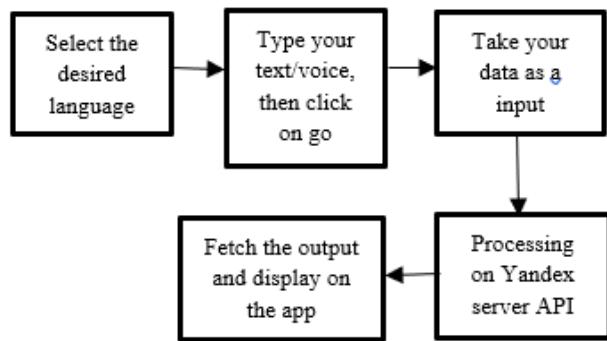


Fig.6. Text/Voice Recognition

IV. EXPERIMENT RESULTS

This specific portion explains the results of the work we have done. Two criteria are used to calculate the performance of the proposed system, namely Precision and Accuracy.

▪ Precision

This can also be interpreted as identification levels for characters. Term precision is defined as the number of characters recognized correctly to the total number of characters tested where tc is the number of characters recognized correctly, and Tc tells the total number of characters

$$\text{Precision} = tc/Tc$$

▪ Accuracy

It is defined as the number of correctly recognized images (signboard) to the total number of test images (signboard). Here tw is the number of properly recognized characters, and Tw is the number of total characters.

$$\text{Accuracy} = tw/Tw$$

Up to 80 signboard images have experimented with our approach, and near about 72 instances are accurately converted to the language you want. That is, the machine reaches 90 percent accuracy. Figure 7 illustrates the image and the results achieved from the sample sign. We have experimented with this method with different fonts for the text, and the device recognizes and correctly translates them. Apart from that, we have checked the voice translation feature over more than 40 voice samples of different-different languages and found this application 94.4% accurate.

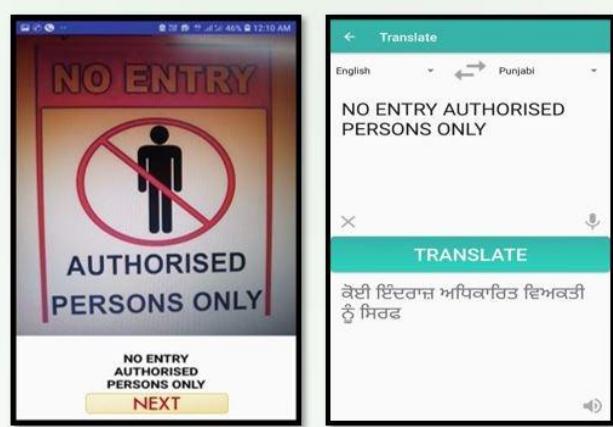


Fig.7. (a) Captured image of Signboard. (b) Extracted text is in English and converted into Punjabi.

V. CONCLUSION

The proposed system successfully translated 50+ languages practiced and spoken across the globe, and results meet our expectations. We have used a hybrid machine (CatBoost algorithm) translation instead of single machine translation to make better use of technology and found that hybrid machine translation is better than single machine translation in terms of accuracy and efficiency. At last, we can say that this application will be more helpful to minimize miscommunication among people due to the language barrier.

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