

# Review of Feature Extraction and Classification Techniques for OHCR in Indian Scripts

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**Abstract:** The development of OHCR (Online Handwritten Character Recognition) is an interesting area in the field of Pattern Recognition. Sustainability of traditional script input methods needs attention in the preservation of neutrality of languages. In the context of Indian scripts, automatic recognition of handwritings contributes a lot in preserving its vital aspects. The South Indian family of languages (Malayalam, Tamil, Kannada, and Telugu) share characteristics of ancient Brahmi script from which they are derived. Research work for accurate OHCR engines for various Indian scripts is still in progress. The paper is intended for familiarizing the works in Online Handwriting Recognition of various Indian scripts. Various features and classification techniques used with recognition accuracies were compared from existing literature. This will be beneficial for the quality selection of features and classification techniques in future works.

**Keywords:** OHCR, Feature Extraction, Indian scripts, Handwriting Recognition

## I. INTRODUCTION

Methods for annihilating the gap in human-machine coalition has decades of sophisticated studies. With the invention of computers the promptness for a method to input diverse scripts span around the globe has been a necessity for the entire community apart from the scientific realm. In order to cope up with the levels of the society, the preservation of traditional script input methods obtained relevance in machine interaction.

Inputting script in one's own language is a difficulty associated with data processing machines. This will be a weighty task in India, a multi script-language country. An easy way to input regional languages in a traditional way as inputs for data processing is desirable. This leads to the studies for natural interface for data input to the machine.

The force thus made Online Handwriting Recognition a frontier area of research in computer science. In OHCR, traditional handwriting input is linked with modern technologies to preserve the natural script input style. Works are still in progress for high recognition OHCR engines for Indian languages.

The rest of the paper is arranged as Online Handwriting Recognition, Indian languages, Features and Classification methods used in OHCR of various Indian scripts and finally the conclusion and future works are mentioned.

## II. ONLINE HCR

Online Handwriting Recognition involves writing on digital touch pads using stylus, writing in paper using digital pen or writing in touch screen displays using fingers and automatically recognized by an OHCR engine.

The strokes corresponding to each character includes coordinate points along the path, pen up, pen down information, time sequence and structural information [1]. The specialty of online handwriting recognition is the preservation of traditional styles followed in the past era. Even though the writer is ignorant of technologies, he can enter the scripts in the data processing machine.

Online handwriting replaces keyboards, which is the commonly used input device for data entry. The complexity of using various scripts [2] in keyboards can be eliminated through online handwriting recognition. Nowadays online recognition is a popular method in mobile devices, which shows the wide admissibility of the technique.

### A. Steps in OHCR

In [3] Charles C. Tappert, Suen, C. Y. and Wakahara, T describes the state of the art of Online Handwriting as an active research area for decades to come. Online HCR is a sequential process comprising Data acquisition, Pre-processing, Feature extraction, Classification, Post processing. Data acquisition involves obtaining pen trajectories of the writings. The data is acquired as a sequence of points with time and device specific values. The acquired points were pre-processed for removing noises and structural variations pertaining to user writing styles.

The data acquisition is a complex task in Online Handwriting and certain issues pertaining to them is to be addressed. Data collected from the users may be affected with a lot of noises and must be cleansed before analysis. Pre-processing varies in various scripts and usually

includes duplicate point elimination, de-hooking, smoothening, resampling and normalization [4]. The features are the essential structural and statistical properties which describe the character and the features extracted from the pre-processed co-ordinate values are used in the classifiers. Quality selection of feature affects classification rate [5], hence the features must be selected based on the classification technique to be employed. Commonly used features are coordinate positions, bumps, crossings, cusps, loops, writing directions, angles and aspect ratios [6].

Classification includes the problem of recognizing to which of the available classes a new class member fit. Commonly used classifiers are Hidden Markov Models (HMM), Support Vector Machines (SVM), K Nearest Neighbour Classifier (k-NN), Neural Networks (NN), Dynamic Time Warping (DTW), Fourier Descriptors and Daubachies Wavelet Transform (DWT) [7]. Post processing is done after classification for the alphanumeric code matching of characters and any dis-ambiguity in confusing sets and applying language specific constructs to improve the recognition phase.

### III. OHCR IN INDIAN SCRIPTS

In the following section, we present a review on various techniques proposed in the literature to recognize online Indian scripts. We outline the contributed works for major Indian scripts as follows.

#### A. Devanagari.

One of the earliest work in Devanagari OHCR has been presented by Scott-D-Connell et al in 2000 [8]. They have worked with both online and offline HCR using five different classifiers. By using five classifiers they have achieved the average accuracy of 86.5%. A.M Nambodiri et al used 11 spatial and temporal features extracted from the strokes of the words for recognizing online Devanagari script [9]. They have attained an accuracy of 87.1% on a database of 13,379 words. Joshi et al developed a test vector is based the structural features, such as mean (x, y) values, positional cues and directional codes at the stroke level [10]. A subspace method is used by the system for classification in which each character class is represented by a basis vector which is a set of N eigenvalues. The system has achieved the accuracy of 93.05% on a set of 100 frequently occurring characters.

Abhimanyu and Samit Battacharya proposed a stroke based recognition using Hidden Markov Model [11]. They focussed on recognition of isolated Devanagari characters in the iPhone. In another work Swethalakshmi proposed a two stage proximity analysis technique for the identification of Devanagari characters from a sequence of strokes. Strokes are represented based on spatiotemporal, spectral and spatiostructural features are explored. They have achieved the average accuracy of 95.29%. Lajish V.L and Sunil Kumar presented an online

handwritten Devanagari script recognition [12]. They introduced Fuzzy Directional Features (FDF) which can produce directional variance in the handwritten primitives. In another work they introduced a new online feature set called the extended directional feature which can discriminate similar looking strokes [13].

#### B. Bangla

U Garain et al proposed a work in Bangla character recognition which is focussed on utilizing the cues from the pen trajectory to derive features when tackling the problem of stroke order variation [14]. Directional code feature has been introduced in a work proposed by Bhattacharya et al for recognition of Bangla OHCR [15]. MLP has been used as the classifier. A string of features are derived at the sub stroke level by Nakai et al [16]. Based on the similar shape of the graphemes that constitute the ideal character shape. The classification algorithm used is HMM.

T Mondal et al presented a comparative study of the performance of the classifiers HMM and a nearest neighbour classifier (based on DTW) [17]. An analytic recognition approach depends on the position of the headline, is adopted by Bhattacharya et al to segment the input word to a set of substrokes [18].

#### C. Malayalam

As one of the initial work in Malayalam, Sreeraj M., and Idicula S proposed an OHCR system with normalised (x, y) coordinates and context bit map features [19]. The classification technique used is Kohonen networks. The system exhibited a recognition accuracy 88.75% with a recognition time of 15-32 milliseconds. A combination of time domain features and dynamic representation of writing direction along with its curvature is proposed by Sreeraj M., and Idicula S in another work [20]. The features used are writing directions, normalised (x, y) coordinates and aspect ratio of curvature. By using KNN classifier, the system achieved an accuracy of 98.125% even with a small sample size.

A recognition engine using wavelet transform is presented by Primekumar K. P. and Idiculla S. M. [21]. Time domain features of the strokes are used with wavelet transform. The classification algorithm used is simplified fuzzy ARTMAP network, which requires comparatively very less time for training. An accuracy of 97.81% is reported in the work.

Indhu, T. R., and Bhadrans, V. K presented an online handwriting recognition system using simplified fuzzy ARTMAP technique [22]. Enormous features are used in the work which includes (x, y) coordinates, start quadrant, end quadrant, horizontal and vertical point density, loop, cusp, stroke length. The system achieved 98.26% recognition accuracy with simplified fuzzy ARTMAP classification technique. Prime Kumar, K. P., and Idiculla, S. M. extended their work using HMM and SVM [69]. They achieved an improved accuracy of 97.97%

for SVM using Gaussian kernel. But the recognition rate dropped to 95.24% for HMM and recognition time was also increased compared to SVM.

Chacko, B. P., and Babu Anto P achieved 96.83% recognition accuracy using OSELM and SLFN by division point features [24]. Sampath A., Tripti C., and Govindaru V presented a Neural Network based model for handwritten recognition [25]. The direction information of the written character is recorded based on the 8 connected Freeman chain code. The direction of the pen movement is recorded as feature vector. Back propagation Neural Network is used for classifying characters. Additional disambiguation technique is used in post processing stage to identify confusing pairs. A recent work of Steffy Maria and Joseph achieved a recognition accuracy of 90% using SVM. The directional and curvature features are extracted and trained in LIBSVM tool available in Matlab [26].

#### D. Tamil

Majority of the work in Dravidian scripts have been reported in Tamil. In the work, Niranjana Joshi., Sita G., Ramakrishnan A. G., and Madhvanath S demonstrated a template based elastic matching technique [27]. The features used are (x, y) coordinates, quantized slopes and dominant point coordinates, the system combined different features and formed seven different schemes. The work describes recognition accuracy, recognition speed and number training templates with dynamic time warping distance measure. The system achieved a maximum of 95.90% accuracy with a speed of 32.6 characters per second.

A novel approach has been described by Aparna K. H., Subramanian V., Kasirajan M., Prakash G. V., and Chakravarthy V. S with a string of shape features [28]. Using this string representation, an unknown stroke is identified by comparing it with a database of strokes using a flexible string matching procedure and finite state automation. A recognition rate of 77.84% has been achieved by the system. Neural networks have been adopted in Ishwarya and Kannan R. J. with multilayer perceptron [29]. The feature used is Fourier descriptor. Test results indicate that Fourier descriptors with back propagation network provides a recognition accuracy of 97%. Shashikiran K., Prasad K., Kunwar R., and Ramakrishnan, A. analysed the performance of Tamil Scripts with HMM and Statistical Dynamic Time Warping (SDTW) [39]. The results show that the best result of 85% for HMM is possible using a simpler SDTW model.

The work of Murthy and Venkatesh N involves two stage recognition with PCA and NN classifier adopted at the first stage followed by combined feature combination and DTW [30]. The system uses (x, y) coordinates, quantised slope, coordinates of accurate dominant points and Quartile Features. Based on the primary classifier output and prior knowledge, a classifier is chosen for the second phase.

The system achieved 90.2% recognition accuracy for Tamil Scripts. Rituraj Kanwar and Ramakrishnan A. G has projected a fractal coding method with features as fractal dimensions [31]. This technique exploits the redundancy in data, thereby achieving better compression and usage of lesser memory. The fractal code features were applied in separate classifiers (HMM, SVM, SDTW). The recognition achieved an accuracy of 90% with SDTW.

#### E. Kannada

Wavelet features of the Kannada script is used in the recognition method proposed by Samuel R. D. S., Srinivasa Rao Kunte [32]. The system achieved a recognition accuracy of 95% for basic Kannada characters. The structural and spatio-temporal features are employed in a work (Prasad M. M., Sukumar M., and Ramakrishnan G. [33]) in the recognition of Kannada Scripts. The strategy used here is Divide and Conquer. A character is segmented into three strokes units and recognized separately. The method is proposed to overcome the complexity of huge number compound characters. The segmented units are preprocessed and the extracted features are mapped to sub-space using PCA. The system achieved an accuracy of 81% with KNN classifier.

Kunwar R., Mohan P., Shashikiran K., & Ramakrishnan A. G [34] presented an unrestricted recognizer using a smoothed first derivatives of (x, y) coordinates and SDTW as the classifier. The system achieved 88% of recognition accuracy with faster recognition time compared to the recognition time of DTW.

A recognizer for mobile devices has been suggested by Keerthi Prasad, Imran Khan and Naveen R Chanukotimath [35] using two approaches PCA and DTW. The reported recognition accuracy is 88% for the PCA and 64% for DTW. The recognition time of PCA approach is 0.8 seconds which is fairly better compared to DTW with a recognition time of 55 seconds. Venkatesh N and A G Ramakrishnan proposed a novel method with two stage classifiers [30]. The features used are quartile features, preprocessed (x, y) coordinates, quantized slopes and accurate dominant points. The primary classifier used is DTW. Based on the result of primary classifier and prior knowledge, secondary classifier is applied to obtain a recognition rate of 92.6% for Kannada characters.

#### F. Telugu

H. Swethalakshmi developed a system for online HCR of Telugu and Devanagari writing using Support Vector Machines (SVMs) [36]. Each stroke is represented as an n-dimensional feature vector depending on the choice of the number of points for stroke representation. The features chosen to represent the curve are the coordinates of points in the pre-processed stroke. The stroke is then classified using SVM with 73.30% of accuracy. Babu V. J., Prasanth L., Sharma R. R., and Bharath, A presented an online handwritten symbol recognition system for Telugu

is based on Hidden Markov Models (HMM)[37]. Normalized x-y coordinates, normalized first derivatives, normalized second derivatives, curvature and aspect ratio are used as the features. The system achieved the accuracy of 98.7% on a dataset containing 29,158 train samples and 9,235 test samples.

Mandalapu D., Prasanth L., Babu V. J., Sharma R. R., and Rao G. V. P. described character based elastic matching using local features for recognizing online Telugu handwritten data [38]. Dynamic Time Warping (DTW) has been used with four different feature sets: x-y features, Shape Context (SC) and Tangent Angle (TA) features, Generalized Shape Context feature (GSC) and the fourth set containing x-y, normalized first and second derivatives and curvature features. Nearest neighbourhood classifier with DTW distance was used as the classifier with the recognition accuracy of 89.77%.

#### IV. CONCLUSION

The work presented a comparison of different recognition techniques used in OHCR of Indian scripts. Maximum recognition rate of 98% is reported with DFT, DCT and SVM classifier in Tamil script. In Malayalam, structural and directional features on SFAM technique produced a recognition rate of 98.26%. The existing literature in Telugu reported a recognition rate of 98.7% with HMM classifier and frequency, time domain features. An accuracy of 95% is reported for Kannada characters with wavelet features and Neural Networks. The maximum recognition rate attained in various Indian scripts uses different methods. Most of the classifiers used structural features or its refined form extensively. No similarity in the recognition scheme is visible in maximum recognition rate. But majority of the work employs DTW and SVM. The comparison of various recognition schemes reveals that feature selection has a major role in recognition accuracy in Indian scripts. A method used in one of the script gives lesser recognition rate in another script is also observed.

The study recommends that even better recognition methods are needed to achieve more promising recognition rate with minimum time for both writer independent and dependent OHCR methods in Indian scripts. The future works in OHCR of Indian scripts may be focused on feature dimensions, class size and recognition time with optimum standards and Benchmark databases.

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