SVM CLASSIFIER FOR THE PREDICTION OF ERA OF AN EPIGRAPHICAL SCRIPT

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ABSTRACT

Inscriptions are the main source for reconstructing the history and culture of ancient civilizations. The scripts of modern Indian languages have evolved over centuries and finally transformed to the present form. Modern readers find difficulty in interpreting a script of olden days. The characters have changed over time. Hence for reading ancient scripts the period has to be determined, so as to have knowledge of which character set of ancient days is to be employed for automatic reading. Prediction of the era of a given ancient script is a follow-on member of an ancient script recognition system and can be used as a component of the OCR system for ancient scripts. This knowledge can be used by archaeologists and historians for further explorations. In this paper we demonstrate period identification of various ancient Kannada scripts using SVM classifier. A system is proposed for prediction of the era and it is being done by examining a few characters in Kannada script of various periods referred to as test characters. These test characters are sampled from the script automatically and matched with the characters available for different periods using machine intelligence.. This classifier is tested on quite number samples of Kannada epigraphical document images belonging to nine different periods.

Keywords

Inscription, Epigraphy, Paleography, Support Vector Machine (SVM), Optical Character Recognition (OCR)

1. INTRODUCTION

Inscriptions have remarkable importance to mankind. The information culled out from the inscriptions provides us the knowledge of history, culture, astronomy, medicine, management, political, religious, social, economic, tax, administrative, educational conditions and also languages and scripts that prevailed during yester years[1][2][3][4]. **Epigraphy** is the science of identifying the inscriptions on rocks, pillars, temple walls, copper plates and other writing materials. It is a primary tool of archaeology when dealing with literate cultures. Many inscriptions do not contain enough historical details to fix their authorship conclusively. For example, from the inscriptions with the name Rajaraja, it is not always clear whether Rajaraja the First or the Second or the Third is intended. To assign dates to such inscriptions and to identify the kings, palaeography is the main tool. **Paleography** is the study of ancient handwriting and the practice of deciphering and reading historical manuscripts. The paleographer must have the knowledge of: first, the language of the text and second, the historical usages of various styles of handwriting, common writing customs, and scribal/notarial abbreviations.

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Scripts denote the writing systems employed by the languages. Any language can be written in any script. Having or not having 'own script' is neither a status nor any impediment for language. Three important varieties of scripts that were prevalent in ancient India were: Indus valley script, Brahmi Script and Kharosti script. The scripts of modern Indian languages have evolved from one of these scripts over the centuries. The evolution of the script is dependent on many factors: the writing material, (Stone, Copper, Palm leaf, Paper etc), writing tools, modes of writing and the background of the scribes. In India, prior to invention of writing / printing papers, Palmyra leaves and birch leaves were used for writing purposes. As they could not be long lasting, engraving on rocks, pillars and plates made of copper/ gold / silver came into practice. In India currently there are 13 scripts and 23 official languages for communicating at state level. Apart from these, there are many languages & dialects used by number of people.

The Kannada script has been used to write in Kannada language. Kannada which belongs to highly acclaimed Dravidian language family is the official language of the State of Karnataka and is one of the most enriched languages in India with is long historical heritage. Its earliest written records dates back to about the third century B.C. and the modern script has evolved gradually from the ancient script known as Brahmi. Kannada script has been evolved in its present form, undergoing several twins and turns. In fact, the Indian linguists have demarcated the whole of this evolutionary process in to four broad phases.

- **Poorvada Halegannada or Pre-ancient Kannada** The first written record in the Kannada language is found in Emperor Ashoka's Brahmagiri decree, which dated back to as early as 230 BC.
- Halegannada or Ancient Kannada- marks the second phase, covering a time period of the 9th to 14th centuries CE.
- **Nadugannada or Middle Kannada-** third phase started from the 14th century and continued till 18th century CE.
- Hosagannada or Modern Kannada the Kannada works that were produced at the end of the 19th century and also much later are categorized as Hosagannada or Modern Kannada.

Kannada script has traveled a long way from the earlier Brahmi model as indicated in Figure 1. The fonts have evolved over the centuries and it has undergone many changes till now. It has undergone a number of changes during the regimes of Shatavahana, Kadamba, Ganga, Rashtrakuta, Chalukya, Hoysala, Vijayanagara and Wodeyar dynasties, and transformed finally into a shape which we are using today [2]. This poses a challenge in understanding these ancient scripts. The stages of evolution, though generally understood by epigraphists, are not known to the educated layman of that province. There was also a certain amount of regional variation. Thus even for the experts, it is difficult to assign dates to many inscriptions, whether complete or fragmentary. The expert epigraphists, who decipher these scripts and translate them into the regional languages, are expected to become extinct in near future and also the importance of inscriptions to mankind is remarkable. Hence there is a dire need for automation of classification and recognition of ancient epigraphical scripts. This work helps us in integrating the new technological concepts with underlying cultural features of the scripts, thereby minimizing the divide between "Man and the History". The age of the document reflects to which era the script belonged to and the particular dynasty in which it existed.

In automation of deciphering epigraphical scripts, the characters present in the script are to be identified and recognized. The characters have evolved over years and transformed to the shape we are using today. Every era has its own character set, which is to be determined before

recognition. Hence there is a need for dating of inscriptions to a particular period, which is the initial step towards decipherment.

The ancient scripts are engraved on rocks, copper, silver or gold plates. The images of these inscriptions are captured and later subjected to digitization process. The script of document images that has been digitized is to identified and later recognized. Hence the present problem addresses off line character recognition. Handwriting identification and recognition are of great practical interest in the extraction of discriminating and invariant information from a handwritten specimen. One of the major difficulties in off-line word recognition originates from the great variations observed in different samples of scripts from the same writer over time or from different scriptors. There is no perfect mathematical model that can describe such extreme variations and hence it is very difficult to find the characteristic features that are invariant with different writing styles.



Figure 1: Changes in character shape since 3rd century B.C to

18th century A.D

2. LITERATURE REVIEW

High accuracy OCR systems are reported for English with excellent performance in presence of printing variations and document degradation. Recognizing English characters is much simpler as there are only 26 letters and each letter is quite distinct from others compared to recognition of Indian language characters. For Indian and many other oriental languages- OCR systems are not yet able to successfully recognize printed / handwritten document images of varying scripts, quality, size, style and font. Compared to European languages, Indian languages pose many

additional challenges. Many researchers have been working on script recognition for more than three decades but there are very few tools to identify these scripts. Indian languages are characterized with the properties:

- (i) Large number of vowels, consonants, and conjuncts.
- (ii) Have a base character along with vowels attached, forming single character called the compound character.
- (iii) Most scripts spread over several zones.
- (iv) Lack of standard test databases of the Indian languages.

There are many conventional OCR systems available for present day Kannada Script, but very few work on ancient Kannada epigraphical scripts are reported. Ancient Script recognition poses additional challenges.Some of researchers are making special efforts to make this knowledge more easily available.Nevertheless, it remains to be one of the most challenging problems in pattern recognition.

Anasuyadevi [2000] has worked on recognition of ancient Indian scripts [5]. She has proposed [2003] a fuzzy neural network approach for the recognition of Brahmi characters.

A hybrid neural network architecture for age identification of ancient Kannada Scripts was proposed by K Harish Kashyap, Bansilal, P Arun Koushik [2003]. After pre-processing the characters, the work is implemented in two phases. The first phase which identifies the base character incorporates an Artificial Neural Network (ANN). ANN is trained by Back propagation algorithm to identify the present day base character corresponding to input character. In the second phase, for identification of age pertaining to the base character, a Probabilistic Neural Network (PNN) - a Bayesian classifier is used taking the advantage that no training is involved prior to classification [6].

The research work by Srikanta Murthy K [2005] for transforming epigraphical objects into machine recognizable form, involves the preprocessing techniques - for removal of noises, segmentation of lines and characters, thinning and classification of the epigraphical document belonging to different period [16]. Two preprocessing techniques for removal of noise have been proposed – the first algorithm is based on a rectangle fitting wherein the height of the character to be retained is assumed to be greater than the noisy pixel. The second algorithm employs a template to obtain the minimum majority of white pixels. Segmentation of lines and characters are carried out using- a Partial Eight Direction Based Line Segmentation (PEBLS) algorithm wherein horizontal Projection profile is applied to identify the base and supplementary reference lines. The second approach is based on Nearest Neighbor Clustering (NNC), could be used even when the document is skewed. Three thinning algorithms - two-step algorithm, fully parallel thinning algorithm and rotation invariant 4- step algorithm have been designed. Classification of the epigraphical document belonging to different period is carried out using - a method based on texture features, a method based on invariant moments which are invariant to rotation, translation and scaling, and for accurate estimation of the period, a neural network based approach is adopted [17][18][19].

3. PROPOSED METHODOLOGY

Age identification of ancient scripts is important, which enables to know which character set is to be employed for automatic reading of age old scripts. It is a problem characterized under the

genre of pattern recognition and image analysis. Figure 2 illustrates the steps towards classification of epigraphical document images to their respective era.

The Age prediction classifier involves the following steps:

- I. *Image Acquisition*: The present work proposes to take the objects of the inscriptions as the inputs. These objects may be camera grabbed or scanned if the inscriptions are engraved on a plate. The photographs of inscriptions on rock or pillars may not result in good objects. Hence estempages are produced [Dani A.H, 1986; Sircar D.C, 1965] and later scanned to get the object image of the inscription.
- II. The input image of the inscription may be degraded due to the presence of the broken characters, erased characters, touching characters, distortion due to fossils settled, irrelevant symbols engraved by the scribes and so on. The non uniform spacing between the lines and characters of epigraphical document and the skew could complicate the process of deciphering the script. Hence *the input document image is preprocessed for removal of noise and skew correction, followed by segmentation of characters*.
- III. Appropriate *Feature extraction method is devised* for measuring the relevant shape information contained in a pattern so that the task of classifying the pattern is made easy by a formal procedure.

The features extracted have the following properties:

- Easy to extract, which reduces the complexity of the program.
- Distinct, which eases the classification process.
- Independent of font type and size, which is a big advantage.



Figure 2. Age Identification System

There are totally14 features extracted from the character of which 4 of them are for the whole image as listed below:

- 1) Height / Width
- 2) Number of black pixels / number of white pixels image
- 3) Number of horizontal transitions
- 4) Number of vertical transitions

In addition, the image is divided into four regions as shown in Figure 3 and the following features are extracted from these regions:

- 5) Black Pixels in Region 1/White Pixels in Region 1
- 6) Black Pixels in Region 2/White Pixels in Region 2
- 7) Black Pixels in Region 3/White Pixels in Region 3
- 8) Black Pixels in Region 4/White Pixels in Region 4
- 9) Black Pixels in Region 1/Black Pixels in Region 2
- 10) Black Pixels in Region 3/Black Pixels in Region 4
- 11) Black Pixels in Region 1/Black Pixels in Region 3
- 12) Black Pixels in Region 2/Black Pixels in Region 4
- 13) Black Pixels in Region 1/Black Pixels in Region 4
- 14) Black Pixels in Region 2/Black Pixels in Region 3
- IV. The characters have evolved gradually over a course of time, leading to varying character sets from time to time. Hence the era to which an epigraphical script belongs to has been predicted so as to further enable automatic recognition of the epigraphical characters. For the classification of the script to their respective period, a Support Vector Machine (SVM) Classifier using linear kernel function is devised. The principle of SVM is to map the input data onto a higher dimensional feature space and determine a separating hyper plane with maximum margin between two classes in the feature space. We use SVM as a multiclass-classifier system which is trained on the distinct/unique characters in the Kannada script of various periods. Each vector is given to SVM which classifies it into a class in the vector space. The important criteria for deciding the final character are: the number of points belonging to a particular class and the minimum distance from a class. It represents the degree to which a particular snapshot resembles a character. We search for the unique (test) characters identified aprior, of various periods in the given ancient script. The number of matches of these unique characters of different era, with the characters of input document image is accounted. The period of input epigraphical document image of ancient Kannada script is the one with maximum number of matches of characters of input ancient script with unique (test) characters. Thus the input document image is classified to the period of any of these dynasties: Ashoka dynasty, Shatavahana dynasty, Kadamba dynasty ,Badami Chalukya dynasty, Rashtrakuta dynasty, Kalyan Chalukya dynasty, Hoyasala dynasty, Vijayanagar dynasty or Mysore Wodeyar dynasty.



Figure 3. Dividing the image to 4 regions and extracting features

V. Display the era of the Script

Thus the input epigraphical document image is classified to the period of any of the nine dynasties that prevailed in ancient Karnataka: Ashoka dynasty, Shatavahana dynasty, Kadamba dynasty, Badami Chalukya dynasty, Rashtrakuta dynasty, Kalyan Chalukya dynasty, Hoyasala dynasty, Vijayanagar dynasty or Mysore Wodeyar dynasty

4. IMPLEMENTATION AND EXPERIMENTAL RESULTS

We have implemented the Age identification System on Window's Xp Operating System, using MATLAB version 7.8.0.347(R2009a) on a PENTIUM IV 2.6 GHz processor. The functionalities of Image processing toolbox is used for developing GUI, preprocessing and segmenting input document image.

In this section we provide sample experimental results out of several experiments conducted. We have tested the proposed method on more than 50 samples of ancient Kannada Epigraphical documents. Figure 4 to Figure 6 illustrates the snapshots of GUI developed, the output of segmentation and feature extraction, and output display of the classifier respectively. Figure 7 to Figure 9 depicts sample results of Classification and Display of the period for the given input epigraphical script.

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AGE IDENTIFICATION OF ANCIENT KANNADA SCRIPT										
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Figure 4. GUI of the Period prediction System for ancient Kannada Script

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Figure 5. Snapshot of Segmentation and Feature extraction for input epigraphical document



Figure 6. Snapshot of classification of input epigraphical document into respective period and corresponding dynasty



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Output image with details pertaining

to the period predicted







Input epigraphical document

Output image with details pertaining

to the period predicted





Input epigraphical document

Output image with details pertaining to the period predicted

Figure 9. A Sample Kannada Script of Wodeyar Dynasty being dated

5. CONCLUSION & FUTURE ENHANCEMENT

Epigraphical sources reflect the importance during regime of several rulers. Script identification is a key step that arises in document image analysis, especially for the documents which belong to ancient days. The study of Epigraphical scripts is vital in knowing the civilized past and hence classification of character belonging to various periods is imperative before using the character bank of the particular period. The proposed system for the prediction of the era is being done by examining a few characters in Kannada script of various periods referred to as test characters. These test characters are sampled from the script automatically and matched with the characters available for different periods using machine intelligence. The proposed system here has various modules like Preprocessing and Segmentation (involves binarization, Noise removal, Skew removal & Segmentation), feature extraction and finally classification of input epigraphical documents into their respective eras using Support Vector Machine. This classifier is tested on more than 50 samples of Kannada epigraphical document images belonging to nine different periods. To sum up, the research issue taken here is to produce a computer perceivable image from a raw epigraphical script which are the inscriptions on rocks or pillars or plate and then predict the era of the ancient script. Prediction of period of ancient scripts is the first step in automatically deciphering epigraphical scripts. Automatic period identification for a given document image, of a script facilitates the selection of the script specific OCR in an environment where scripts of various periods are given as input. The work can be further extended to facilitate the computer dating of inscriptions of various other scripts that were prevalent in ancient India and other non Indian regions, during the regime of various rulers.

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