

Cognitive based Emotion Analysis of a Child Reading a Book

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Abstract - - In this COVID-19 situation many schools have started online coaching. Students are reading many books online. There is no app that capture student's expression based on their reading which might be trouble sometime. It's a problem because children might give wrong expression to the context which he is reading also it needs to be observable that to the sentence he is reading should give correct expression. Sentiment analysis of sentence to the joy, anger, fear, excited, sad, happy are ongoing process. Also understanding human expression from face specifically children is also required. In this project we are going to develop a system which captures a children's expression from their face using CNN and the context they are reading with the help of sentiment analysis.

Key Words: Data Mining, NLP, WordNet, SCAM, KMP

1. INTRODUCTION

Detection of emotion is a difficult area for researchers for a very long time. We express our feelings with facial expressions. Once we interact with others, our expressions show some essential signs such as our level of interest, our willingness to participate in speaking and to respond continuously. It helps in improving social communication. However, there are problems in detecting the emotions of people with autism. Facial expression is the most natural way of inner world revelation. It plays a vital role in our social interactions. With it, we can express our feelings, and infer other people's attitude and intention. We are capable of telling other people's facial expressions at a glance. Facial Expression Recognition (FER) system enables machines to infer our intentions through reading the facial expression, that is of great help for Human-Computer Interaction (HCI). A typical FER system consists of three stages: 1) First, face detection and localization. 2) Next, extract expression information from located faces. 3) Finally, a classifier (like an SVM) is trained on the extracted information to output the final expression labels. We provide a CNN-based approach to address the FER problem in the following three steps: (i) first we describe several different structured CNNs, these subnets are trained separately on the training set; (ii) then these trained CNNs are assembled together by removing the output layers and concatenating the last but one layers together.

1.1 OBJECTIVES

Our main objective of project is

1. Sentence classification: -The paragraph that children reading needs to be classified based on joy, anger, fear, excited, sad, happy. For this purpose, we are going to train the model using SVM, naïve bayes, random forest model. Models will be generated from twitter dataset and sentence will be classified by giving input to this field.

2. Facial expression classification: Facial expression images dataset can be found online for dataset but this dataset is for adult. For children dataset we will capture the face of children. And train the dataset using CNN (convolutional neural network) We will create a model based on face emotion using CNN.

3. Application development: - we will develop flask based application in which we will show a paragraph to person and camera will be started at same time the text will be shown to user at the same time face will be captured and output will be generated log will be maintained and output will be shown at the end time of after 3-4 minute.

1.2 SCOPE

Students: -This project can be used in any educational program where children's expression is important
Paragraph analysis on audience books publisher: -Suppose we want to analyze the books response and expression this can be used for generating books rating.

INPUT: Text of paragraph and face expressions. (Fear, happy, sad, normal, anxiety, anger)

PROCESS: sklearn, OpenCV, sentiment analysis Dlib, face landmark detection python libraries.

1.3 FEATURES

1. This System help to find the emotion of child.
2. It will detect emotions using CNN and SVM algorithms.
3. Emotion analysis based on Sentence is also possible.

2. LITERATURE SURVEY

Identification of Relations from IndoWordNet for Indian Languages using Support Vector Machine

Inspection of the SVM classifier for identification of relations in Indian Languages apart from English SVM successfully implemented for identification of relations in Indian Languages from the IndoWorldNet. This aims at implementing SVM in the field of text classification

Sentiment Classification based on Ontology and SVM Classifier

SVM has been used for sentiment classification by making SVM act as a binary classifier to classify sentiments In the document level and sentence level sentiment classifications assume that each document or sentence focus on a single object and contains only one opinion or opinion from a single opinion holder SVM is better at text classification and use of ontology makes SVM suitable for the application

User Intention Understanding from Scratch

The main evaluation criterion of the performance is the prediction accuracy of all users in the test set. We can observe that on overall, the method Threshold-V achieves the best accuracy of 67 percent. The drawback of the system is that the use of a numerical classifier led to unwanted overheads

The discipline of artificial intelligence is wide-spread across various fields and has a wide variety of sub-domains. They are concisely explained in the diagram that follows:



Figure-1: Classification of Algorithm

Support Vector Machines is a Machine learning algorithm

Which can be used for the following purposes:

- Classification
- Natural Language Processing

It attempts to label the vector according to pre-defined classes using a dividing hyper-plane. Finding the right hyper-plane is searching for the saddle point of the Lagrange function. It is equivalent to quadratic programming containing dual variables. SVM requires solution of the following optimisation problem. This is a

supervised algorithm and hence needs a training data set to train the classifier and then, that classifier can be tested by using the testing data set.

SVM uses specific kernel functions to construct a hyper plane that is used for classification. These kernels are the basis for finding the correct hyper-plane among the many possible hyper-planes in the given situation that divides a particular vector. The 4 basic kernel functions are as follows:

- Linear Kernel
- Polynomial Kernel
- Radial Basis Function Kernel
- Sigmoid Kernel

Any of these kernel functions are used for designing the classifier according to the requirements of the problem. These kernels have the following values associates with it:

- Kernel Function Type
- Values of Kernel Function Parameters
- Values of Regularisation Parameters

These values should be calculating meticulously irrespective of the choice of the kernel because they largely influence the results of the classifier and its accuracy. Any Error or miscalculations in these values can have a heavy toll on the output and the end-results.

LIBSVM

LIBSVM is a library for Support Vector Machines. This package has been actively developed since 2000. This package aims at easily applying SVM to their applications. This library is used in machine learning along with many other areas. It is used for the following purpose

- Support Vector Classification
- Support Vector Regression
- One-class Support Vector Machines

A typical use of LIBSVM involves two steps: first, training a data set to obtain a model and second, using the model to predict information of a testing data set. For SVC and SVR, LIBSVM can also output probability estimates. The LIBSVM package is structured as follows.

- Main directory:

Core C/C++ programs and sample data. In particular, the file svm.cpp implements training

and testing algorithms, where details are described in this article.

- The tool sub-directory:

This sub-directory includes tools for checking data format and for selecting SVM parameters.

- Other sub-directories contain pre-built binary files and interfaces to other languages/software.

3. FLOWCHART

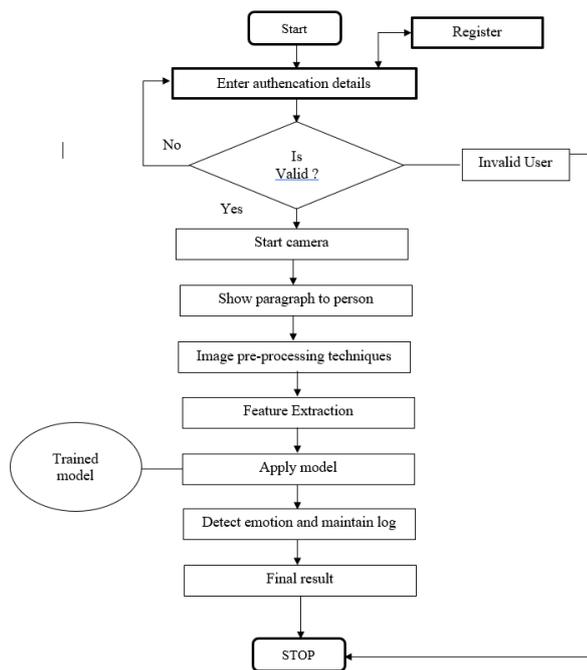


Figure-2: Flowchart

User have to register himself then after enter the authentication details and if the user is invalid then he has to move to reenter authentication details if user is valid then the camera starts and the particular user have to read the paragraph while reading preprocessing technique start its work and extracts the features from the face with the help of trained model. With the help of trained model, we find the actual result.

4. ALGORITHMS

1. SVM (Support Vector Machine)

Support Vector Machines (SVM) have recently gained prominence in the field of machine learning and pattern classification. Classification is achieved by realizing a linear or non-linear separation surface in the input space. In Support Vector classification, the separating function can be expressed as a linear combination of kernels associated with the Support Vectors as

$$f(x) = \sum_{x_j \in S} \alpha_j y_j K(x_j, x) + b$$

where x_i denotes the training patterns, $y_i \in \{+1, -1\}$ denotes the corresponding class labels and S denotes the set of Support Vectors. The dual formulation yields

$$\min_{0 \leq \alpha_i \leq C} W = \frac{1}{2} \sum_{i,j} \alpha_i Q_{ij} \alpha_j - \sum_i \alpha_i + b \sum_i y_i \alpha_i$$

where α_i are the corresponding coefficients, b is the offset, $Q_{ij} = y_i y_j K(x_i, x_j)$ is a symmetric positive definite kernel matrix and C is the parameter used to penalize error points in the inseparable case. The Karush-Kuhn-Tucker (KKT) conditions for the dual can be expressed as

$$g_i = \frac{\partial W}{\partial \alpha_i} = \sum_j Q_{ij} \alpha_j + y_i b - 1 = y_i f(x_i) - 1$$

and

$$\frac{\partial W}{\partial b} = \sum_j y_j \alpha_j = 0$$

This partitions the training set into S the Support Vector set ($0 < \alpha_i < C, g_i = 0$), E the error set ($\alpha_i = C, g_i < 0$) and R the well classified set ($\alpha_i = 0, g_i > 0$) [2]. If the points in error are penalized quadratically with a penalty factor C , then, it has been shown that the problem reduces to that of a separable case with $C = \infty$ [3]. The kernel function is modified

$$K'(x_i, x_j) = K(x_i, x_j) + \frac{1}{C} \delta_{ij}$$

where $\delta_{ij} = 1$ if $i = j$ and $\delta_{ij} = 0$ otherwise. The advantage of this formulation is that the SVM problem reduces to that of a linearly separable case. It can be seen that training the SVM involves solving a quadratic optimization problem which requires the use of optimization routines from numerical libraries. This step is computationally intensive, can be subject to stability problems and is non-trivial to implement.

2. CNN (Convolution Neural Network):

Convolutional Neural Networks are based on the Neural Networks, they have neurons that are learnable weights and biases. Each neuron gets some inputs from the input layer and a dot product is performed and then non-linearity is applied. They also have a loss function and all the techniques we used for Neural Networks are also applied here. ConvNet allows us to encode some properties into the architecture. This makes the forward function more efficient and reduces the parameters. Convolutional Neural Network consists of three types of layers, a convolutional layer, pooling layer, and fully-connected layers.

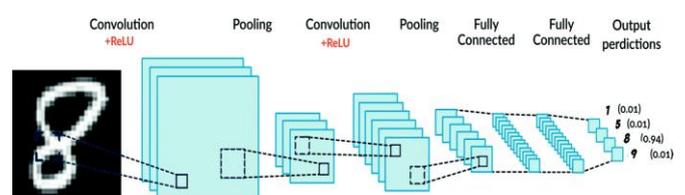


Figure-3: CNN Layers

The convolutional layer learns features from the given image. It is composed of several convolution kernels which are used to compute feature maps. Each neuron of a feature map is connected to neighbouring neurons in the previous layer. This is called the receptive field of a single unit/neuron. A new feature map can be obtained by convolving the given input with a kernel and then applying element-wise nonlinear activation function on the output of the kernel.

Operation	Formula
Convolution	$\mathbf{z}^l = \mathbf{h}^{l-1} * W^l$
MaxPooling	$h_{xy}^l = \max_{i=0,\dots,s,j=0,\dots,s} \mathbf{h}_{(x+i)(y+j)}^{l-1}$
Fully-connected layer	$\mathbf{z}^l = W^l \mathbf{h}^{l-1}$
ReLU	$\text{ReLU}(z_i) = \max(0, z_i)$
Sigmoid	$\sigma(z_i) = \frac{1}{1+e^{-z_i}}$
Softmax	$\text{softmax}(z_i) = \frac{e^{z_i}}{\sum_j e^{z_j}}$
Batch Normalization	$\text{BN}(z_i) = \gamma_i z_i + \beta_i,$ $\hat{z}_i = \frac{z_i - E[z_i]}{\sqrt{\text{Var}[z_i]}}$

\mathbf{z}^l : pre-activation output of layer l
 \mathbf{h}^l : activation of layer l
 $*$: discrete convolution operator
 W, γ, β : learnable parameters

Figure-4: CNN Formula

To generate a feature map the kernel is used by all spatial locations of the input given to the input layer. The final feature maps are obtained by different kernels in the CNN.

5. CONCLUSION

In this project, we will analyse the emotion distribution in a children’s story corpus and have demonstrated that the emotion of a sentence in children’s story is affected by its context. An HMM based method has been proposed to model the emotion sequence in an article. Important features have been listed by comparing the emotional and neutral sentences. Experiments demonstrated that the proposed model is affected less by the training sets than traditional methods. Classifier fusion has been applied to achieve better classification results.

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BIOGRAPHIES



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