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Texture Feature Abstraction based on Assessment of HOG and 1 GLDM Features for Diagnosing Brain Abnormalities in MRI 2 Images 3 Sudheesh K V^1 and L.Basavaraj² 4 ¹ Visveswaraya Technological University 5 Received: 9 December 2017 Accepted: 3 January 2018 Published: 15 January 2018 6

Abstract 8

Recognition of vehicles has always been a desired technology for curbing the crimes done with 9 the help of vehicles. Number imprinted on plates of cars and motorbikes are consist of 10 numerals and alphabets, and these plates can be easily recognized. The uniqueness of 11 combination of characters and numbers can be easily utilized for multiple purposes. For 12 instance, fines can be imposed on people automatically for wrong parking, toll fee can be 13 automatically collected just by recognizing the number plate, apart from these two there may 14 be several numbers of uses can be accommodated. Computer vision is comprehended as a sub 15 space of the computerized reasoning furthermore software engineering fields. Alternate ranges 16 most firmly identified with computer vision are picture handling, picture examination and 17 machine vision. As an exploratory order, computer vision is apprehensive with the counterfeit 18 frameworks that concentrate data from pictures and recordings. The picture information can 19 take numerous structures, for instance, segmentations of videos, taken from several cameras. 20 This thesis presents a training based approach for the recognition of vehicle number plate. 21 The whole process has been divided into three stages i.e. capturing the image, plate 22 localization and recognition of digits over the plate. The characteristics of HOG have been 23 utilized for training and SVM has been used for adopted for classifying while recognizing. 24 This algorithm has been checked for more than 100 pictures. 25

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Index terms— histogram of gradient, gray level difference method, feature extraction. 27

Texture Feature Abstraction based on Assessment of HOG and GLDMF eature and GLDMF eature1 28

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Vidyavrdhaka College of Engineering Introduction maging technique in biomedical field has helped the doctors in 32 depicting the inner parts of the body for easier diagnosis. It also has helped doctors to make keyhole surgeries for 33 reaching the particular interior parts without really opening too much of the body. X-ray is a beam produced by a tube which contains X-ray by sending through the body. This had helped doctors in investigating the subject 34 very detailed manner. But the drawback from this was when the subject was x-rayed many times there was some 35 side-effects. CT scanner, Ultrasound and Magnetic Resonance Imaging have overcome x-ray imaging by making 36 easier way for doctors. Technique of imaging using ultra-sound utilizes high frequency band sound waves which 37 will be in terms of megahertz range that are reflected in imaging. The technique is usually diagnosis of fetus in 38 pregnant women. It can be used to image the breast, muscles, tendons, abdominal organs, heart, arteries and 39

veins. But it provides less information when compared to techniques such as CT or MRI. Powerful magnetic fields 40 and radio waves which produce detailed interior information of body are used in Magnetic resonance imaging 41 technique. MRI is one of the improvised technique in medical imaging which has proven to be a successful tool 42 43 in the realizing the human brain. The abundant information provided by this concerning of the soft tissue and anatomy of brain has improved the quality of diagnosis and its treatment. utilized for examining any part of 44 the body including the injuries due to sports, problems related to musculoskeletal, vessels of heart and blood 45 can be analyzed. Internal organs, brain and spinal cord and bones and joints. Results obtained can be provide 46 ample amount of information for treatment, planning for further and shows how effective the previous treatment 47 was. MRI scan include a special methods that ensures extra information to physician. An intracranial neoplasm 48 (brain tumor) happens when undesirable cells form within the brain. There are mainly malignant or cancerous 49 tumors and benign tumors or noncancerous tumors. Malignant tumors are the tumors which are caused due 50 to inflammation in other parts of the body. In malignant, primary tumors which start within the brain and 51 secondary tumors which will be spreading from other body parts known as brain metastasis tumors. Benign 52 tumors are tumor cells which grow in brain. They grow very slow where as malignant cells grow very rapidly. 53 Brain tumors has the symptoms determined on the amount of the brain involved. Some of the symptoms are 54 55 headaches, problem with vision, seizures, vomiting and mental changes. Cause of most brain tumors is usually 56 unknown. The common type in children is a malignant medulloblastoma. Anticonvulsant medication used for 57 treatment of seizures. Primary brain tumors originate in your brain. They can develop from the brain cells, 58 nerve cells and glands.

⁵⁹ **3 II.**

60 4 Background Study a) Preprocessing

Removing noise from signal or image is the main aim of pre-processing. Photography field uses Preprocessing 61 because an image is degraded which needs to be enhanced so that it will printed as required. Getting back 62 a high quality magnetic resonance image medical for diagnosing is crucial where it injures the subject more if 63 the machines are passing high level Magnetic resonance sound for taking an image. Noise will be introduced 64 65 to image at the time acquisition or during the transmission. The corruption rate in an image is decided by its 66 quantification. Noises sources for an image are: i. Environmental conditions may affect the sensors which are 67 used for imaging. ii. Inadequate light levels and sensor temperature produce the unwanted information in an 68 image. iii. Transmission channel interruption cause noise generation.

⁶⁹ 5 b) Feature Extraction

70 Dimensionality reduction is known as feature extraction. When input is large enough to be processed by an 71 algorithm and it is having an impression to emphasize the quality redundant. This input information will be 72 made to appear into a decreased few features. Features extracted are picked up and expected features set will 73 extract the appropriate information from the input which will perform the desired task being reduced instead of 74 the full size input.

Extracted features from an image are the properties that elaborates the full image. The aim is to reduce the 75 original information set by measuring certain features. Generally feature extraction can be classified as Shape 76 based, Texture based and Intensity based. Circularity, shape, irregularity, area, perimeter etc. are the commonly 77 used shape features. Intensity features are mean, median intensity, variance, standard variance, skewness and 78 kurtosis. Commonly used texture features are contrast, sum of square variance correlation, entropy, energy, 79 homogeneity. Wavelet transform is used in the modern feature extraction method for MRI as the WT supplies 80 localization in both spatial and spectral domains. Wavelet transform can be decomposed into different levels. 81 82 Middle frequencies information is provided by decomposition coefficients. These information has excellent usage for image segmentation. The features extracted from wavelet coefficients are combined and applied as input to 83 the segmentation stage. The work mainly concentrates on extracting the features like histogram of gradient and 84 grey level difference method. III. 85

⁸⁶ 6 Methodology

A system comprises of pre-processing of input image, feature extraction and segmentation of the image. The block diagram representation for various steps carried out in this work is depicted in the Figure 1.

89 7 Input MRI

Image Segmentation Feature Extraction Pre-Processing The input MRI image is acquired by using MRI scan technology. The image acquisition is a difficult task as the image needed from the scanner is of an important organ of the body and once if it is damaged curing those damages is difficult. The user positions his body on the scanner. The user will go through high magnetic resonance imaging so that the machine captures the image in a proper manner. Then collect the images through the help of MRI scanning centers. The scanner produces the images 5mm slice thickness whose resolution varies from 256X256 to 958X958. One of the T2 weighted brain MRI image (normal and abnormal) considered in the database.

⁹⁷ 8 b) Brain MRI Image Preprocessing

The MRI image can be seen as a combination of many noises such as salt and pepper, speckle noise, Rican noise, Gaussian and so on. To improvise the information present in the image and remove associated noises preprocessing techniques are used. In this proposed algorithm we use a hybrid technique known as KSL technique. KSL filtering algorithm is the combination of kernel, sobel and low pass filter. Kernel filter is applied on to the MRI image, where kernel matrix is applied to each pixel in the image. Various kernel filters are used to remove different types of noises. This provides LPF and HPF using a kernel. Next pass the output obtained through sobel filter which does the work of 2-D spatial gradient measurement on an image.

118 The calculation of the gradient is done by:2 2 x y G G G = +

- 119 The closest Magnitude can be estimated as:
- 120 X Y

¹²¹ 9 G G G = +

Non-brain tissue has to be eliminated from the MRI images which is performed by skull stripping methods. 126 Non-brain tissues which are removed is a basic step in the processing of brain MR images. This process by 127 default includes some intensity normalization, spatial normalization, and repositioning of the brain, but to a 128 certain extent those can be switched off. The actual skull stripping is a modified version of the BET algorithm, 129 expanding a spherical surface iteratively until it envelopes the brain. The output can be a skull stripped (masked) 130 131 brain, the mask itself or different surface formats. Skull-Strip is called by a set of other functions. Therefore no separate block for skull-stripping but options can be adjusted as parts of the respective blocks. It increases 132 the speed and accuracy of diagnosis which is the preceding step in various medical applications. It's been know 133 that removes non-cerebral tissues such a skull, scalp, and dura which will from images of brain. S3 is named 134 after algorithm simple skull stripping which depends on brain anatomy and intensity of image characteristics. 135 It uses intensity adaptive thresholding which helps in increased robustness, morphological operations of the 136 magnetic resonance (MR) images. The value of the threshold is evaluated by adaptively knowledge based in 137 brain MR images database. The performance of S3 algorithm is used to estimate the similarities with three 138 popular algorithms known as brain extraction tool, brain surface extractor, and robust brain extraction using 139 140 standard validity indices.

¹⁴¹ 10 c) Brain MRI Image Feature Extraction

Analysis of the texture of an image is the main role in the medical image segmentation. The basic principles is 142 the use of characteristics regarding text, image and background objects. Features are removed in this processes to 143 identify image, text and background objects by the help of various techniques. Few techniques used for excluding 144 the features from the image are depending on shape of the subject, textures and the intensity based. The simple 145 and straight forward statistical features are used to distinguish objects from one another. Principles depends on 146 observing the image pixel colors for example pixels lighter than from background in gray scale images indicate 147 characteristic pattern of the brain tissues. Intensity of the image having high values the WM, pixels of lesser 148 values are GM and pixels with least values are CSF. 149

¹⁵⁰ 11 ? Histogram of Gradients

Object detection in computer vision and image processing are mainly defined through histogram of gradient. The high intensity features are extracted by this method. Dense means that it extracts features for region of interest and all locations in the image as reflected to only the local neighborhood of key-points like SIFT. This technique counts the availability of gradient orientation of an image in localized portions. This is similar to shape Year 2018 ()D © 2018 Global Journals (2) (3) (4) (6) (7) contexts edge orientation histograms. It is uniformly spaced cells of dense grid and this ensures overlapping local contrast for normalization. It will obtains the information of the region by capturing the gradients. It can be done through making the image into small parts. Each cell which are divided has a particular number of gradient orientation bins. Pixels are bi-linearly interpolated to reduce aliasing. Pixel vote for its orientation bin and for the neighboring orientation bins. Distances of pixels from the centre decides the weights. Depending on the energy across the blocks histograms are normalized. Usually each step size has a cell where in each cell has 4 blocks .Therefore these define four different version of histogram.

163 ? Gray Level Difference Method (GLDM)

The GLDM depends on the availability of two pixels which are having difference in grey level. They 164 are separated by specific displacement is ?. In the process of texture analysis usually Haralick features are 165 extracted and from those set, algorithms select the reduced features. The statistical texture features of a digital 166 mammogram or the digital image are extracted using this technique. Contrast can be defined as the difference 167 between the high and low intensity levels of the image. The measure of homogeneity of the image is angular 168 second moment. Entropy is correlated to energy. Measure of inverse difference moment is the distribution of 169 elements in the GLDM in diagonal form. Gray level difference method be defined by g (n, m). For any given 170 displacement The main aim of this network is to create a computational device for looking after the brain to 171 perform many tasks faster than the traditional systems.() i, j? =?? Let ()()() S i, j | S i, j S i i, j j = ? + 172 173 ? + ? Consider difference = 2,

ANN does various tasks as matching of pattern and classification, optimization function, quantization of vector. ANN is an effective information processing system which is similar as biological neural network characteristics. ANN possesses large number of nodes and neurons. These are inter connected to each other. These neurons and nodes have the capability of modeling the networks of the original brain manages it.

Single layered is defined by taking an element to be processed and combining it with other processing element. A layer indicates a stage going stage by stage means input stage and output stage are linked with each other. These lined environments leads to formation of various network architectures. When a layer of the processing nodes is formed the inputs can be connected to these nodes with different weights, resulting in series of outputs one single node. This is called single layered feed forward network. A multilayered is defined by the connection of several layers. The input layer receives the input and has no function as of buffering the input signal. The output layered generates the output of the network called the hidden layers. The hidden layers are the inner layers to

the network. This has no contact with external environment. The layers may vary from zero to N numbers.

¹⁸⁶ 12 IV. Experimental Results

Pre-processing is concerned with predominantly decreasing any variability in the input. The input image for processing is read and shown in Figure 2. The different noise removal algorithms are tested and simulated. Denoising process is carried out for five set of brain MRI images using the different standard impulse lessening filters such as Median mesh (filter), and proposed filtering approach called as Kernel Sobel Low-pass filter. The Table ?? indicates the comparative analysis of performances of different denoising algorithms evaluated using statistical noise parameters such as PSNR, SNR, SSI and MSE.

¹⁹³ 13 Table 2: Comparison Analysis of noise removal techniques

After noise removal the next step is brain image segmentation. But for segmentation of brain image the skull 194 stripping process is a preliminary step. Removal of non-cerebral tissues of the brain is essentially termed as Skull 195 Stripping. The non-cerebral tissues are normally bright and may hamper the segmentation process because MRI 196 images contains patient label (Film Artifacts), noise like salt and pepper noise and skull regions. Hence the direct 197 usage of brain MRI images without skull stripping as a main preprocessing step will affects the accuracy of both 198 segmentation and classification process. Therefore it is necessary to remove the skull region which is not required 199 in the detection of abnormalities. Since morphological operations is said to give better results than tracking 200 algorithms morphological operations based skull stripping algorithm is explored in this work. Related to the 201 methodology we proposes a pivotal scheme which is to create a mask of the brain which is free from noncerebral 202 tissues and multiply with the original image to obtain the resultant image free from skull region. The images 203 obtained after opening operation, creation of mask and after skull removal for brain MRI images are as in Figure 204 5 and the features extracted from the images are HOG and GLDM are as shown in Table 3. 205

²⁰⁶ 14 Conclusion and Future Work

The tumor that is being spread widely have to be identified in prior stages. The tumor which is being captured by MRI scanner in MRI images will provide a wide range of information when compared with the other techniques such as CT scanner, X-ray Scanner. These images are first pre-processed to remove the noise as well as the skull region in the image. Segmentation process is performed to detect the tumor regions. 1 2 3

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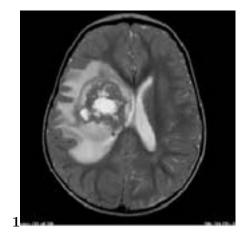


Figure 1: Fig. 1 :

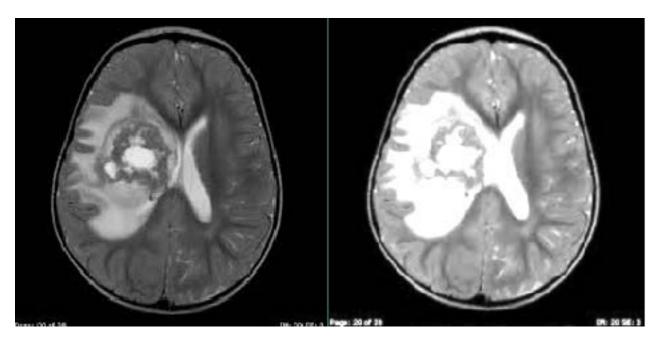


Figure 2:

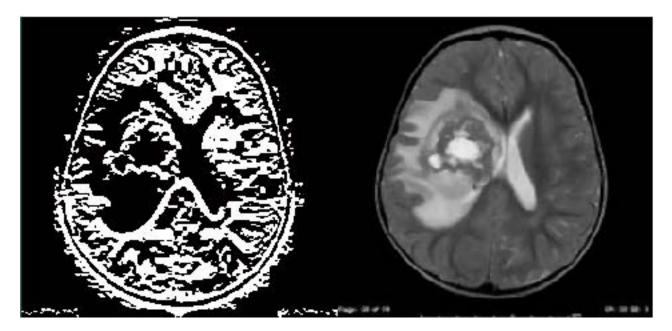


Figure 3:

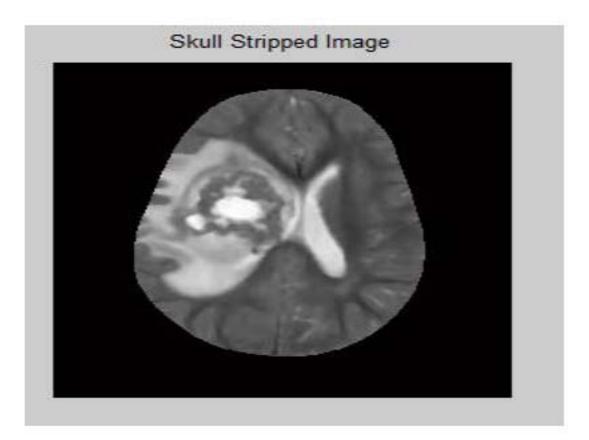


Figure 4:

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| | | | | | $\Delta 9$ |
|--------|-------|-----------|-----------|-----------------------------|------------|
| Test I | mages | Features | Gradients | Method Difference Gray Leve | el () |
| | | Histogram | of | Features from | D |
| Image | 1 | 5720 | | 24979 | |
| Image | 2 | 10248 | | 27571 | |
| Image | 3 | 13949 | | 29589 | |
| Image | 4 | 16692 | | 31292 | |
| Image | 5 | 18974 | | 32827 | |
| Image | 6 | 21025 | | 34345 | |
| Image | 7 | 222839 | | 35539 | |
| Image | 8 | 24525 | | 36760 | |
| Image | 9 | 25910 | | 37906 | |
| Image | 10 | 27201 | | 38935 | |
| V. | | | | | |
| | | | | | |

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Figure 5: Table 3 :

| Test | SNR | SNR | PSNR | PSNR |
|---------|-------|----------|-------|----------|
| Images | (KSL) | (MEDIAN) | (KSL) | (MEDIAN) |
| Image 1 | 26.90 | 8.19 | 36.79 | 28.60 |
| Image 2 | 30.96 | 31.23 | 38.84 | 29.07 |
| Image 3 | 28.51 | 14.63 | 36.36 | 29.31 |
| Image 4 | 22.67 | 16.59 | 33.46 | 30.70 |
| Image 5 | 25.03 | 5.18 | 31.82 | 28.90 |

Figure 6:

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- ²¹¹ [Nag et al. ()] 'A review of image segmentation methods on brain mri for detection of tumor and related ²¹² abnormalities'. S Nag, I Kanta, S K Roy. International Journal of Advanced Research in Computer Science
- and Software engineering 2014. 4 p. .
- [Otsu] 'A Threshold Selection Method from Gray-Level Histogram'. N Otsu . IEEE Trans.on System Man
 Cybernetics 9 p. .
- [Singh and Kaur ()] 'Classification of Abnormalities in Brain MRI Images Using GLCM, PCA and SVM'. D
 Singh , K Kaur . International Journal of Engineering and Advanced Technology 2012. 1 (6) p. .
- [El-Dahshan et al. ()] 'Computeraided diagnosis of human brain tumor through mri: A survey and a new algorithm'. E.-S A El-Dahshan , H M Mohsen , K Revett , A.-B M Salem . *Expert systems with Applications* 2014. 41 (11) p. .
- [Gopinathan and Poornima ()] 'Enhancement of images with speckle noise reduction using different filters'. S
 Gopinathan , S Poornima . International Journal of Applied Science and Engineering Research 2015. 4 (3) p.
 .
- [Mokji and Bakar ()] 'Gray Level Co Occurrence Matrix Computation Based on Haar Wavelet'. M M Mokji , S
 A R Bakar . Computer Graphics, Imaging and Visualization, 2007.
- [Mohanaiah et al. ()] 'Image Texture Feature Extraction using GLCM Approach'. P Mohanaiah , P Satya narayana , L Gurukumar . International Journal of Scientific and Research Publication 2013. 3 (5) p. .
- [Chinnu ()] 'Mri brain tumor classification using svm and histogram based image segmentation'. A Chinnu .
 International Journal of Computer Science and Information Technologies 2015. 6 (2) p. .
- [Patil et al. ()] 'Preprocessing to be Considered for MR and CT Images Containing Tumors'. Sonali Patil , V R
 Dr , Udupi . IOSR Journal of Electrical and Electronics Engineering 2012. 1 (4) p. .
- [Bandhyopadhyay and Paul ()] 'Segmentation of brain mri image-a review'. D S K Bandhyopadhyay , T U Paul
 International Journal of Advanced Research in Computer Science and Software Engineering 2012. 2 (3) .
- [Albregtsen ()] Statistical Texture Measures Computed From Gray Level Co Occurrence Matrices, Fritz Albregtsen
 2008. p. . Image Processing Laboratory, Department of Informatics, University of Oslo
- [Rajeswari and Jeyaselvi ()] 'Support Vector Machine Classification for MRI Images'. Theiva S Rajeswari ,
 Jeyaselvi . International Journal of Electronics and Computer Science Engineering 1956. 1 (3) p. .
- [Kumari ()] 'SVM Classification An Approach on Detecting Abnormality in Brain MRI Images'. Rosy Kumari .
 International Journal of Engineering Research and Applications 2013. 3 (4) p. .
- [Harclick et al. ()] 'Texture Features For Image Classification'. R Harclick , K Shanmugam , I Dinstein . IEEE
 Transaction 1973. 3 (6) p. .