

## **An Efficient Technique for Tumor Detection and Classification Using K-Means Clustering Algorithm**

**Sangeeta<sup>1</sup>, Dr.Nagendra.H<sup>2</sup>**

Electronics and communication engineering, PDA College of engineering,  
Kalaburagi, Karnataka, India  
Associate professor, Electronics and communication engineering, PDA College of  
engineering, Kalaburagi, Karnataka, India

(\* Corresponding author's e-mail: [sangeeta2246@gmail.com](mailto:sangeeta2246@gmail.com))

### **Abstract**

The unusual cells structure in cerebrum is brain tumor. There are two sorts those are benign and malignant tumor. Identification and characterization of tumor assume a significant part lately. Division of cerebrum Tumor is a one of the fundamental advance in discovery as well as order of tumor. The way toward apportioning a picture keen on a few districts is alluding to picture division. Cerebrum pictures are dissimilar fit as a fiddle; contain commotion, complex in surface, inadequately tested, varieties in picture quality, organic changeability as well as dissimilar elements make a hard errand for mind tumor division. There are algorithms and technique accessible for picture division yet there necessities to construct up a proficient strategy of clinical picture division. In malignant growth determination the clinical imaging strategies assume a significant part these days. The quite possibly the main procedure is Magnetic resonance imaging (MRI) method which is utilized to distinguish as well as find the tumor in the cerebrum. This paper present productive picture divisions approach utilizing K-implies bunching. It is trailed via sifting, thresholding, otsu binarization as well as segmentation stages to give a precise mind tumor discovery. The Median channel is sifting strategy to eliminate the commotion in MRI picture. The proposed procedure can get advantages of K method grouping for picture segmentation through least handling instance. The proposed strategy is approved on BRATS 2015 dataset.

**Keywords:** Median filter, otsu binarization, K means cluster, MRI images, KSVM

### **Introduction**

The improvement of strange cells in the cerebrum is mind tumor. There are two fundamental sorts of tumors are Benign as well as malignant. The essential malignant tumor is begin in actual mind as well as Secondary cerebrum tumors spread as of somewhere else. The auxiliary cerebrum tumor is likewise called metastasis tumors. There are four evaluations of cerebrum tumors as indicated via the American Brain Tumor Association plus World Health Organization [9]. Kind tumors, which be moderate rising be Grade I as well as grade II, and are otherwise called poor quality tumors. Threatening tumors, which be fast developing are grade III as well as grade IV and are called high-grade tumors. It is imperative to distinguish mind tumor at beginning phase and it is significant to recognize the tumor territory plus segment the tumor pictures.

Computed Tomography (CT), Single-Photon Emission Computed Tomography (SPECT), Positron Emission Tomography (PET), Magnetic Resonance Spectroscopy (MRS) as well as Magnetic Resonance Imaging (MRI) are medical Imaging method. These method give significant data about shape, dimension, area plus digestion of cerebrum tumors aiding finding. X-ray is

considered as the standard procedure because of its enormous delicate tissue contrast goal plus, picture control, utilize non-ionizing radiation moreover multi plane imaging technique.[11]

During picture securing, coding as well as transmission the clamor is consistently present in advanced picture. Thus, in picture preparing channel strategy are for most part used to stifle either the elevated frequencies in picture implies smoothing the picture or low frequencies in picture implies recognizing the edges in picture. There are assorted kind of picture clamor separating strategy in picture handling those are Median channel, mean channel, reciprocal channel, weiner channel, Anisotropic channel as well as Gaussian channel. On the off chance to salt plus pepper commotion is available in picture, at to tip meadian channel is acceptable strategy, poisson clamor is available in picture at to tip mean channel is acceptable strategy as well as spot commotion is loathe in picture then weiner filter method is good. [12]

Picture segmentation is parceling a mind picture keen on numerous portions which be uniform as well as homogenous as for certain attribute, for instance, shading, power or surface. Segmentation procedures be edge strategy, edge based segmentation, area based segmentation, grouping based segmentation, watershed based segmentation as well as fake neural organization based segmentation.

Bunching is gathering comparative information as well as diminishes information measurement plus it learns the state of dataset via over as well as over drawing its neurons nearer to information focuses. Grouping technique be k-mean bunching, improved k-mean bunching, Fuzzy C-mean as well as improved Fuzzy C-mean bunching. In K-implies, information will be remembered for one specific bunch. In FCM, information can be remembered for all current bunches, however through shifting level of participation in a scope of qualities [0 1]. the execution instance is less in K-Means contrasted through Fuzzy C Means grouping approach, in light of fact to quantity of emphases of Means is fewer than Fuzzy C Means bunching .[1]

## Literature survey

**Nimeesha et al.(2013)** They planned K means as well as FCM clustering for segmentation of brain tumor. They composed 100 MRI brain picture as of publically obtainable dataset as well as they achieve 95% of accuracy via using K-means clustering as well as 84.3% of accuracy via using Fuzzy C Means.

**Bangare et al.(2015)** They planned K means as well as FCM for segmentation of brain tumor plus they used Magnetic Resonance Imaging (MRI) furthermore CT-Scanned picture. They gave the output of K-Means algorithm is used as input to this Fuzzy C-Means algorithm as well as they did area computation.

**Praveen, G et al. (2015)** planned Least Squares Support Vector Machine classifier through Multilayer perceptron bit for arrangement of cerebrum tumor as well as rapid bouncing box is utilize for division of tumor part as of mind. They gathered mix of dataset a) Real instance MRI check got at Goa Medical College, Goa plus b) Synthetic picture which were open as a piece of Brats dispute, MICCAI 2013. They utilized 100 picture through 75 are tumor MR picture plus 25 are typical MR picture as well as They accomplished 95.6% characterization exactness.

**Dubey et al. (2016)** They planned *c*-implies (FCM) grouping calculation for division of cerebrum MR images. They gathered information as of The Internet Brain Segmentation Repository (IBSR)

is a WorldWideWeb asset bountiful admittance to attractive reverberation mind picture information as well as they contrast distinctive bunching strategies plus FCM.

**Bhima, K et al.(2016)** They planned Watershed as well as Marker based Watershed Algorithm for division of mind tumor. They gathered three distinctive Datasets of Brain MRI picture plus they accomplished division precision 98.02% via considering Datasets-1, 96.81% via considering Datasets-2 plus 97.21% via considering Datasets-3.

**Sudhakar et al.(2017)** They planned Curvelet Decomposition as well as Grey Level Co-occurrence Matrix (GLCM) for spatial Feature Extraction as well as Probabilistic Neural Network (PNN) enemy order of cerebrum tumor. They gather 50 MRI pictures 40 be tumor plus 10 ordinary mind picture as of WHO (World Health Organization), WBA (World Brain Atlas) Website. They recognize the tumor via utilizing PNN classifier.

**Gupta et al.(2017)** They planned fluffy c-implies for tumor division as well as support vector machine (SVM) for tumor classification plus gray level run length matrix (GLRLM) for comprise extraction. They gather 100 MRI pictures of 120 patients as well as they accomplished 91.77% of accuracy via utilizing Linear Kernel capacity plus 90.01% of precision via utilizing RBF Kernel work.

**Patil et al.(2017)** They planned Fuzzy Transform as well as Morphological activity for division plus recognition of cerebrum tumor. They gathered 20 mind MRI pictures as of WHO 2007 moreover they recognize the tumor part.

**Ilhan et al.(2017)** They planned edge approach for division plus Morphological activities to distinguish the limit in a picture. They gather 100 MRI pictures of 70 be anomalous as well as 30 are typical as of TCIA (The Cancer Imaging Archive, 2017) moreover they accomplished Sensitivity (94.28%), Specificity (100%) as well as Accuracy (96%).

**Rashid et al. (2018)** They planned Anisotropic unraveling tactic to eliminate the commotion as well as for division they utilize SVM classifier plus to recognize the influenced zone as of MRI pictures of mind tumor they utilized morphological tasks. They gathered solitary unusual MRI picture as of The National Center for Biotechnology Information as well as they accomplished 83% precision in division.

**Vidhyavathi et al.(2018)** planned Adaptive Neuro-Fuzzy Interface System (ANFIS) classifier for characterization of cerebrum tumor. They gather information as of clinical school site; it contain 85 pictures; amongst this 38 pictures be typical as well as remaining pictures be unusual as well as they proficient accuracy as 96.23%, affectability as 92.3%, plus particularity as 94.52%.

**Deepa et al., (2019)** planned method which comprise of highlight extraction, preprocessing, combination to achieve elevated precision in order, as well as choice. The normal channel is utilized to lessen diversity in force of pictures in preprocessing step. The direction, region, as well as recurrence are alienated via Gabor wavelet include extraction which give surface statistics to arrange. The little subset of highlights is chosen via kernel principal component analysis (KPCA) to progress the implication as well as to smother the repetition of element. The GRBF of highlight combination give the statistics recognized as of highlights having numerous sets. To order an intertwined highlight versatile firefly back engendering neural organization is utilized. They gathered picture as of the BRATS 2015 dataset counting 81 MRI picture, through 11 ordinary pictures, 55 pictures having hazardous tumor as well as 15 pictures having amiable tumor. They

accomplished 99.85% of explicitness, 99.84% of Accuracy also 97.24% of affectability.

**Rajesh T et al., (2019)** The planned approach incorporate order of tumor as well as extraction of highlights. For extraction of highlights, they utilized (RST) Rough set theory as well as the characterization of MRI cerebrum pictures as anomalous plus ordinary particle swarm optimization neural network (PSO) is tried. They gathered 90 pictures, 60 pictures are use for testing as well as 30 for preparing moreover they consummate 95% of precision when contrast through RST-FFNN (Feed-forward neural Network) plus RST-FSVM.

**Sajid et al.(2019)** They planned convolutional neural network (CNN) for division plus this work comprise of preprocessing venture, in which pictures be standardized as well as predisposition field adjusted, a feed-forward pass through a CNN plus a post-handling step, which is utilized to eliminate little bogus positives around skull partition. They gathered pictures as of BRATS 2013 dataset as well as they accomplished 86% of Sensitivity moreover 91% of Specificity.

**Alam et al. (2019)** They planned format based K method for division of mind tumor as well as enhanced fluffy C methods (TKFCM) computation for identify human cerebrum tumors in an magnetic resonance imaging (MRI) picture. They gathered 40 MRI pictures plus They accomplished 97.43% Sensitivity, 100% of Specificity, 97.5% of exactness.

**Ahmed et al.(2019)** planned edge division to fragment the picture plus mix of GrayWolf Optimizer (GWO) as well as Support Vector Machine (SVM) through Radial Basis Function (RBF) piece to cluster the cerebrum tumor. They gather 80 MR cerebrum pictures at which 20 be kindhearted tumors as well as 60 are threatening tumors as of Harvard Medical School Website as well as The second dataset is BRATS 2015 dataset which comprise of 60 High Grade Glioma (HGG) pictures plus 20 Low Grade Glioma (LGG) picture .They accomplished 98.75% characterization precision.

**Chinnam et al.(2019)** planned Gabor wavelet alter for Feature extraction, Principle Component Analysis (PCA) for Feature reduce as well as backing vector machine-Pearson VII widespread bit (SVM-PUK) is utilized for grouping. They gathered information as of benchmark dataset BRATS 2018 plus they accomplished 91.31 % of accuracy.

**Rajan et al.(2019)** They planned K- K-means clustering, integrated through Fuzzy C-Means (KMFCM) for division as well as order of cerebrum tumor. They 41 MRI cuts of patients' picture gathered as of ANBU clinics in Madurai. They accomplished 99.25% of arrangement precision as well as they reduces the execution instance moreover.

**Akram et al.(2019)** They planned coordinated effort of BMM(Beta mixture model)plus LA(Learning automata) computation for cerebrum tumor division as well as for order of mind tumor is ended via utilizing support vector machine (SVM), K-nearest neighbor (KNN) plus decision tree (DT).They gather 79 MR pictures as of TCIA dataset as well as Harvard Medical School plus they talented 98.74% of arrangement accuracy via utilizing SVM classifier through direct piece.

**Choudhury et al.(2020)**They planned convolutional neural network (CNN) for highlights extraction as well as 3 layer CNN replica to order the tumor plus they accomplished 97.47% of characterization exactness via utilizing preparing information.

**Bhandari et al.(2020)** They planned convolutional neural network (CNNs) to part the cerebrum tumor. They gathered preparing dataset as of Multimodal Brain Tumor Segmentation (BraTS) benchmark as well as commencing their own college. They accomplished division precision as of 82 to 89%.

**Abd Alreda et al. (2020)** They planned Feed Forward Back Propagation neural network (FFBPNN) as well as support vector machine (SVM) for cerebrum tumors order plus Discrete wavelet alter moreover gray level co- occurrence matrix (GLCM) for include extraction. They gather 308 pictures of 83 cases for prepare as well as 37 cases for testing from the cancer imaging archive (TCIA) as well as 59 cases for preparing plus 29 cases for test as of Harvard clinical cerebrum information bases also they gather 66 cases for preparing as well as 34 cases for testing as of General Hospital of Baquba-Iraq of MRI mind pictures as well as they proficient 97% of characterization precision.

**Chander et al., (2020)** They suggested to MRI picture can be alienated keen on dissimilar portions via versatile k-implies bunching as well as via utilizing Support Vector Machine classifier alienated pictures be ordered. They gathered forty MR pictures of harmful as well as generous tumors as of Harvard University clinical Image Repository. They accomplished 93% of accuracy utilizing straight piece strategy plus division precision is 99.7%.

**Deb et al.(2020)** planned two-sided sifting policy to eliminate the commotion ,GLCM is utilize for include extraction , Adaptive Squirrel search algorithm (ASSA) is use for segmentation as well as AFDNN (Adaptive Fuzzy Deep Neural Network) is utilize for grouping of mind tumor. They gather information as of BRATS 2012, BRATS 2015 as well as BRATS 2016 plus they accomplished 99.6% arrangement precision via utilize BRATS 2016 dataset.

**Rai et al.(2020)** planned Low Layered U-Net (LU-Net), Le-Net as well as VGG-16 CCN replica for recognition as well as division of cerebrum tumor. They gather of 253 pictures of elevated pixels as of freely accessible open-source dataset moreover they accomplished 98.00% of accuracy.

**Toğaçar et al.(2020)** planned Brain MRNet is novel convolutional neural network replica for recognition of mind tumor. They gather 253 pictures in which 155 pictures are strange plus 98 pictures are usual. Here pictures are in JPEG organization as well as those accomplished 96.05% of accuracy.

**Kalaiselvi et al.(2020)** planned Convolutional Neural Networks (CNN) for order. They gather dataset as of BRATS 2013 dataset as well as a clinical datasets gathered as of The Whole Brain Atlas (WBA), via Women's Hospital, Harvard Medical School, Boston, USA. They utilize 6 model of CNN plus they accomplished 96-99% of characterization accuracy.

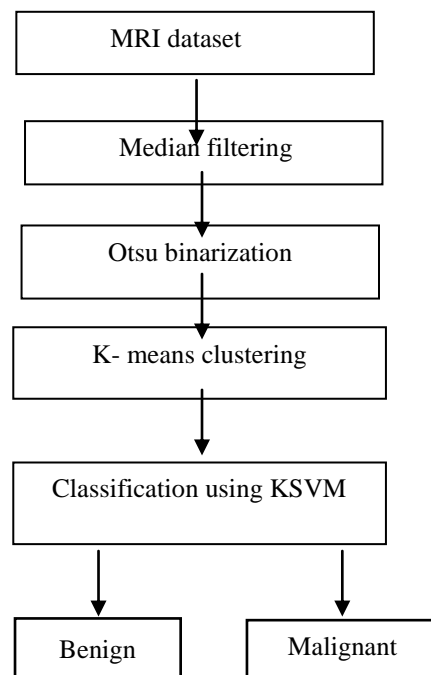
**Kumar et al. (2020)** planned versatile k-closest neighbor classifier to arrange the tumor as typical otherwise unusual as well as for division they planned the ideal possibilistic fluffy C-implies grouping computation. They gather 1000 pictures as of BRATS MICCAI cerebrum tumor dataset as well as another is publically accessible dataset picture which be gather as of web. They accomplished 96.5% arrangement accuracy,100% Sensitivity, 93%% of Specificity.

**Amarapur et al.(2020)** They planned Cognition based Modified Level Set Segmentation method for division of cerebrum MRI picture as well as for order of mind tumor is ended via utilizing Adaptive Artificial Neural Network (AANN) arrangement policy .They gathered information as

of BRATS-2015 data set as well as they accomplished 98% of grouping precision .

**SivaSai et al. (2021)** planned Adaptive Bilateral Filter to eliminate the commotion plus Fuzzy Recurrent Neural Network (FR-Net) is utilized for division of cerebrum tumor. They gather 1900 MRI pictures from Kaggle plus they accomplished 87.8% of exactness.

## Materials and methods



**Figure 1: Brain tumor segmentation and detection methodology**

**Dataset:** Gathered 100 multi-methodology MRI picture as of BRATS 2015 benchmark. X-ray scanned pictures be either shading, Gray-scale or force picture through size of 220×220.if it is Gray-scale picture , a Gray-scale distorted over picture is characterize via utilizing a huge grid whose passages be mathematical qualities somewhere in range of 0 and 255, where 0 relates to dark as well as 255 compare to white . Picture division as well as edge location is considered as two main stage for mind tumor recognition.

**Pre-processing stage:** It is essential stage which is utilized to eliminate the undesirable commotion as of pictures like patients name, age, address as well as other additional subtleties be taken out during this interaction. The picture is resized as well as alter of RGB to grayscale is perform. Because of warm impact, there might be commotion through MRI scrutiny, as well as it is significant to kill this clamor. The Median channel is nonlinear channel which is utilize to eliminate clamor lacking influencing picture data as well as it give elevated goal outcome contrasted through dissimilar channels, for instance, spatial channel, mean channel, two-sided channel, weiner channel, Anisotropic channel as well as Gaussian channel plus versatile channels. The dim scale picture comprise of Salt as well as Pepper commotion, this clamor is eliminate via utilizing middle channel.

**Image Segmentation:** It is perform via two stage those be Otsu binarization as well as K -

implies bunching. Choosing legitimate division strategy is troublesome assignment as a outcome of incredible verities of the injury shape, size, as well as tones alongside assorted skin type as well as surfaces. The downside of FCM bunching for picture division is to its aim work doesn't contemplate any spatial reliance amongst pixels of picture yet manage pictures equivalent to isolate focuses. Second drawback of FCM grouping approach is to participation work is usually settled via  $(xk, Vi)$ , which gauge the closeness among the pixel force as well as the bunch place. Higher enrollment relies upon nearer force esteem to cluster place. Subsequently participation work is exceptionally delicate to commotion [18]. They can be extensively delegated thresholding, edge-based otherwise district based, regulated as well as unaided characterization method

- ☐ Threshold segmentation
- ☐ Water shed segmentation
- ☐ Gradient Vector Flow (GVF)
- ☐ K-mean Clustering
- ☐ Fuzzy C-means Clustering

### 1) Otsu method:

Otsu binarization was utilize to alter over the picture keen on its parallel arrangement as well as it naturally find the binarization edge. It is frequently utilize thresholding policy. It is utilized to naturally perform bunching based picture thresholding, or, diminish of a grey level picture to a parallel picture in PC vision as well as picture prepare. Otsu's thresholding process depends on a discriminate investigation which parcels the picture keen on classes  $C_1$  plus  $C_2$  at grey levels 'k' through end goal to  $C_1 = \{0, 1, 2, \dots \dots k\}$  plus  $C_2 = \{k+1, k+2, \dots \dots L-1\}$  where, 'L' is whole numeral of grey levels of picture. Let 'n' be all out numeral of pixels in specified picture as well as 'ni' be quantity of pixels at  $i^{th}$  grey level. The likelihood of event of dim level is characterized as,

$$P_i = \frac{n_i}{n}$$

$C_1$  as well as  $C_2$  be two classes representing the region of interest plus background.

### 2) K-means clustering

K-means is an unsupervised iterative cluster method as well as it split the specified statistics keen on K predefined distinct clusters. The cluster is defined as a collection of statistics points exhibiting certain similarity. The objective purpose telling k-means clustering can be written as:

$$S = \sum_{b=1}^k \sum_{a=1}^x ||x_a^{(b)} - c_b||^2$$

Where  $||x_a^{(b)} - c_b||^2$  is the predefined distance as of statistics point  $x_a^{(b)}$  as well as cluster center  $c_b$ . S is the pointer of distance of N point as of their respective cluster center.

Partition the data set via using these following points:

- 1) Each information point has a place through a group among the closest mean.
- 2) Data focuses having a place through one group encompass serious level of likeness.
- 3) Data focuses having a place through assorted bunches have serious level of uniqueness.

The algorithm steps of k-means clustering can be articulated as:

Step 1:

- 1) Choose the quantity of group 'K'
- 2) Arbitrarily select any 'K' information

Stage 2:

- 1) Select group habitats so to they are just about as farther as conceivable as of one another.

Stage 3:

- 1) Compute the distance among every information point as well as each bunch community.
- 2) The distance is determined either specified distance work or via utilizing Euclidean distance recipe.

Stage 4:

- 1) Assign each information highlight some group.
- 2) A information tip is relegating to group where focus is closest to information point.

Stage 5:

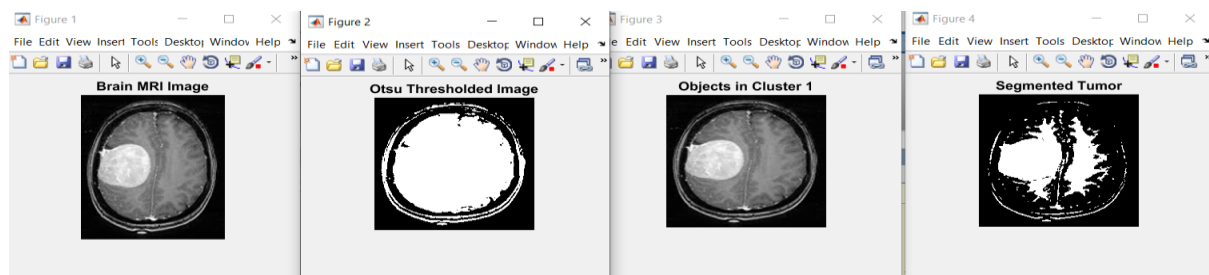
- 1) Recomputed the focal point of recently framed bunches.
- 2) The focus of a group is registered via taking mean of all information focuses contained in to bunch.

Stage 6:

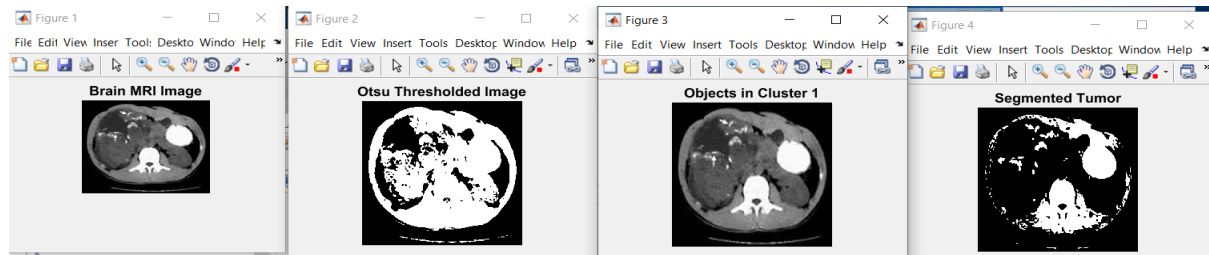
- 1) Keep rehashing the technique as of Step 3 to Step 5 until any of accompanying stopping model is met.
- 2) Center of recently shaped group don't change
- 3) Data focuses stay present same group.
- 4) Maximum quantities of cycles are reached.

## Results and discussion

This section presents trial aftereffects of segmentation of cerebrum tumor utilizing K-implies grouping to recognize the contaminated district. We encompass gather 100 pictures of both ordinary as well as anomalous cerebrum tumor pictures as of BRATS 2015 dataset as well as applied K-implies bunching computation to distinguish the contaminated district. Underneath show 10 pictures consequence of preprocessing plus segmentation for ordinary picture, picture through kindhearted tumor moreover picture through malignant tumor.







**Figure 2: Brain MRI image with segmented region**



**Figure 3: Classification of tumor by using K-means clustering**

## Conclusion

In this world a brain tumor is one of main infections. Thus, it is vital to distinguish as well as classify mind tumor at beginning phase. X-ray mind pictures be broadly used to classify cerebrum tumors. Mind tumor segmentation utilizing otsu binarization as well as K-implies bunching. In this exploration work K-Means grouping is utilize to fragment as well as recognize the phase of cerebrum tumor. Likewise these computation encompass beaten the downsides of thresholding computation as well as it is usually proficient through instance intricacy. The principle bit of leeway of proposed strategy is decrease in execution instance. In the following composition the arrangement utilizing a kernel support vector machine (KSVM) to accomplish a decent order exactness will be specified.

## References

- [1] Nimeesha, K. M., and Rajaram M. Gowda. "Brain tumour segmentation using Kmeans and fuzzy c-means clustering algorithm." *Int J Comput Sci Inf Technol Res Excell* 3 (2013): 60-65. DOI: 10.13140/RG.2.1.3961.5841
- [2] Menze, Bjoern H., Andras Jakab, Stefan Bauer, Jayashree Kalpathy-Cramer, Keyvan Farahani, Justin Kirby, Yuliya Burren et al. "The multimodal brain tumor image segmentation benchmark (BRATS)." *IEEE transactions on medical imaging* 34, no. 10 (2014): 1993-2024.
- [3] Bangare, Sunil L., Madhura Patil, Pallavi S. Bangare, and S. T. Patil. "Implementing tumor detection and area calculation in MRI image of human brain using image processing techniques." *Int. Journal of Engineering Research and Applications* 5, no. 4 (2015): 60-65.
- [4] Praveen, G. B., and Anita Agrawal. "Hybrid approach for brain tumor detection and classification in magnetic resonance images." In *2015 Communication, Control and Intelligent Systems (CCIS)*, pp. 162-166. IEEE, 2015.

- [5] Dubey, Yogita K., and Milind M. Mushrif. "FCM clustering algorithms for segmentation of brain MR images." *Advances in Fuzzy Systems* 2016 (2016).
- [6] Bhima, K., and A. Jagan. "Analysis of MRI based brain tumor identification using segmentation technique." In *2016 International Conference on Communication and Signal Processing (ICCSP)*, pp. 2109-2113. IEEE, 2016.
- [7] Işın, Ali, Cem Direkoğlu, and Melike Şah. "Review of MRI-based brain tumor image segmentation using deep learning methods." *Procedia Computer Science* 102 (2016): 317-324. <https://doi.org/10.1016/j.procs.2016.09.407>
- [8] Sudhakar, T., J. Bethanney Janney, D. Haritha, M. Juliet Sahaya, and V. Parvathy. "Automatic Detection and Classification of Brain Tumor using Image Processing Techniques." *Research Journal of Pharmacy and Technology* 10, no. 11 (2017): 3692-3696.DOI: 10.5958/0974-360X.2017.00669.2
- [9] Gupta, Gaurav, and Vinay Singh. "Brain Tumor segmentation and classification using Fcm and support vector machine." *IRJET* 4, no. 5 (2017): 792-796.
- [10] Patil, Ms Priya, Ms Seema Pawar, Ms Sunayna Patil, and Arjun Nichal. "A Review Paper on Brain Tumor Segmentation and Detection." *IJIREEICE* 5 (2017): 12-15. DOI: 10.17148/IJIREEICE.2017.5103
- [11] Ilhan, Umit, and Ahmet Ilhan. "Brain tumor segmentation based on a new threshold approach." *Procedia computerscience* 120(2017):580-587.
- [12] Rashid, M. H. O., M. A. Mamun, M. A. Hossain, and M. P. Uddin. "Brain Tumor Detection Using Anisotropic Filtering, SVM Classifier and Morphological Operation from MR Images." In *2018 International Conference on Computer, Communication, Chemical, Material and Electronic Engineering (IC4ME2)*, pp. 1-4. IEEE, 2018.
- [13] RM, Vidhyavathi, Mohamed A. Elsoud, and Majid Alkhambashi. "Optimal feature level fusion based ANFIS classifier for brain MRI image classification.", p p 1-12,2018.
- [14] Deepa, A. R., and WR Sam Emmanuel. "An efficient detection of brain tumor using fused feature adaptive firefly backpropagation neural network." *Multimedia Tools and Applications* 78, no. 9, p.p 11799-11814, 2019 – Springer.
- [15] Rajesh, T., R. Suja Mani Malar, and M. R. Geetha. "Brain tumor detection using optimisation classification based on rough set theory." *Cluster Computing* Volume 22, no 6, p.p 13853-13859, 2019 - Springer.
- [16] Sajid, Sidra, Saddam Hussain, and Amna Sarwar. "Brain tumor detection and segmentation in MR images using deep learning." *Arabian Journal for Science and Engineering* 44, no. 11, p.p 9249-9261 2019 springer.
- [17] Alam, Md Shahariar, Md Mahbubur Rahman, Mohammad Amazad Hossain, Md Khairul Islam, Kazi Mowdud Ahmed, Khandaker Takdir Ahmed, Bikash Chandra Singh, and Md Sipon Miah. "Automatic Human Brain Tumor Detection in MRI Image Using Template-Based K Means and Improved Fuzzy C Means Clustering Algorithm." *Big Data and Cognitive Computing* 3, no. 2,p p 27,2019.
- [18] Ahmed, Heba M., Bayumy AB Youssef, Ahmed S. Elkorany, Zeinab F. Elsharkawy, Adel A. Saleeb, and Fathi Abd El-Samie. "Hybridized classification approach for magnetic resonance brain images using gray wolf optimizer and support vector machine." *Multimedia Tools and Applications* 78, no. 19, p p 27983-28002, 2019.

- [19] Chinnam, Siva Koteswara Rao, Venkatramaphanikumar Sistla, and Venkata Krishna Kishore Kolli. "SVM-PUK Kernel Based MRI-brain Tumor Identification Using Texture and Gabor " International information and engineering technology association, vol 36, No 2, p 185-191, 2019.
- [20] Rajan, P. G., and C. Sundar. "Brain tumor detection and segmentation by intensity adjustment." *Journal of medical systems* 43, no. 8 (2019): 282. <https://doi.org/10.1007/s10916-019-1368-4>.
- [21] Edalati-rad, Akram, and Mohammad Mosleh. "Improving Brain Tumor Diagnosis Using MRI Segmentation Based on Collaboration of Beta Mixture Model and Learning Automata." *Arabian Journal for Science and Engineering* 44, no. 4 (2019): 2945-2957. <https://doi.org/10.1007/s13369-018-3320-1>
- [22] Choudhury, Chirodip Lodh, Chandrakanta Mahanty, Raghvendra Kumar, and Brojo Kishore Mishra. "Brain Tumor Detection and Classification Using Convolutional Neural Network and Deep Neural Network." In *2020 International Conference on Computer Science, Engineering and Applications (ICCSEA)*, pp. 1-4. IEEE, 2020.
- [23] Bhandari, A., Koppen, J. & Agzarian, M. Convolutional neural networks for brain tumour segmentation. *Insights Imaging* 11, 77 (2020). <https://doi.org/10.1186/s13244-020-00869-4>
- [24] Abd Alreda, Ban Mohammed, and Thamir Rashed Saeid Hussain Kareem Khalif. "Automated Brain Tumor Detection Based on Feature Extraction from The MRI Brain Image Analysis." *Iraqi Journal For Electrical and Electronics Engineering (IJEED)*, p p 1-9, 2020, DOI: 10.37917/ijeed.16.2.6.
- [25] Chander, P. Sharath, J. Soundarya, and R. Priyadharsini. "Brain Tumour Detection and Classification Using K-Means Clustering and SVM Classifier." In *RITA 2018*, p p. 49-63, 2020 Springer, Singapore.
- [26] Deb, Daizy, and Sudipta Roy. "Brain tumor detection based on hybrid deep neural network in MRI by adaptive squirrel search optimization." *Multimedia Tools and Applications*, p p 1-25, 2020.
- [27] Rai, Hari Mohan, and Kalyan Chatterjee. "Detection of brain abnormality by a novel Lu-Net deep neural CNN model from MR images." *Machine Learning with Applications*, p p 100004, 2020.
- [28] Toğaçar, Mesut, Burhan Ergen, and Zafer Cömert. "BrainMRNet: Brain tumor detection using magnetic resonance images with a novel convolutional neural network model." *Medical Hypotheses* 134, p p 109531, 2020.
- [29] Toğaçar, Mesut, Zafer Cömert, and Burhan Ergen. "Classification of brain MRI using hyper column technique with convolutional neural network and feature selection method." *Expert Systems with Applications* 149, p p 113274, 2020.
- [30] Kalaiselvi, Thiruvankadam, Thiyagarajan Padmapriya, Padmanaban Sriramakrishnan, and Venugopal Priyadharshini. "Development of automatic glioma brain tumor detection system using deep convolutional neural networks." *International Journal of Imaging Systems and Technology* (2020).

- [31] Kumar, D. Maruthi, D. Satyanarayana, and MN Giri Prasad. "MRI brain tumor detection using optimal possibilistic fuzzy C-means clustering algorithm and adaptive k-nearest neighbor classifier." *Journal of Ambient Intelligence and Humanized Computing*, p p 1-14,2020.
- [32] Amarapur, Basavaraj. "Computer-aided diagnosis applied to MRI images of brain tumor using cognition based modified level set and optimized ANN classifier." *Multimedia Tools and Applications* 79, no. 5, 3571-3599,2020.
- [33] SivaSai, Jalluri Gnana, P. Naga Srinivasu, Munjila Naga Sindhuri, Kola Rohitha, and Sreesailam Deepika. "An Automated Segmentation of Brain MR Image Through Fuzzy Recurrent Neural Network." In *Bio-inspired Neurocomputing*, pp. 163-179. Springer, Singapore,2021.

## Author details



Sangeeta Working as assistant professor at Godutai engineering college for women, Klaburagi. She is pursuing part time Ph.D. in Poojya Doddappa Appa College of Engineering, Kalabuaragi, Karnataka, India. Her research area is Biomedical image processing.



Dr. Nagendra H received his Ph.D. in Biomedical Signal Processing (Electrical Engineering Department) from Indian Institute of Technology Roorkee, Roorkee, India, 2016 and M.Tech. in Power Electronics (PDA Engg. College, Kalaburagi) from Vishwesharaya Technological University Belagavi, karnataka, India, 2002. B.E. in Electronics and Communication Engineering (PDA Engg. College, kalabuaragi) from Gulbarga University, kalaburagi, Karnataka, India, 1993. he has published many research papers in National and International Journals / Conferences and also presented her research findings in various National and International conferences. Working as Associate Professor in Electronics and Communication , Poojya Doddappa Appa College of Engineering, Kalabuaragi, Karnataka, India from May 2005 to present. he is guiding four students for their Ph. D degree under VTU in biomedical signal and image processing and evaluation of cognitive behaviour under various interventions such as yoga, video games, physical excises etc