

Impact of Noise in the Quatification of ILD Patterns in Lung CT Images

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Abstract

Biomedical image processing accomplishes breath taking developments and is considered integrating explorative area that has scope for research in engineering domains and medicine. Computerized problem identification of the disease has formerly evolved into decisive component in analytic and scientific trails and endeavours. Contemporary prophylactic interpretation approaches affordspromising overtures in disciplines like information and medicine as they outturn allegiance models and advances in automated form of memory and determination procedures for early detection, conclusion and treatment of cancer. The challenge lies in adequately excerpt, assess and enact the data for acquisition of morphological changes in the structural functions of body parts that are exemplified. For clinical practices, images are accumulated and stored in digital representation procured from MRI, CT, Ultrasound etc., in PACS, processing and thereby for diagnosis. Medical image segmentation is an important traipse for successive illustrations of patterns. The goal of lung image segmentation is to separate Region of Interest (ROI) for extracting lung abnormalities of Interstitial Lung Disease (ILD) impressions like Sarcoidosis, Idiopathic pulmonary fibrosis, Malignant nodules, and Sarcoidosis structures. These ROI's are affected by background regions and exhibit various levels of quality and brightness. In this paper morphology-based segmentation is used to extract honey comb and sarcoidosis patterns that are used in as much as pronouncement and prognostication of pneumonic disease. The performance appertaining to proposed method is evaluated in noisy environment. Salt and pepper noise, Speckle noise, Poisson noise, Gaussian noise are added to original DICOM image for the evaluation of noise effect. Comparison is implemented for honey comb and sarcoidosis patterns extracted from noisy and noiseless images. Noise reduction capabilities of proposed method on a particular noise type is validated based on correlation co-efficient and peak Signal-to-Noise ratio.

Keywords: Lung cancer, Nodules, Reconstruction, Morphology, Noise, correlation coefficient, peak Signal-to-Noise ratio.

Introduction

Refinement of representations pertaining to medicine (Medical image processing)manages complicated accessions to embellish rough pathological inputs for conscientious determination and some emphasize on applicable theory and applications [1, 28]. The major strength of computer applications rely on the refinement of the data based on the illustrations and accessions required for perceptible reasoning. Restorative models are generally ocular but are correlated by restraints of distractions due to inaccuracy, debility and defined sense where interpretations are subjective element. Computer analysis improves diagnostic

accuracy and confidence with thorough examinations using segmentation, registration and visualisation [2, 26]. Cancer is an invasive and emotive subject where millions of people worldwide are struggling without a final cure. Survival rate can be increased by early detection, lung cancer observed second in both men and women with death toll over 70%. Survival rates can be increased to 49% when detected at early stage and spreading can be restricted to lung out of lymph [2, 27]. Medical imaging field retains momentous systematic and experimental explorations related to victim's dossier through imaging approaches like CT, MRI, PET and US. Interpretations of particular modes come up with un-checked ameliorations that help pronouncements and therapeutic guesstimations. Pathological patterns, when subjected to certain series of actions for attaining a particular result will alter augmenting collections of objective reports and considers them in consequences of requirement of explicit medicinal data [3]. For test, dissolution and study of pathological patterns, image processing performance procedures and approaches to suppress unwanted distortions for feature enhancements that cognate to ruckus removal, intensification of representation and contour disclosure. Quality of counter parts are extricated to endorse the scope of significance that may either be lump, abrasion or aberrancy [14]. When incredulous linked components are identified, anomalous symbols or models should be substantially comprehended and separated to understand the progressive facts. Sovereign bio-medical germaneness supports illustrations for severance needed for supplementary transforms for refinement which is a must for esteemed accoutrement of dossier adumbration [3, 7]. Computed tomography majorly acknowledged as CAT reinforces encephalogram for computing in order to bring about 3D presumption of physical structure and material substances of human anatomy. CT counterparts recognises noticeable lung abnormalities like ILD and sarcoma. Apparent bronchi motif can be clearly examined from CT models where demarcation of lung promotes nature and pattern of image along with its examination [2]. Interstitial lung diseases (ILD) abide over and above 200 contrasting classes that are assorted associations of alveolar or bronchiole clutter because of septa congeal, cell connective tissue conception or decay and pulmo scarring or laceration. Preponderance of ILD's happen by reason of anonymous sources. Acknowledged ILDs are due to sniffing of anatomical or inessential filths, grime, cinders, effluvium, stimulant, emissions or may be tobacco or weed consumption. ILDs are type casted as acute, chronic and episodic. Acute ILDs are owing to contagion, aversion or contamination. Episodic ILDs are on account of edema tissue leukaemia or exotic naïve inflammation of lung air sacs. Chronic ILDs are induced mainly by cause of narcotic aftereffects in which the utmost prevalent are carcinoma, IPF, sarcoidosis and honey comb [3]. Nodule in pleura is a spot permeating abrasion in lung which will be of 3cm or may be still less in across the module. This lesion when spread larger than the measurement, it is considered as mass and beyond any doubt termed as cancer [22]. Identification and acknowledgement of these masses against legitimate forms target on indistinguishable segments of alveolus connective tissues [17].

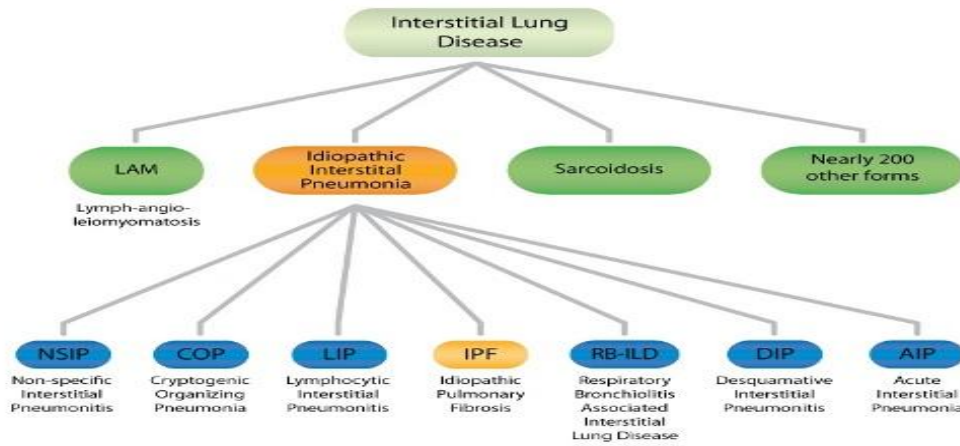


Fig.1. ILD Assortment.

Swellings that form like tumours or nodules acquire inconsistent attitudinal feautresuch as hoop, outstretched and developed at contour verges [20]. Typecasting of ILDs is shown in Fig. 1. Disparatesamples in consequence to ILDs considering lung protuberance are displayed in Fig. 2a, 2b & 2c.

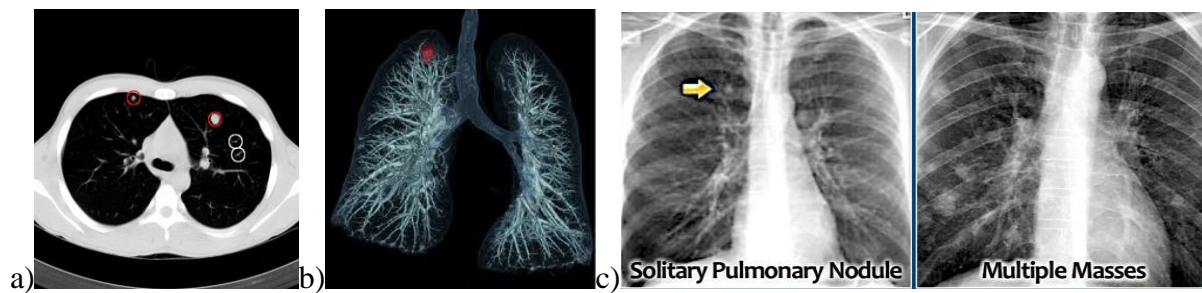


Fig.2. (a) Protuberance ofpulmonary lung inducing lung cancer. (b) Burl in Alveolous. (c) Mass.

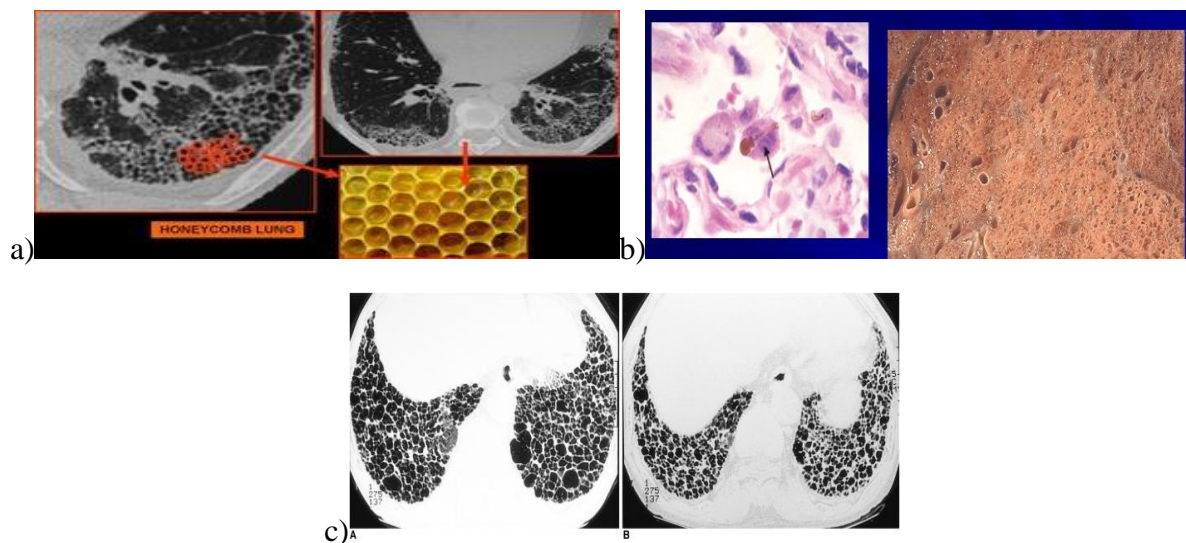


Fig.3. (a) Honey comb lung. (b) Extremity level of Honey comb model. (c) Extremity Honey comb visual.

Honey comb invokes CT revelations as disseminated connective tissue thickening and scarring causing tumour. Aggregated cyst tumour can be measured to 3 to 10mm yet far-reaching spreading can be 2.5cm. Honey comb impressions along with overwrought lung are demonstrated in Fig. 3a, 3b & 3c. Sarcoidosis affliction implicates aberrant accumulation as concerned with incendiary vacuoles systemised as clump or mass popular as granulomas. Anomaly of granulomas are displayed in the Fig. 4a, 4b & 4c. Idiopathic pulmonary fibrosis traumatise alveolus matter precipitating and aggravating eupnea paucity.

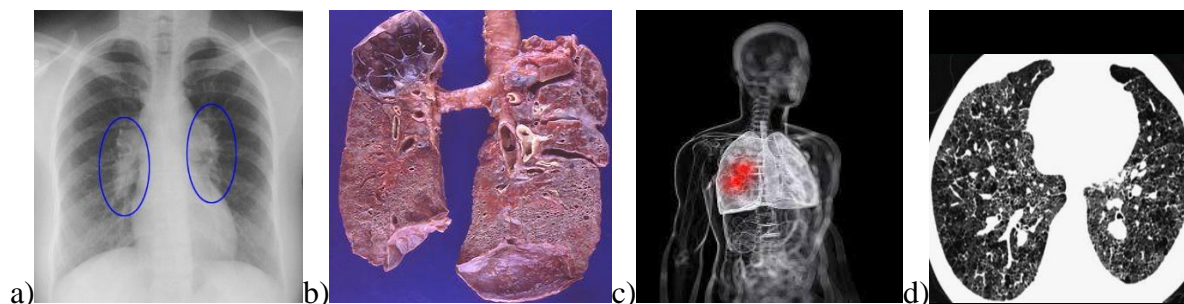


Fig. 4. (a) Sarcoidosis. (b) Sarcoidosis lung. (c) Sarcoidosis Lung in 3D. (d) Seriously affected lung with sarcoidosis.

2 Literature Review

Severance of pathological versions are epitomised schemes that are done in three concatenations. The elementary technique is finding of straight forward process to ascertain the degree of bioluminescence and its akin specks. The predated practice obtains through perception, drafting and transforming approaches [23]. The third system accesses apportionment by applying concoction component to the info. Determination of restorative representation is done in a persuasive process using demarcation method [8]. Stratified assembling techniques are probably used contemporarily as they accomplish appearances in distinct to main figure. This mode congregates counterparts by taking account of model intensity, manifestation and contiguous proportions. Regardless of stratifies procedures, there exists predicament in establishing the array reckoning which is a must for model severance. So these methods are of less importance considering these impacts. To overwhelm long-established mechanism deficiencies, this periodical conceives a principle to study forms and apart the scope to extract the diseased part from the body [24]. Demarcation of a model holds proceedings by acquainting scientific examination and analytical means separating the model motif from its countenance. If the model's primitive procedure is considered, its ROI is analysed from its object actualization features [25]. Quantity or frequency stationed methods are incorporated by using neighbourhood expanding, discontinuity and closure operations. Linguistic and diagnostic approaches segregate infinitesimal 3D contusions in alveolus by examining its extent, luster and the scope of concern [11]. Relegated hypothesis are entrenched to scrunch momentous characteristics that overrides the appearance range, display quality and aberration. Determined distinguished approaches are even supported for diagnosis by demarcating unsuspected characteristics of the object and also confide upon restraints such as sort of affliction and its condition which variates it from the general format. The above mentioned developments incite an arduous assignment as much the analyst has to aim over the composition that concerns to the illustrations [8].

3 Medical Image Segmentation

Segmentation endures as admissible artifice for contemporary refinement of restorative models and also directed towards attainment of info for improved point about class, condition and spot of the disparate ailments. Pathological changes and its elucidations cognizes disclosure of deep, superficial and pulmonary veins by refractometry, concludes edema muscular walls swellings, lacerations, gash exigency, calibrating cancerous anomalies and their operation and medication, restrained plasma poll, atypical surge in soft nervous tissue, spot divergent ranges of phylum that endures bosom carcinoma out of mammogram and culling cardiac and muscular muscles from fluoroscopy and also testimonies contusion. In consequence to the progress related to therapy, severance has a vital part to figure out anatomical attributes by their structural and descriptive considerate [6]. Inevitability of precision and determination of segregating a counterpart from its background purely depends on the compatible segmentation accession. Demarcation of an object construes its confines from the backdrop. Tinge, effulgence rate and format or aspect ratio determines appearance of an object in a realm. Prevalent contours of a representation will be exceptional if the scope is resolved and figured out [2]. Appliance of pathological model demarcation arbitrates the location for diagnosis. ROI concerned performances governs certain assertions of significance [8]. Tinge and aspect ratio vitally discovers silhouette that fall in analogous parts of the representation [12]. This bit of discovery process on meaningful ROI is crucial for pathological representations and their examination. Such medical supporting systems evaluate feasible groupings to set out interconnected breaks to acquire anatomical qualities of particular ROI [10]. Prophylactic depiction partitioning consummates observable differentia of the model belonging to the similar vigour of the significance [20, 16]. Severance is conspired to permeate region of significance that confers congregation on certain suspicious suburbs which gives rise to the augmentation of ailment. The split up effects drawn ensues aid ophthalmologist to determine medical peculiarities and treatment. Various pathological model observations find their relevancies in finding out lumps, flesh impression as shadows, realise and extract fundamental features, assortment and acknowledge certain parts of significances [14].

4 Concept of Morphology in Medical Image Processing

Morphology of models characterise imprecisely determined processes that akin toward silhouette, aspect and appearance of the representation [1]. This concept has ample purview on conformation of models [4]. Operations based on this study implies a grouping component to develop a model and gives an outcome by waiving the digression. Estimating the amount of precision can be possible by reconstruction method for which it uses class of operative elements. Binate applications of morphology benefits monochrome display models [16, 18]. These operative elements may be of different sizes and structure consonantly a 3x3 extended having its seed point from the equidistance pixel. Phonemic operations that adopts contour discovery discerns procedures just as deterioration, distention, opening and closing. Distention advances the model whereas deterioration dwindles the portions of representation. Opening adjournments the substance lineament and closing fillers the object lineament elements [20]. The apperception on ROI by CT modality is possible from the elicitation of beneficial structural characteristics. Counterparts are excerpted for early diagnosis by selecting the operative elements size and structure [21]. Grouping elements set applied on monochrome models have level sets [5, 17]. It is a tough task of identifying the confines dearth's likely the discontinuity of outlines and also the skimpy characteristic differences between the image and its backdrop. Watershed finding uses mathematical morphology for

its application. This model relies upon neighbourhood dependency techniques that forms as water lines that flow through the fields during agriculture. This appears like a corrugation and entire process pauses once the utmost innuendo point is attained [4, 13].

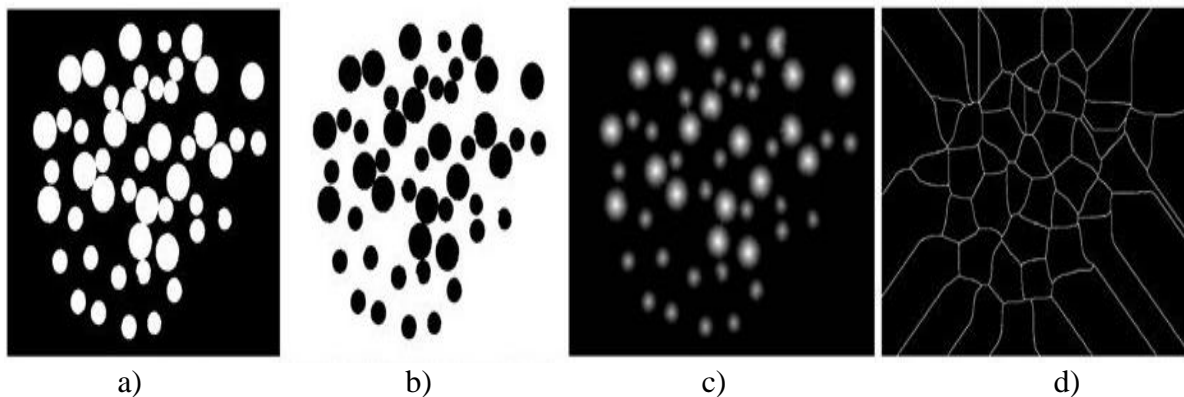


Fig. 6. (a) Original image. (b) Negative of a. (c) transform of b. (d) watershed of c.

The whole image is sectioned as rations that look like walls called watershed ridge lines. This appearances are attained from lexemic restoration. These ridge lines show the border regions of the counterparts and also seem like to be like their resemblance [20]. These lines are extracted through the severance process and the model is exhibited in Fig. 6 (a, b, c & d).

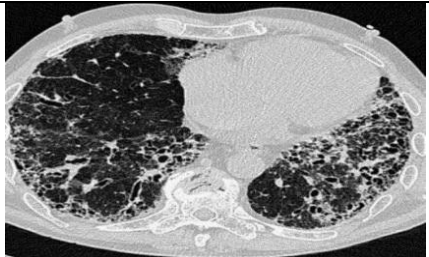

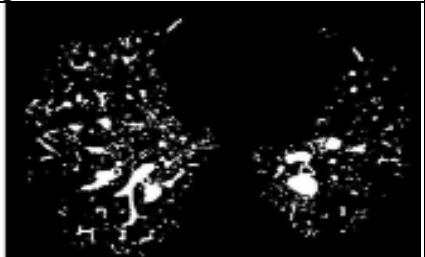
5 Reconstruction of Lesions

Lump restoration is consummated from proper excerption of the alveolus ROI. DICOM can never be in a layout of portfolio or quietly a processor. It brandishes pathological representations, accrues and stashes information, delivers indiscriminate data inconsiderate with something that augments [13]. DICOM will not degrade the quality of counterpart and liberates 2d models being refined with upright support by maintaining actual details. Noise is foreseen antagonistic form of energy which will misrepresent and abjures the file and is highly arduous to clear away. If it is adduced in computerised communication during simulacra procurement, compilation, organisation, transference or its subjection to series of action in order to achieve a result. When files are supplemented with noise they yield to capricious strange data [6]. This noise added to images is because of haphazard lop-sidedness in its luster and adumbration. It is forged from computers, thermionic apparatus many and optical scanning devices and gradually stashes factual files [8]. Many formats while processing undergo sequence of events by inducing noise to finally obtain legitimate data. The analogy of factual noise image that is prone and then recognised by noise filters will help for identification of disease [12]. **Gaussian noise** is ascribed as extra dissonance that unfolds overly in recurrence of any data expeditiously. **Poisson noise** familiar as photon noise occurs due to equivocal fluorescent disparities from apparatus when the data is being redeemed like during galvanism issuance from a typical poison dispersion [13]. **Detector noise** ensues against independent vitality transmission. **Speckle noise** happen to particularized info that is desist by the reason of transference infidelity. This noise is always vague, exotic and dubious and it turns up just as blanched drops over the object or representations which is also termed as texture [15]. **Salt and pepper noise** inhabits as a glitch when binal or multiple images are sent at identical instants. They appear as salt and pepper on the counterparts. The eminence of pepper noise nearby 0 and salt approaching 225 [18].

In order to excerpt ROI while it is inflicted to conduct something for achieving a denouement when the data is synthetically prone to noise in order it should not mystify the examinations and determinations. Counterparts are lay over with dissonant cacophonies to choose the proficiency of offbeat operators at the restoration concoct [23]. In this article DICOM image is allied with four variant noises like Speckle noise, Poisson noise, Gaussian noise and salt & pepper noise in order to bring out the lung ROI. DICOM format data can be checked out after it is sundered from the background organs. Demarcation process is delineated as cleaving data into assorted pieces. The idea behind demarcation is to clearly picturise the various functional attributes of the object and to specify the extremity of the portions even in the noise circumstances. To disqualify the noise that is added, lexemic and watershed algorithms are preferred as they can bring the ROI that is needed for restoration of the image [24]. Lexemic clarifications reveal the modification that occur to an image in its aspect, element and peculiarities of the counterparts by expunging superfluous info. The series of actions that are done to achieve the result will elucidate the reckonings eludes impediments. Watershed method highly adept to extract the ROI as it follows ridges too definitely to identify and separate the backgrounds and clear away noise details [25].

6 Results and Evaluation

Operators that distinguishes fringes are checked out on DICOM format of lung input along with noise and excluding noise for the restoration of lesions. Input data is depraved with four types of noises at the beginning of preliminary processing. Corroborations are done for all the noises by taking the evaluation limits such as correlation coefficient and peak signal to noise ratio. Correlation coefficient is principle measure to approximate the extent to which deuce data is compared. Peak signal to noise ratio characterizes the ratio of signal power to the noise power corrupting the signal. A model for lesion recognition of honeycombing patterns and sarcoidosis are demonstrated in Fig 7. Emanated outcomes from the figures can bolster medical doctors to diagnose and treat the disease at early stages of its growth by the end of analysis. The purpose of aforementioned evaluation estimates the noise reduction feature and aids to percept the distortion added to the CT format if it is transferred over a network where there is a prone to noise.

Original image	Lesion details of Honey comb pattern	Lesion details of Sarcoidosis pattern
		
Image with Salt & Pepper noise	Lesion details of Honey comb pattern with Salt & Pepper noise	Lesion details of Sarcoidosis pattern with Salt & Pepper noise




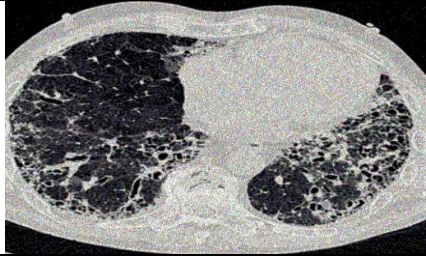
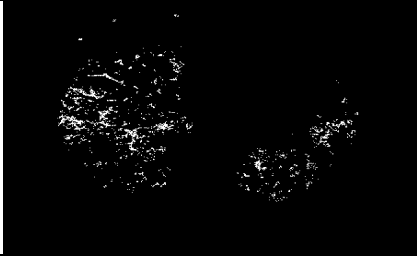
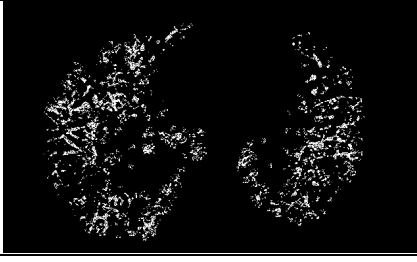

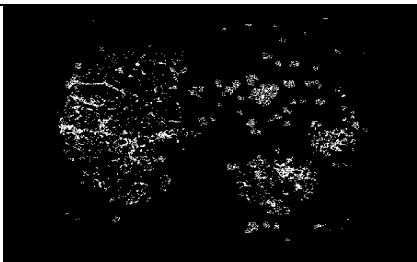
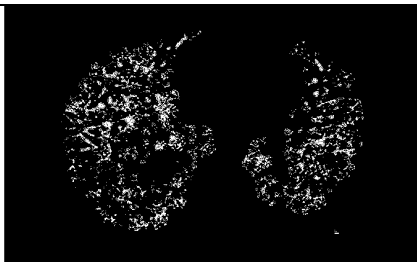
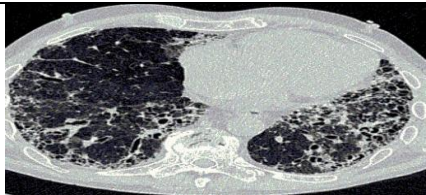
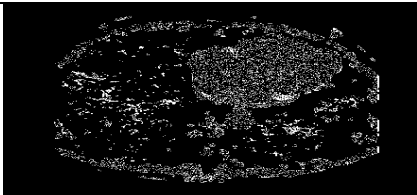

		
Image with Speckle noise	Lesion details of Honey comb pattern with Speckle noise	Lesion details of Sarcoidosis pattern with Speckle noise
		
Image with Gaussian noise	Lesion details of Honey comb pattern with Gaussian noise	Lesion details of Sarcoidosis pattern with Gaussian noise
		
Image with Poisson noise	Lesion details of Honey comb pattern with Poisson noise	Lesion details of Sarcoidosis pattern with Poisson noise
		

Fig. 7. Lesion extraction with different noise for sarcoidosis and Honey comb lung patterns.

The performance evaluation parameters PSNR and correlation coefficient are calculated for original image and reconstructed ROI and tabulated in Table 1.

Noise	PSNR		Correlation coefficient	
	Honey comb pattern	Sarcoidosis	Honey comb pattern	Sarcoidosis
Speckle	19.5911	17.7365	0.7437	0.8633
Gaussian	19.4712	19.3380	0.7542	0.9005
Poisson	22.9894	22.9868	0.8831	0.9556
Salt & Pepper	16.8879	16.5626	0.6751	0.8334

Table.1: Correlation Co-efficient and PSNR values of different ROI images in noisy environment for Honey comb and Sarcoidosis lung patterns.

7 Conclusions and Future scope

Preponderance of pulmonate abnormalities and associated apperceptions be sure of pertaining to the ameliorations of their shapes, borders and angels between the lung areas to the other parts to obtain ROI. The significance of ROI exhibit various levels of quality, brightness, and shapes. In this project, morphology-based segmentation is used to extract Sarcoidosis ILD pattern is developed and implemented. Morphology based region of interest segmentation is implemented to extract various lung patterns by lung ROI severance to detail about diseased portion. The final images extracted affords awesome features for discernible investigation to assess extracted ILD patterns of honeycomb and sarcoidosis. These results will assist the medical physicians for antecedent forecast, pronouncement, hospitalisation or medication of disease. In addition, the proposed method is evaluated in the presence of noise to estimate the noise reducing capability of the method for a particular noise. The peak signal to noise ratio and correlation coefficient are computed for extracted honeycomb and Sarcoidosis patterns from original and noise added images. It is observed that the proposed method has reduced the effect of Poisson noise compared to all other noises. This is identified by high values of PSNR and Correlation Co-efficient for Poisson noise compared to other noise.

Assessing the abnormality, growth progression and descriptive characteristics can be taken as future work. Additionally this method can be accomplished and embellished to estimate the elicited imitates by altering the grouping components using phonemic ROI severance and also for different ILD patterns.

References

1. Zhao Yu qian , Gui Wei-hua "Medical Images edge detection Based on Mathematical Morphology "Proceedings of 2005 IEEE Engineering in Medicine and Biology 27th Annual Conference, sep-2005.
2. V. Vijaya Kishore, R V S. Satyanarayana, "Computer-Aided Diagnosis Tool for Honeycomb Detection by using Morphology and WaveletTransform in Lung CT Images",IFRSA's International Journal Of Computing|Vol2|issue 2|April 2012.
3. Felix Ritter, Tobias Boskamp, André Homeyer, Hendrik Laue,MichaelSchwier, Florian Link, and Heinz-Otto Peitgen, "Medical Image Analysis", Proceedings of CVPR'96 IEEE PULSE, pp 2154-2287 November/December 2011

4. Mohamed Roushdy, "Comparitive Study of Edge detection Algorithms Applying on the Gray Scale Noisy Image using Morphological Filter", GVIP Journal, Vol-6, Issue-4, Dec-2006.
5. Albovik, "Handbook of Image and video Processing", Academic Press, 2000.
6. V. Vijaya Kishore, R V S. Satyanarayana, "A Matlab based GUI for Detection and Classification of Lung Nodules in CT Images", International Journal of Biomedical Research and Analysis, 4 (1), May-June 2013, 01-03.
7. Ping-Lin Chang and Wei-GuangTeng, "Exploiting the Self-Organizing Map for Medical Image Segmentation", IEEE International Symposium on Computer-Based Medical Systems (CBMS'07)0-7695-2905-4/07.
8. Tian Shen, Hongsheng Li & Xiaolei Huang, "Active Volume Modelo for Medical Image Segmentation", IEEE Transactions on Medical Imaging, VOL. 30, NO. 3, MARCH 2011.
9. Adegoke, B. O., Olawale, B. O., Olabisi, N.I., "Overview of Medical Image Segmentation", International Journal of Engineering Research and Development, Volume 8, Issue 9 (September 2013), PP. 13-17.
10. V. Vijaya Kishore, R V S. Satyanarayana, "A Modified Mathematical Morphology-Based Approach for Medical Image Edge Detection and Restoration", The IUPJournal of Electrical and Electronics Engineering, Volume 5, Issue 2, 2012.
11. Tuba Sirin, Mehmet IzzetSaglam et al. "A Comparative Evaluation of Competitive Learning Algorithms for Edge detection Enhancement".
12. V. Vijaya Kishore, R V S. Satyanarayana, "Comparative Study of Performance of Edge Detecting Methods on Medical Image in The Presence of GaussianAnd PoissonNoise", International Journal of Electronics and Electrical Engineering, Volume 5, Issue 7, June 2009.
13. P. Garcia, F. Pla, I. Garcia, "Detecting Edges in colour Images using dichromatic Differences", Seventh International IEEE Conference on Image processing and ITS Applications, volume: 1, 1999.pp.363-367.
14. Kushal Kr. Roy, Amit Phadikar, "Automated Medical Image Segmentation: A Survey", Proc. of Int. Conf. on Computing, Communication & Manufacturing 2014.
15. S. G. Chang, Y. Binand, M.Vetterli, "Adaptive Wavelet Thresholdingfor Image Denoising And Compression", IEEE Trans. On Image Processing, vol.9, no.9, pp. 1532-1546, sep 2006.
16. V. Vijaya Kishore, R V S. Satyanarayana, "A Multi-Functional Interactive Image processing tool for lung CT images", International Journal of Biomedical and clinical Engineering, 2(1), 1-11, January-June 2013.
17. Cronin, P. et al. Solitary Pulmonary Nodules: Meta-analytic Comparison of Cross-sectional Imaging Modalities for Diagnosis of Malignancy. Radiology. 2008, 246:772-782.
18. H. E Durdick, Digital Imaging: Theory and Applications, McGraw-hill, 1977.
19. Janaki. R and Dr.Tamilarasi. A, "Enhanced ROI (Region of Interest Algorithms) for Medical Image Compression", International Journal of Computer Applications (0975 – 8887) Volume 38– No.2, January 2012.
20. V. Kalpana, Dr S Varadarajan and T. Milindapurna, "Performance Evaluation of DICOM Lung ROI with Different Sizes of Morphological Structuring Elements and Noise", International journal of applied engineering research, vol. 10, no.6 pp.4991-4996, 2015.
21. Ahmed soliman, Fah mi khalifia, nil u, Neal dunlap and Brain waye, "Image Based CAD System for Accurate Identification of Lung Injury", IEEE 2016.
22. V. Kalpana, Prof GK Rajini, "Segmentation of Lung Lesion Nodules Using DICOM With Structuring Elements andNoise -A Comparative Study", IEEE digital xplore library, DOI: 10.1109/UPCON.2016.7894661.

23. V. Vijaya Kishore, V. Kalpana, “ROI Segmentation and Detection of Neoplasm Based on Morphology Using Segmentation Operators”, ISSN 1876-1100 ISSN 1876-1119 (electronic), Lecture Notes in Electrical Engineering, volume 569, Emerging Trends in Electrical, Communications, and Information Technologies, ISBN 978-981-13-8941-2 ISBN 978-981-13-8942-9 (eBook), <https://doi.org/10.1007/978-981-13-8942-9>, pp.501-510.
24. V. Kalpana, V. Vijaya Kishore, “A Common framework for the extraction of ILD patterns from CT image”, ISSN 1876-1100 ISSN 1876-1119 (electronic), Lecture Notes in Electrical Engineering, volume 569, Emerging Trends in Electrical, Communications, and Information Technologies, ISBN 978-981-13-8941-2 ISBN 978-981-13-8942-9 (eBook), <https://doi.org/10.1007/978-981-13-8942-9>, pp.511-520.
25. V. Vijaya Kishore, V. Kalpana, “Effect of Noise on Segmentation Evaluation Parameters”, Springer Nature Singapore Pte Ltd. 2020, M. Pant et al. (eds.), Soft Computing: Theories and Applications, Advances in Intelligent Systems and Computing 1154, https://doi.org/10.1007/978-981-15-4032-5_41.
26. Joyjitpatra, Himadrinathmoulick, Arunkanti manna, “Biomedical image processing with morphology and segmentation methods for medical image analysis”, American Journal of Engineering Research (AJER), e-ISSN: 2320-0847 p-ISSN: 2320-0936, Volume-02, Issue-07, pp-227-244.
27. HuseyinPolat, HomayDanaeiMehr, “Classification of Pulmonary CT Images by Using Hybrid 3D-Deep Convolutional Neural Network Architecture”, Appl. Sci. 2019, 9, 940; doi:10.3390/app9050940.
28. Dr. J. Thirumaran, S. Shylaja, “Medical Image Processing – An Introduction”, International Journal of Science and Research, ISSN (Online): 2319-7064, Volume 4 Issue 11, November 2015.