

Computerized Tomography Images Processing using Artificial Intelligence Techniques

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SCHOOL OF
MATHEMATICS SCIENCES AND
INFORMATION TECHNOLOGY

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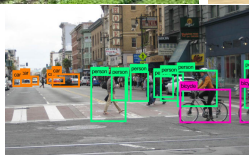
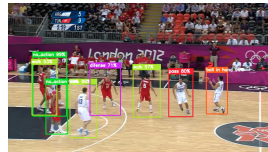
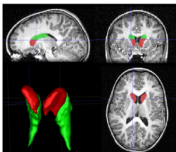
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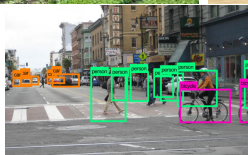
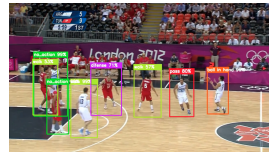
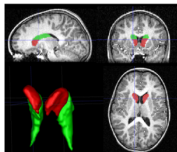
Study of Images

- Image analyses are vital in medicine, agriculture, sports, and object recognition.



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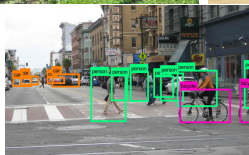
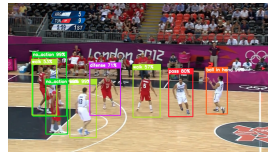
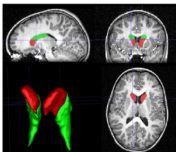
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- In the medical domain, images reveal their importance to improve the processes to treat patients.
- There is extensive research to contribute to medicine by segmentation and identifying areas of interest (e.g., organs)

Study of Images

- Researches has been focuses on Computerized Tomography (CT) images to apply image processing techniques due to its essential role in detecting diseases on the liver [1], lungs [2], brain [3], skull [4], and Covid-19 disease [5].

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- The segmentation of CT images is a challenging task due to several factors such as the complex background, dataset size, dimensions, noise, illumination, and shadows.
- Artificial Intelligence (AI) has different ways to automated processes like Machine Learning (ML), Artificial Neural Networks (ANN), and Convolutional Neural Networks (CNN).

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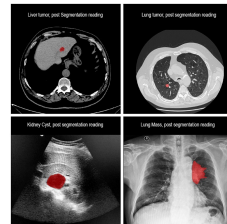
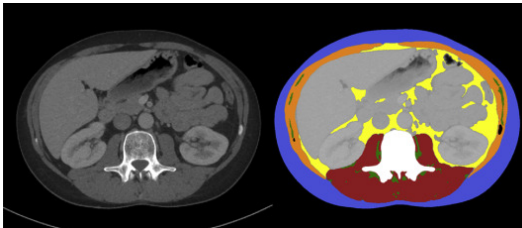
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Objective

This work aims to present state-of-the-art image segmentation using Computed Tomography images.



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Artificial Intelligence

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Artificial Intelligence

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- One of the most considerable advantages is on medicine by detecting anomalies or pathologies from digital images.



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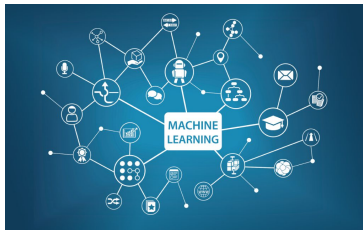
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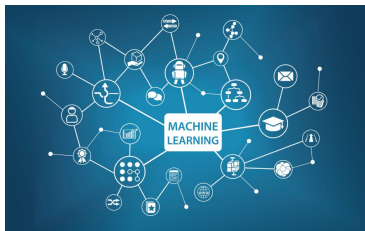
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- There are three subareas of ML : Supervised Learning, Unsupervised Learning, and Reinforcement Learning.
- A supervised model learns from a labeled dataset, and its task is to predict the correct output or label in a test dataset.



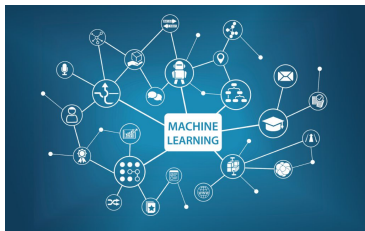
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- The reinforcement algorithms use AI agents to choose the best actions to maximize a numerical reward signal [7].



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- DL has dramatically impacted research areas, from image analysis to automotive, finance, and medical industry [9].
- The impact of DL is on different studies like recognition of breast cancer [10, 11, 12, 13], extraction of brain tumor [14, 15, 3], and organ segmentation [1, 16, 9].

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- Application of CNNs are: image classification [18, 19], edge detection [20], decoding facial recognition [21, 22], and analyzing documents [23, 24].

Convolutional Neural Networks

- A CNN has three fundamental layers: a convolutional layer, a pooling layer, and a fully connected layer

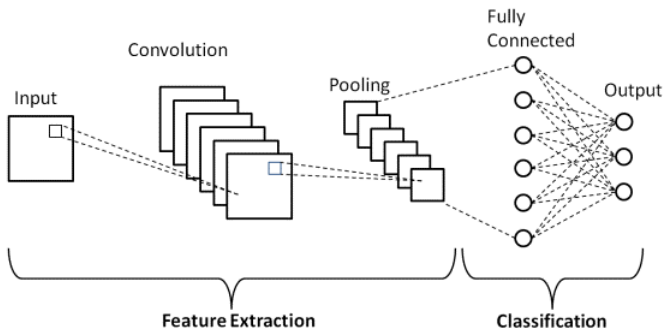


Figure 1: A CNN architecture showing the layers: convolutional, pooling, and fully connected [25].

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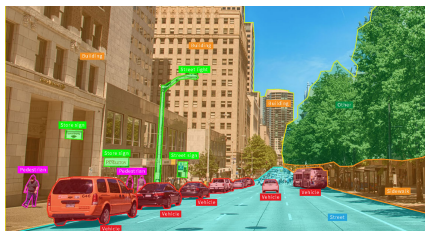
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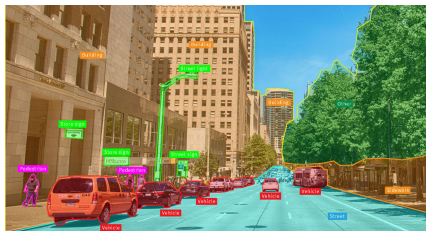
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- There are many applications of computer vision such as machine inspection, retail, 3D model building, medical imaging, automotive safety, and biometrics [26].



Computer Vision

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- Image processing applications involve the medical, agriculture, sports, and object recognition fields.



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- **Computerized Tomography image** refers to an x-ray image. It can show the skeleton, organs, and tissues as well as any abnormalities the physician is trying to identify.



Figure 2: Example of a Computerized Tomography Image from Lung [9]

Computer Vision

- **Computerized Tomography image** refers to an x-ray image. It can show the skeleton, organs, and tissues as well as any abnormalities the physician is trying to identify.
- Several challenges arise when processing images such as noise, complex background, dataset size, and shadows.



Figure 2: Example of a Computerized Tomography Image from Lung [9]

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Related Works

- Susomboom et al. [1] worked on a hybrid method to achieve a single-organ segmentation automatically. They used CT images of the liver in a DICOM format¹. The algorithms used were the split-and-merge and region-growing.

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- Seng et al. [4] designed a framework to extract objects from CT images using Reinforcement Learning (RL).
- Guo et al. [2] combine thresholding and the Chan-Vese algorithm to segment many types of images.

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- Wang et al. [28] propose a framework for multi-organ segmentation using Organ-Attention-Network with Reverse Connections (OAN-RCs). The OAN is a deep convolutional network that reduces the complex background
- Sakboonyara et al. [29] propose an efficient method for liver segmentation. They mention the use of U-Net and the mean-shift clustering algorithm. Also, they introduce statistical thresholding to improve the accuracy.

Related Works

- Zhou et al. [16] made an implementation based on multi-organ segmentation. They use partially-labeled datasets of the pancreas. The authors applied the Prior-aware Neural Network (PaNN) to overcome the context background problem.

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- Dabiri [30] proposed a study of medical images to find the middle axial slice at the L3 level automatically from a full or partial body CT image. The algorithm uses a fully convolutional classifier.
- Yan et al. [5] use a new deep convolutional neural network called COVIDSegNet to segment infected regions and the entire lung from chest CT images. The proposed network focuses on Feature Variation (FV) block and Progressive Atrous Spatial Pyramid Pooling (PASPP).

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Approaches	Year	Method	Neural Networks	Deep Learning	Machine Learning	Other Methods
Yan et al. [5]	2021	Architecture with Feature Variation and Progressive Atrous Spatial Pyramid Pooling	x			
Dabiri et al. [30]	2020	Algorithm of L3		x		
Wang et al. [28]	2019	OAN - RCs and statistical fusion	x			
Sakboonyara et al. [29]	2019	U-Net and Mean-Shift Histogram	x			
Zhou et al. [16]	2019	PaNN	x			
Xu et al. [27]	2017	Membership function convolution neural network	x			
Guo et al. [2]	2015	Iterative Chan-Vese				x
Seng et al. [4]	2011	Reinforcement Learning			x	
Susomboon et al. [1]	2006	Pixel-based, split-and-merge, and region growing algorithms				x

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- Image processing is crucial in many domains such as computer sciences, industry, security, and medicine.
- In recent years, the study of Computerized Tomography (CT) images reveals its importance in the medical field by helping to detect diseases like tumors, cancer, or control organs with problems.
- The automation of algorithms using advanced techniques such as Convolutional Neural Networks can improve the accuracy and performance of results.
- Previous work reveals that it is necessary to use traditional image processing techniques in the data preparation stage of the algorithms.

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- [1] R. Susomboon, D. Raicu, J. Furst, and D. Channin, “Automatic single-organ segmentation in computed tomography images,” in *Sixth International Conference on Data Mining (ICDM’06)*, (Los Alamitos, CA, USA), pp. 1081–1086, IEEE Computer Society, dec 2006.
- [2] S. Guo and L. Wang, “Automatic ct image segmentation of the lungs with an iterative chan-vede algorithm,” in *2015 International Conference on Informatics, Electronics and Vision (ICIEV)*, (Los Alamitos, CA, USA), pp. 1–4, IEEE Computer Society, jun 2015.

- [3] S. E. kaitouni and H. Tairi, “Segmentation of medical images for the extraction of brain tumors: A comparative study between the hidden markov and deep learning approaches,” in *2020 International Conference on Intelligent Systems and Computer Vision (ISCV)*, (Los Alamitos, CA, USA), pp. 1–5, IEEE Computer Society, jun 2020.
- [4] W. Seng and M. Chitsaz, “Medical image segmentation by using reinforcement learning agent,” in *Digital Image Processing, International Conference on*, (Los Alamitos, CA, USA), pp. 216–219, IEEE Computer Society, mar 2009.
- [5] Q. Yan, B. Wang, D. Gong, C. Luo, W. Zhao, J. Shen, J. Ai, Q. Shi, Y. Zhang, S. Jin, L. Zhang, and Z. You, “Covid-19 chest ct image segmentation network by multi-scale fusion and enhancement operations,” *IEEE Transactions on Big Data*, vol. 7, pp. 13–24, jan 2021.

- [6] M. A. Shahin, “State-of-the-art review of some artificial intelligence applications in pile foundations,” *Geoscience Frontiers*, vol. 7, no. 1, pp. 33–44, 2016.
- [7] R. S. Sutton and A. G. Barto, *Reinforcement learning: An introduction*. MIT Press, 2018.
- [8] A. Abraham, “Artificial Neural Networks,” *Handbook of Measuring System Design*, 2005.
- [9] S. Stefaniga and M. Gaianu, “An approach of segmentation method using deep learning for ct medical images,” in *2019 21st International Symposium on Symbolic and Numeric Algorithms for Scientific Computing (SYNASC)*, (Los Alamitos, CA, USA), pp. 273–279, IEEE Computer Society, sep 2019.

- [10] A. Chorianopoulos, I. Daramouskas, I. Perikos, F. Grivokostopoulou, and I. Hatzilygeroudis, "Deep learning methods in medical imaging for the recognition of breast cancer," in *2020 11th International Conference on Information, Intelligence, Systems and Applications (IISA)*, (Los Alamitos, CA, USA), pp. 1–8, IEEE Computer Society, jul 2020.
- [11] M. Jannesari, M. Habibzadeh, H. Aboulkheyr, P. Khosravi, O. Elemento, M. Totonchi, and I. Hajirasouliha, "Breast cancer histopathological image classification: A deep learning approach," in *2018 IEEE International Conference on Bioinformatics and Biomedicine (BIBM)*, (Los Alamitos, CA, USA), pp. 2405–2412, IEEE Computer Society, dec 2018.

- [12] W. Zou, H. Lu, K. Yan, and M. Ye, “Breast cancer histopathological image classification using deep learning,” in *2019 10th International Conference on Information Technology in Medicine and Education (ITME)*, (Los Alamitos, CA, USA), pp. 53–57, IEEE Computer Society, aug 2019.
- [13] A. Chorianopoulos, I. Daramouskas, I. Perikos, F. Grivokostopoulou, and I. Hatzilygeroudis, “Deep learning methods in medical imaging for the recognition of breast cancer,” in *2020 11th International Conference on Information, Intelligence, Systems and Applications (IISA)*, (Los Alamitos, CA, USA), pp. 1–8, IEEE Computer Society, jul 2020.

- [14] Z. Xiao, R. Huang, Y. Ding, T. Lan, R. Dong, Z. Qin, X. Zhang, and W. Wang, "A deep learning-based segmentation method for brain tumor in mr images," in *2016 IEEE 6th International Conference on Computational Advances in Bio and Medical Sciences (ICCABS)*, (Los Alamitos, CA, USA), pp. 1–6, IEEE Computer Society, oct 2016.
- [15] Y. Li, X. Xie, S. Liu, X. Li, and L. Shen, "Gt-net: A deep learning network for gastric tumor diagnosis," in *2018 IEEE 30th International Conference on Tools with Artificial Intelligence (ICTAI)*, (Los Alamitos, CA, USA), pp. 20–24, IEEE Computer Society, nov 2018.

- [16] Y. Zhou, Z. Li, S. Bai, X. Chen, M. Han, C. Wang, E. Fishman, and A. Yuille, "Prior-aware neural network for partially-supervised multi-organ segmentation," in *2019 IEEE/CVF International Conference on Computer Vision (ICCV)*, (Los Alamitos, CA, USA), pp. 10671–10680, IEEE Computer Society, nov 2019.
- [17] S. Albawi, T. A. M. Mohammed, and S. Alzawi, "Understanding of a Convolutional Neural Network," *IEEE*, 2017.
- [18] A. Ramanath, S. Muthusrinivasan, Y. Xie, S. Shekhar, and B. Ramachandra, "Ndvi versus cnn features in deep learning for land cover clasification of aerial images," in *IGARSS 2019 - 2019 IEEE International Geoscience and Remote Sensing Symposium*, pp. 6483–6486, 2019.

- [19] M. T. Islam, B. M. N. Karim Siddique, S. Rahman, and T. Jabid, "Food image classification with convolutional neural network," in *2018 International Conference on Intelligent Informatics and Biomedical Sciences (ICIIBMS)*, vol. 3, pp. 257–262, 2018.
- [20] W. Xue, X. Wenxia, and L. Guodong, "Image edge detection algorithm research based on the cnn's neighborhood radius equals 2," in *2016 International Conference on Smart Grid and Electrical Automation (ICSGEA)*, pp. 115–119, 2016.
- [21] D. Qu, Z. Huang, Z. Gao, Y. Zhao, X. Zhao, and G. Song, "An automatic system for smile recognition based on cnn and face detection," in *2018 IEEE International Conference on Robotics and Biomimetics (ROBIO)*, pp. 243–247, 2018.

- [22] Y. Zhang, M. Zhao, L. Yan, T. Gao, and J. Chen, "Cnn-based anomaly detection for face presentation attack detection with multi-channel images," in *2020 IEEE International Conference on Visual Communications and Image Processing (VCIP)*, pp. 189–192, 2020.
- [23] S. C. Kosaraju, M. Masum, N. Z. Tsaku, P. Patel, T. Bayramoglu, G. Modgil, and M. Kang, "Dot-net: Document layout classification using texture-based cnn," in *2019 International Conference on Document Analysis and Recognition (ICDAR)*, pp. 1029–1034, 2019.
- [24] L. Chen, S. Wang, W. Fan, J. Sun, and N. Satoshi, "Deep learning based language and orientation recognition in document analysis," in *2015 13th International Conference on Document Analysis and Recognition (ICDAR)*, pp. 436–440, 2015.

- [25] S. Malathy, R. R. Karthiga, K. Swetha, and G. Preethi, “Disease detection in fruits using image processing,” in *2021 6th International Conference on Inventive Computation Technologies (ICICT)*, pp. 747–752, 2021.
- [26] R. Szeliski, *Computer vision: algorithms and applications*. Springer Science & Business Media, 2010.
- [27] J. Xu and H. Liu, “Segmentation of pulmonary ct image by using convolutional neural network based on membership function,” in *2017 IEEE International Conference on Computational Science and Engineering (CSE) and IEEE International Conference on Embedded and Ubiquitous Computing (EUC)*, vol. 1, (Los Alamitos, CA, USA), pp. 198–203, IEEE Computer Society, jul 2017.

- [28] Y. Wang, Y. Zhou, W. Shen, S. Park, E. K. Fishman, and A. L. Yuille, “Abdominal multi-organ segmentation with organ-attention networks and statistical fusion,” *Medical Image Analysis*, vol. 55, pp. 88–102, 2019.
- [29] B. Sakboonyara and P. Taeprasartsit, “U-Net and Mean-Shift Histogram for Efficient Liver Segmentation from CT Images,” *2019 11th International Conference on Knowledge and Smart Technology, KST 2019*, pp. 51–56, 2019.
- [30] S. Dabiri, K. Popuri, C. Ma, V. Chow, E. M. Feliciano, B. J. Caan, V. E. Baracos, and M. F. Beg, “Deep learning method for localization and segmentation of abdominal CT,” *Computerized Medical Imaging and Graphics*, vol. 85, no. August, p. 101776, 2020.

Thanks!