

**SISTEM DIAGNOSIS PENYAKIT GINJAL BERDASARKAN CITRA
CT-RADIOGRAPHY MENGGUNAKAN *CONVOLUTIONAL NEURAL
NETWORK (CNN) MODEL RESNET-50***

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ABSTRAK

SISTEM DIAGNOSIS PENYAKIT GINJAL BERDASARKAN CITRA CT-RADIOGRAPHY MENGGUNAKAN CONVOLUTIONAL NEURAL NETWORK (CNN) MODEL RESNET-50

Penyakit ginjal dialami oleh lebih dari 10% populasi global yang berjumlah lebih dari 800 juta individu pada tahun 2022. Pada tahun 2040, penyakit ginjal diperkirakan menjadi urutan ke-5 sebagai penyebab kematian utama. Penyakit ginjal paling umum yang dapat mengganggu fungsi ginjal diantaranya Renal Cell Carcinoma (tumor ginjal), kista ginjal, dan nefrolitiasis (batu ginjal). Sebelum dilakukan pengobatan, diagnosis penyakit ginjal penting dilakukan untuk menentukan jenis dan tingkat keparahan penyakit ginjal yang dialami oleh seorang pasien, salah satunya menggunakan citra Computer Tomography (CT) Radiography. Berdasarkan data citra yang diperoleh, terdapat data normal dan 3 jenis penyakit ginjal, yaitu kista, tumor, dan batu ginjal. Tujuan dari penelitian ini yaitu mengetahui performa Convolutional Neural Network (CNN) model ResNet-50 dalam mendiagnosis penyakit ginjal berdasarkan hasil klasifikasi penyakit ginjal. Hasil yang optimal pada penelitian ini diperoleh pada uji coba learning rate 0,001 dengan probabilitas dropout 0,1 dan jumlah mini batch size 32. Uji coba tersebut menghasilkan akurasi, sensitivitas, dan spesifisitas masing-masing 100%, 100%, dan 100% dengan waktu komputasi 15 jam 53 menit 34 detik. Hasil tersebut menunjukkan bahwa metode CNN model ResNet-50 mampu mendiagnosis penyakit ginjal dengan baik.

Kata kunci: CNN, Penyakit ginjal, ResNet-50

ABSTRACT

DIAGNOSIS SYSTEM OF KIDNEY DISEASES BASED ON CT-RADIOGRAPHY IMAGES USING CONVOLUTIONAL NEURAL NETWORK (CNN) RESNET-50 MODEL

Kidney disease affects more than 10% of the global population of more than 800 million individuals in 2022. In 2040, kidney disease is predicted to be the 5th leading cause of major death. The most common kidney diseases that can impair kidney function include Renal Cell Carcinoma (kidney tumor), kidney cysts, and nephrolithiasis (kidney stones). Before treatment can be administered, the diagnosis of kidney disease is essential to determine the type and severity of kidney disease experienced by a patient, one of which can be done using Computer Tomography (CT) Radiography images. Based on the obtained image data, there are normal data and three types of kidney diseases, namely cysts, tumors, and kidney stones. The aim of this research is to determine the performance of the Convolutional Neural Network (CNN) ResNet-50 model in diagnosing kidney disease based on the results of kidney disease classification. Optimal results in this research were obtained in a learning rate of 0.001 with a dropout probability of 0.1 and a mini batch size of 32. The trial produced accuracy, sensitivity and specificity of 100%, 100% and 100% respectively with a computing time of 15 hours 53 minutes 34 seconds. These results show that the CNN ResNet-50 model method is able to diagnose kidney disease good.

Keywords: Kidney disease, CNN, ResNet-50

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DAFTAR PUSTAKA

- Abhirawa, H., Jondri, and Arifianto, A. (2017). Pengenalan Wajah Menggunakan Convolutional Neural Network. In *e-Proceeding of Engineering*, volume 4, pages 4907–4916.
- Agrawal, S., Patel, M., and Sinhal, A. (2021). *An Enhance Security of the Color Image Using Asymmetric RSA Algorithm*. Springer Singapore.
- Agrawal, S. A., Rewaskar, V. D., Agrawal, R. A., Chaudhari, S. S., Patil, Y., and Agrawal, N. S. (2023). Advancements in NSFW Content Detection: A Comprehensive Review of ResNet-50 Based Approaches. *International Journal of Intelligent Systems and Applications in Engineering*, 11(4):41–45.
- Ahsan, M. M., Luna, S. A., and Siddique, Z. (2022). Machine-Learning-Based Disease Diagnosis : A Comprehensive Review. *Healthcare*, pages 1–30.
- Alelign, T. and Petros, B. (2018). Kidney Stone Disease: An Update on Current Concepts. *Advances in Urology*, 2018.
- Ali, K., Shaikh, Z. A., Khan, A. A., and Laghari, A. A. (2022). Multiclass Skin Cancer Classification Using EfficientNets – A First Step Towards Preventing Skin Cancer. *Neuroscience Informatics*, 2(4):100034.
- Allaam, M. R. R. (2021). Klasifikasi Genus Tanaman Anggrek Menggunakan Metode Convolutional Neural Network (CNN) Program Studi Sarjana. In *e-Proceeding of Engineering*, volume 8, pages 3147–3179.

- Allport, P. M. (2023). *Applications of Fault-Tolerant Software Architecture Principles in The Detection of Adversarial Attacks*.
- Almalkawi, I. T., Halloush, R., Alsarhan, A., Al-Dubai, A., and Al-karaki, J. N. (2019). A Lightweight and Efficient Digital Image Encryption Using Hybrid Chaotic Systems for Wireless Network Applications. *Journal of Information Security and Applications*, 49:102384.
- Alqumboz, A. and Naser, M. (2019). Classification of Avocado Using Deep Learning. *International Journal of Academic Engineering Research (IJAER)*, 3(12):30–34.
- Amin, A., Sari, Y. A., and Adinugroho, S. (2019). Klon Perilaku Menggunakan Jaringan Saraf Tiruan Konvolusional Dalam Game SuperTuxKart. *Jurnal Pengembangan Teknologi Informasi dan Ilmu Komputer (J-PTIIK)*, 3(1):8454–8464.
- Arifin, T. and Ariesta, D. (2019). Prediksi Penyakit Ginjal Kronis Menggunakan Algoritma Naive Bayes Classifier Berbasis Particle Swarm Optimization. *Jurnal Tekno Insentif*, 13(1):26–30.
- Ashiquzzaman, A., Tushar, A. K., Islam, M. R., Shon, D., Im, K., Park, J.-h., Lim, D.-s., and Kim, J. (2017). Reduction of Overfitting in Diabetes Prediction Using Deep Learning Neural Network. In *IT Convergence and Security 2017*, pages 35–43.
- Bakheet, S. and Al-Hamadi, A. (2021). Automatic Detection of COVID-19 using Pruned GLCM-Based Texture Features and LDCRF Classification. *Computers in Biology and Medicine*, 137:1–10.

- Bilal, M. A., Ji, Y., Wang, Y., Akhter, M. P., and Yaqub, M. (2022). Early Earthquake Detection Using Batch Normalization Graph Convolutional Neural Network (BNGCNN). *Applied Sciences (Switzerland)*, 12(7548).
- Brownstein, A. J., Bin Mahmood, S. U., Saeyeldin, A., Mejia, C. V., Zafar, M. A., Li, Y., Rizzo, J. A., Dahl, N. K., Erben, Y., Ziganshin, B. A., and Elefteriades, J. A. (2019). Simple renal cysts and bovine aortic arch: Markers for aortic disease. *Open Heart*, 6(1).
- Cai, S., Shu, Y., Chen, G., Ooi, B. C., Wang, W., and Zhang, M. (2019). Effective and Efficient Dropout for Deep Convolutional Neural Networks. pages 1–12.
- Chhabra, M. and Kumar, R. (2022). An Efficient ResNet-50 based Intelligent Deep Learning Model to Predict Pneumonia from Medical Images. *2022 International Conference on Sustainable Computing and Data Communication Systems (ICSCDS)*, pages 1714–1721.
- Curhan, G. C. (2023). Kidney Stones in Adults: Epidemiology and Risk Factors. *UpToDate*.
- Das, S., Kharbanda, K., M, S., Raman, R., and D, D. E. (2021). Deep Learning Architecture Based On Segmented Fundus Image Features for Classification of Diabetic Retinopathy. *Biomedical Signal Processing and Control*, 68:102600.
- Ding, B., Qian, H., and Zhou, J. (2018). Activation Function and Their Characteristics in Deep Neural Networks. In *The 30th Chinese Control and Decision Conference (2018 CCDC)*, pages 1836–1841.
- Dokuz, Y. and Tufekci, Z. (2021). Mini-Batch Sample Selection Strategies for Deep Learning Based Speech Recognition. *Applied Acoustics*, 171:107573.

- Dwipayanti, I. A. K. (2022). *Gambaran Kristal Urine pada Pekerja Bata Merah di Desa Keramas, Blahbatu, Gianyar*. PhD thesis, Poltekkes Denpasar.
- Fachmi, R., Hidayatno, A., and Soetrisno, Y. A. A. (2020). Sistem Identifikasi Ukuran Tubuh Menggunakan Metode Convolutional Neural Network (CNN). *Transient: Jurnal Ilmiah Teknik Elektro*, 9(1):1–7.
- Feng, B., Zhou, H., Li, G., Zhang, Y., Sood, K., and Yu, S. (2021). Enabling Machine Learning with Service Function Chaining for Security Enhancement at 5G Edges. *IEEE Network*, 35:196–201.
- Filler, G., Salerno, F., McIntyre, C. W., and de Ferris, M. E. G. (2021). Animal, Human, and ^{23}Na MRI Imaging Evidence for the Negative Impact of High Dietary Salt in Children. *Current Pediatrics Reports*, 9(4):110–117.
- Foeady, A. Z. (2019). Sistem Klasifikasi Kanker Kulit Berdasarkan Data Citra Dermoscopic dengan menggunakan Metode Deep Extreme Learning Machine. page 144.
- Foreman, K. J., Marquez, N., Dolgert, A., Fukutaki, K., Fullman, N., McGaughey, M., Pletcher, M. A., Smith, A. E., Tang, K., Yuan, C. W., Brown, J. C., Friedman, J., He, J., Heuton, K. R., Holmberg, M., Patel, D. J., Reidy, P., Carter, A., Cercy, K., Chapin, A., Douwes-Schultz, D., Frank, T., Goettsch, F., Liu, P. Y., Nandakumar, V., Reitsma, M. B., Reuter, V., Sadat, N., Sorensen, R. J., Srinivasan, V., Updike, R. L., York, H., Lopez, A. D., Lozano, R., Lim, S. S., Mokdad, A. H., Vollset, S. E., and Murray, C. J. (2018). Forecasting Life Expectancy, Years of Life Lost, and All-Cause and Cause-Specific Mortality for 250 Causes of Death: Reference and Alternative Scenarios for 2016–40 for 195 Countries and Territories. *The Lancet*, 392(10159):2052–2090.

- Fuadah, Y. N., Ubaidullah, I. D., Ibrahim, N., Taliningsing, F. F., Sy, N. K., and Pramuditho, M. A. (2022). Optimasi Convolutional Neural Network dan K-Fold Cross Validation pada Sistem Klasifikasi Glaukoma. *ELKOMIKA: Jurnal Teknik Energi Elektrik, Teknik Telekomunikasi, Teknik Elektronika*, 10(3):728–741.
- Gonzalez, R. C. and Woods, R. E. (2018). *Digital Image Processing*. 4 edition.
- Hadibrata, E. (2023). Hipertensi Berhubungan Dengan Derajat Batu Ginjal. *Jurnal Penelitian Perawat Profesional Volume*, 5(2):795–802.
- Haji, S. H. and Abdulazeez, A. M. (2021). Comparison Of Optimization Techniques Based on Gradient Descent Algorithm: A Review. *Journal Of Archaeology Of Egypt/Egyptology*, 18(4):2715–2743.
- Hannan, S. A. and Pal, P. (2023). Detection and classification of kidney disease using convolutional neural networks. *J Neurol Neurorehab Res*, 8(August):1.
- Haq, D. Z. (2020). *Klasifikasi Citra Kanker Kulit Menggunakan Convolutional Neural Network Model Googlenet*. PhD thesis.
- Harahap, F. A. A., Mardianson Sinaga, R., Arifin, K., and Saputra, K. (2022). Implementasi Algoritma Convolutional Neural Network Untuk Mendeteksi Penyakit Ginjal. *Jurnal Teknologi Informasi, Komputer dan Aplikasinya (JTIKA)*, 4(2):212–219.
- He, K., Zhang, X., Ren, S., and Sun, J. (2016). Deep Residual Learning for Image Recognition. In *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*, pages 770–778.
- Hossain, M. R., Afroze, S., Siddique, N., and Hoque, M. M. (2020). Automatic

- Detection of Eye Cataract using Deep Convolution Neural Networks (DCNNs). *2020 IEEE Region 10 Symposium, TENSYP 2020*, (June):1333–1338.
- Hsiao, T. Y., Chang, Y. C., Chou, H. H., and Chiu, C. T. (2019). Filter-Based Deep-Compression with Global Average Pooling for Convolutional Networks. *Journal of Systems Architecture*, 95:9–18.
- Hsieh, J. J., Purdue, M. P., Signoretti, S., Swanton, C., Albiges, L., Schmidinger, M., Heng, D. Y., Larkin, J., and Ficarra, V. (2017). Renal cell carcinoma. *Nature Reviews Disease Primers*, 3:1–19.
- Ikechukwu, A. V., Murali, S., Deepu, R., and Shivamurthy, R. (2021). ResNet-50 vs VGG-19 vs training from scratch: A comparative analysis of the segmentation and classification of Pneumonia from chest X-ray images. *Global Transitions Proceedings*, 2(2):375–381.
- Ioffe, S. and Szegedy, C. (2015). Batch Normalization: Accelerating Deep Network Training by Reducing Internal Covariate Shift. *International Conference on Machine Learning*, pages 448–456.
- Islam, M. N., Hasan, M., Hossain, M. K., Alam, M. G. R., Uddin, M. Z., and Soylu, A. (2022). Vision transformer and explainable transfer learning models for auto detection of kidney cyst, stone and tumor from CT-radiography. *Scientific Reports*, 12(11440):1–14.
- Ismael, S. A. A., Mohammed, A., and Hefny, H. (2019). An Enhanced Deep Learning Approach for Brain Cancer MRI Images Classification using Residual Networks. *Artificial Intelligence In Medicine*, 102:1–14.
- Ismail, N. and Lestari, S. (2023). Mendiagnosis Penyakit Ginjal Kronis

- Menggunakan Algoritme C4.5. *Seminar Nasional Hasil Penelitian dan Pengabdian Masyarakat 2023*, pages 25–31.
- Jaya, I. F. (2023). Edukasi Pengetahuan Pembatasan Cairan pada Pasien Gagal Ginjal Kronik yang Menjalani Hemodialisis. *Indonesian Journal Of Community Service*, 3(2):61–68.
- Jianfei, C. and Changming, Z. (2021). Research on Image Recognition Based on Improved ResNet. In *2020 IEEE 6th International Conference on Computer and Communications*, pages 1422–1426.
- Jihaannuriy, M. A. (2023). *Pengenalan Pelat Nomor Otomatis Menggunakan Metode Inception-ResNet v2*. Skripsi, UIN Syarif Hidayatullah Jakarta.
- Kakian, F., Palangi, M. G., and Hadi, N. (2021). Epidemiology of Kidney Stone and Bacterial Strains with Antibiotic Resistance in Shiraz, Southwest of Iran during 2014-2019. *International Journal of Epidemiology and Health Sciences*, 2(3):1.
- Kaya, Y. and Gürsoy, E. (2023). A Novel Multi-Head CNN Design to Identify Plant Diseases using The Fusion of RGB Images. *Ecological Informatics*, 75:101998.
- Kementerian Kesehatan Republik Indonesia (2018). Riset Kesehatan Dasar.
- Khan, M. S., Salsabil, N., Alam, M. G. R., Dewan, M. A. A., and Uddin, M. Z. (2022). CNN-XGBoost Fusion-Based Affective State Recognition using EEG Spectrogram Image Analysis. *Scientific Reports*, 12(1):1–19.
- Kholilurrahman, M., Syafei, W. A., and Nurhayati, O. D. (2023). Image Processing Classification of Rice Leaf Color Images Using the Convolutional Neural Network Method. *Jurnal Ilmiah Sains*, 23(2):175–186.

- Kittipongdaja, P. and Siriborvornratanakul, T. (2022). Automatic kidney segmentation using 2.5D ResUNet and 2.5D DenseUNet for malignant potential analysis in complex renal cyst based on CT images. *Eurasip Journal on Image and Video Processing*, 2022(1).
- Kline, T. L., Edwards, M. E., Fetzer, J., Gregory, A. V., Anaam, D., Metzger, A. J., and Erickson, B. J. (2021). Automatic Semantic Segmentation of Kidney Cysts in MR Images of Patients Affected by Autosomal-Dominant Polycystic Kidney Disease. *Abdominal Radiology*, 46:1053–1061.
- Koklu, M., Cinar, I., and Taspinar, Y. S. (2021). Classification Of Rice Varieties with Deep Learning Methods. *Computers and Electronics in Agriculture*, 187:106285.
- Koller, C., Kauermann, G., and Zhu, X. X. (2022). Going Beyond One-Hot Encoding in Classification: Can Human Uncertainty Improve Model Performance?
- Koundinya, S., Sharma, H., Sharma, M., Upadhyay, A., Manekar, R., Mukhopadhyay, R., Karmakar, A., and Chaudhury, S. (2018). 2D-3D CNN Based Architectures for Spectral Reconstruction from RGB Images. *IEEE Computer Society Conference on Computer Vision and Pattern Recognition Workshops*, pages 957–964.
- Kouretas, I. and Paliouras, V. (2019). Simplified Hardware Implementation of the Softmax Activation Function. In *8th International Conference on Modern Circuits and Systems Technologies, MOCAS 2019*, pages 1–4. IEEE.
- Kovesdy, C. P. (2022). Epidemiology of Chronic Kidney Disease: An Update 2022. *Kidney International Supplements Volume 12, Issue 1, April 2022, Pages 7-11*, 12(1):7–11.

- Krishnan, V. G., Saleem, P. A., Sathyamoorthy, K., Priya, K. H., and Kumar, T. K. (2023). Colorectal Cancer Prediction using ResNet-CNN Classification Method. *International Journal of Advances in Soft Computing and its Applications*, 15(2):54–66.
- Kull, M., Perello-Nieto, M., Kängsepp, M., Filho, T. S., Song, H., and Flach, P. (2019). Beyond Temperature Scaling: Obtaining Well-Calibrated Multiclass Probabilities with Dirichlet Calibration. In *Advances in Neural Information Processing Systems*.
- Kumar, A., Murthy, O. N., Shrish, Ghosal, P., Mukherjee, A., and Nandi, D. (2019). A Dense U-Net Architecture for Multiple Sclerosis Lesion Segmentation. *IEEE Region 10 Annual International Conference, Proceedings/TENCON*, pages 662–667.
- Kurnianto, E. A., Cholissodin, I., and Santoso, E. (2018). Klasifikasi Penderita Penyakit Ginjal Kronis Menggunakan Algoritme Support Vector Machine (SVM). *Jurnal Pengembangan Teknologi Informasi dan Ilmu Komputer (J-PTIHK) Universitas Brawijaya*, 2(12):6597–6602.
- Lee, S., Kang, Q., Madireddy, S., Balaprakash, P., Agrawal, A., Choudhary, A., Archibald, R., and Liao, W. K. (2019). Improving Scalability of Parallel CNN Training by Adjusting Mini-Batch Size at Run-Time. In *Proceedings - 2019 IEEE International Conference on Big Data, Big Data 2019*, pages 830–839. IEEE.
- Legay, C., Krasniqi, T., Bourdet, A., Bonny, O., and Bochud, M. (2022). Methods for The Dietary Assessment of Adult Kidney Stone Formers: A Scoping Review. *Journal of Nephrology*, 35(3):821–830.

- Li, W. and Huang, Y. (2020). A combined method of cross-correlation and PCA-based outlier algorithm for detecting structural damages on a jacket oil platform under random wave excitations. *Applied Ocean Research*, 102:102301.
- Liu, D., Jain, M., Dossou, B. F., Shen, Q., Lahlou, S., Goyal, A., Malkin, N., Emezue, C. C., Zhang, D., Hassen, N., Ji, X., Kawaguchi, K., and Bengio, Y. (2023). GFlowOut: Dropout with Generative Flow Networks. *Proceedings of Machine Learning Research*, pages 1–15.
- Liu, K., Kang, G., Zhang, N., and Hou, B. (2018). Breast Cancer Classification Based on Fully-Connected Layer First Convolutional Neural Networks. *IEEE Access*, 6:23722–23732.
- Lorentius, C. A., Adipranata, R., and Tjondrowiguno, A. (2019). Pengenalan Aksara Jawa dengan Menggunakan Metode Convolutional Neural Network. *Jurnal INFRA*, 7(1):1–7.
- Luyckx, V. A., Cherney, D. Z., and Bello, A. K. (2020). Preventing CKD in Developed Countries. *Kidney International Reports*, 5(3):263–277.
- Ma, M. (2020). Infrared Pedestrian Detection Algorithm Based on Multimedia Image Recombination and Matrix Restoration. *Multimedia Tools and Applications*, pages 9267–9282.
- Maheen, U., Malik, K. I., and Ali, G. (2021). Comparative Analysis of Deep Learning Algorithms for Classification of COVID-19 X-Ray Images. *International Journal of Computer Science Trends and Technology (IJCSST)*, 9(4):73–81.
- Maruyama, T., Hayashi, N., Sato, Y., Hyuga, S., Wakayama, Y., Watanabe, H.,

- Ogura, A., and Ogura, T. (2018). Comparison Of Medical Image Classification Accuracy Among Three Machine Learning Methods. *Journal of X-ray Science and Technology*, 26(6):885–893.
- MathWorks (2021). What is a Convolutional Neural Network?
- Matondang, Z. A., Damanik, R., and Silitonga, P. D. P. (2022). Image Detection Edge Image Using Canny Edge Algorithm. *Journal of Artificial Intelligence and Engineering Applications (JAIEA)*, 1(3):248–255.
- Maulana, M. A., Jamaludin, A., Solehudin, A., and Voutama, A. (2019). Sistem Pakar Diagnosis Penyakit Ginjal Menggunakan Metode Certainty Factor Berbasis Website. *INFOTECH Journal*, 9(4):431–441.
- Melbourne Radiology Clinic (2021). CT Scan of the Kidney - Diagnostic Imaging.
- Monaco, S., Malfatti, G., Culham, J. C., Cattaneo, L., and Turella, L. (2020). Decoding Motor Imagery and Action Planning in The Early Visual Cortex: Overlapping but Distinct Neural Mechanisms. *NeuroImage*, 218:116981.
- Mostafa, A. M., Kumar, S. A., Meraj, T., Rauf, H. T., Alnuaim, A. A., and Alkhayyal, M. A. (2022). Guava Disease Detection Using Deep Convolutional Neural Networks: A case Study of Guava Plants. *Applied Sciences (Switzerland)*, 12(239):1–19.
- Murdika, U., Alif, M., and Mulyani, Y. (2021). Identifikasi Kualitas Buah Tomat dengan Metode PCA (Principal Component Analysis) dan Backpropagation. *Electrician - Jurnal Rekayasa dan Teknologi Elektro*, 15(3):175–180.
- Naufal, M. H. (2023). *Pendeteksian Balon Ucapan Pada Komik Jepang (Manga)*

Dengan Run Length Smooth Dan EfficientNet-B6. Skripsi, Universitas Komputer Indonesia.

Naz, S., Bibi, K., and Ahmad, R. (2022). DeepSignature: Fine-Tuned Transfer Learning Based Signature Verification System. *Multimedia Tools and Applications*, 81:38113–38122.

Negi, A., Kumar, K., Chaudhari, N. S., Singh, N., and Chauhan, P. (2021). Predictive Analytics for Recognizing Human Activities Using Residual Network and Fine-Tuning. *Big Data Analytics*, 13147.

Novitasari, D. C., Wulandari, P., and Haq, D. Z. (2022). Cervical Cancer Diagnosis System using Convolutional Neural Network ResidualNet. *International Journal of Computing*, 21(1):61–68.

Novitasari, D. C. R., Hendradi, R., Caraka, R. E., Rachmawati, Y., Fanani, N. Z., Syarifudin, A., Toharudin, T., and Chen, R. C. (2020). Detection of COVID-19 Chest X-Ray Using Support Vector Machine and Convolutional Neural Network. *Communications in Mathematical Biology and Neuroscience*, 42:1–19.

Pahlevi, R. A. and Setiaji, B. (2023). Analysis of Application Haar Cascade Classifier and Local Binary Pattern Histogram Algorithm in Recognizing Faces With Real-Time Grayscale Images Using Opencv. *Jurnal Teknik Informatika (JUTIF)*, 4(1):179–186.

Pang, S., Qiao, S., Song, T., Zhao, J., and Zheng, P. (2019). An Improved Convolutional Network Architecture Based on Residual Modeling for Person Re-Identification in Edge Computing. *IEEE Access*, 7:106749–106760.

Peryanto, A., Yudhana, A., and Umar, R. (2020). Klasifikasi Citra Menggunakan

- Convolutional Neural Network dan K Fold Cross Validation. *Journal of Applied Informatics and Computing (JAIC)*, 4(1):45–51.
- Pithani, M. B., Sanyal, S., and Shukla, A. K. (2022). Bilinear and Bicubic Interpolations for Image Presentation of Mechanical Stress and Temperature Distribution. *Power Engineering and Engineering Thermophysics*, 1(1):8–18.
- Putra, T. I. Z. M., Suprpto, and Bukhori, A. F. (2022). Model Klasifikasi Berbasis Multiclass Classification dengan Kombinasi Indobert Embedding dan Long Short-Term Memory untuk Tweet Berbahasa Indonesia (Classification Model Based on Multiclass Classification with a Combination of Indobert Embedding and Long . *Jurnal Ilmu Siber dan Teknologi Digital (JISTED)*, 1(1):1–28.
- Qadir, A. M. and Abd, D. F. (2023). Kidney Diseases Classification using Hybrid Transfer-Learning DenseNet201-Based and Random Forest Classifier. *Kurdistan Journal of Applied Research*, (March):131–144.
- Quach, L. D., Quoc, N. P., Thi, N. H., Tran, D. C., and Hassan, M. F. (2020). Using SURF to Improve ResNet-50 Model for Poultry Disease Recognition Algorithm. *2020 International Conference on Computational Intelligence, ICCI 2020*, (October):317–321.
- Rachman, S. A., Bagaskara, D. C., Magdalena, R., and Sa'idah, S. (2023). Classification of Pneumonia Based on X-Ray Images with ResNet-50 Architecture. In *Proceeding of the 3rd International Conference on Electronics, Biomedical Engineering, and Health Informatics*.
- Rakamawati, J. V. N. (2021). *Klasifikasi Diabetic Retinopathy Berdasarkan Foto Fundus Menggunakan Convolutional Neural Network (CNN) Jenis Densenet*. PhD thesis, UIN Sunan Ampel Surabaya.

- Raman, S., Maskeliūnas, R., and Damaševičius, R. (2021). Markerless Dog Pose Recognition in The Wild using Resnet Deep Learning Model. *Computers*, 11(2):1–12.
- Razzak, I., Shoukat, G., Naz, S., and Khan, T. M. (2020). Skin Lesion Analysis Toward Accurate Detection of Melanoma using Multistage Fully Connected Residual Network. *Proceedings of the International Joint Conference on Neural Networks*.
- Reddy, P. C. S., Kumar, G. R., and Yadala, S. (2022). Image Recognition using Deep Residual Learning. *International Journal of Research in Engineering and Science (IJRES)*, 10(5):115–121.
- Rediger, C., Guerra, L. A., Keays, M. A., Wayne, C., Reddy, D., Ksara, S., and Leonard, M. P. (2019). Renal cyst evolution in childhood: a contemporary observational study. *Journal of Pediatric Urology*, 15(2):188.e1–188.e6.
- Safiri, S., Kolahi, A. A., Mansournia, M. A., Almasi-Hashiani, A., Ashrafi-Asgarabad, A., Sullman, M. J., Bettampadi, D., Qorbani, M., Moradi-Lakeh, M., Ardalan, M., Mokdad, A., and Fitzmaurice, C. (2020). The burden of kidney cancer and its attributable risk factors in 195 countries and territories, 1990–2017. *Scientific Reports*, 10(1):1–20.
- Salimi, A. A. (2023). *Simulasi Finite Element Analysis (FEA) dan Fabrikasi Implan untuk Perbandingan Kinerja Implan Kovensional dan Custom Cruciate Retaining (CR) untuk penderita Osteoarthritis Berdasarkan Rekonstruksi Data MRI/CT Scan Tulang Lutut*. PhD thesis, Universitas Islam Indonesia.
- Sanna, E., Loukogeorgakis, S., Prior, T., Derwig, I., Paramasivam, G., Choudhry,

- M., and Lees, C. (2019). Fetal abdominal cysts: Antenatal course and postnatal outcomes. *Journal of Perinatal Medicine*, 47(4):418–421.
- Sarwinda, D., Paradisa, R. H., Bustamam, A., and Anggia, P. (2021). Deep Learning in Image Classification using Residual Network (ResNet) Variants for Detection of Colorectal Cancer. *Procedia Computer Science*, 179:423–431.
- Sathesh, A. and Adam, E. E. B. (2021). Hybrid Parallel Image Processing Algorithm for Binary Images with Image Thinning Technique. *Journal of Artificial Intelligence and Capsule Networks*, pages 243–258.
- Seo, Y. and shik Shin, K. (2019). Hierarchical Convolutional Neural Networks for Fashion Image Classification. *Expert Systems with Applications*, 116:328–339.
- Sharma, H., Saraswat, M., Kumar, S., and Bansal, J. C. (2021). *Intelligent Learning for Computer Vision*, volume 61. Springer, Singapore.
- Shehab, L. H., Fahmy, O. M., Gasser, S. M., and El-Mahallawy, M. S. (2021). An Efficient Brain Tumor Image Segmentation Based on Deep Residual Networks (ResNets). *Journal of King Saud University - Engineering Sciences*, 33:404–412.
- Shi, J., Dang, J., Cui, M., Zuo, R., Shimizu, K., Tsunoda, A., and Suzuki, Y. (2021). Improvement of Damage Segmentation Based on Pixel-Level Data Balance Using VGG-Unet. *Applied Sciences (Switzerland)*, 11:1–17.
- Showkat, S. and Qureshi, S. (2022). Efficacy of Transfer Learning-based ResNet Models in Chest X-ray Image Classification for Detecting COVID-19 Pneumonia. *Chemometrics and Intelligent Laboratory Systems*, 224:104534.
- Sia, M. (2021). Renal pathology.

- Sihotang, J. H. (2019). Implementation of Gray Level Transformation Method for Sharping 2D Images. *Jurnal INFOKUM*, 8(1):15–19.
- Sindagi, V. A., Yasarla, R., and Patel, V. M. (2022). JHU-CROWD++: Large-Scale Crowd Counting Dataset and A Benchmark Method. *IEEE Transactions on Pattern Analysis and Machine Intelligence*, 44(5):2594–2609.
- Srivastava, N., Hinton, G., Krizhevsky, A., Sutskever, I., and Salakhutdinov, R. (2014). Dropout: A Simple Way to Prevent Neural Networks from Overfitting. *Journal of Machine Learning Research*, 15:1929–1958.
- Suharmanto (2021). Ordinal Regression Model to Predict Hypertension. *Indian Journal of Forensic Medicine and Toxicology*, 15(3):4185–4190.
- Suharmanto (2022). Pengetahuan Tentang Pencegahan Penyakit Kronis Berhubungan Dengan Kualitas Hidup Petani. *Jurnal Penelitian Perawat Profesional*, 4(3):1011–1016.
- Suherman, E., Hindarto, D., Makmur, A., and Santoso, H. (2023). Comparison of Convolutional Neural Network and Artificial Neural Network for Rice Detection. *Sinkron*, 8(1):247–255.
- Sujithra, B. S. and Jerome, S. A. (2023). Identification of Glaucoma in Fundus Images Utilizing Gray Wolf Optimization with Deep Convolutional Neural Network-based Resnet50 Model. *Multimedia Tools and Applications*, (0123456789).
- Sukegawa, S., Yoshii, K., Hara, T., Matsuyama, T., Yamashita, K., Nakano, K., Takabatake, K., Kawai, H., Nagatsuka, H., and Furuki, Y. (2021). Multi-Task

- Deep Learning Model for Classification of Dental Implant Brand and Treatment Stage Using Dental Panoramic Radiograph Images. *Biomolecules*, 11(815):1–14.
- Sulistiani, S. L. (2019). Analisis Maqashid Syariah Dalam Pengembangan Hukum Industri Halal Di Indonesia. *Law and Justice*, 3(2):91–97.
- Sumitha, S. and Gokila, S. (2023). Experimental Approach to Identify the Optimal Deep CNN Models to Early Detection of Glaucoma from Fundus CT-Scan Images. *2023 7th International Conference on Intelligent Computing and Control Systems (ICICCS)*, pages 511–515.
- Summers, C. and Dinneen, M. J. (2020). Four Things Everyone Should Know To Improve Batch Normalization. In *8th International Conference on Learning Representations, ICLR 2020*.
- Sun, J., Peng, L., Li, T., Adila, D., Zaiman, Z., Melton-Meaux, G. B., Ingraham, N. E., Murray, E., Boley, D., Switzer, S., Burns, J. L., Huang, K., Allen, T., Steenburg, S. D., Gichoya, J. W., Kummerfeld, E., and Tignanelli, C. J. (2022). Performance of a Chest Radiograph AI Diagnostic Tool for COVID-19: A Prospective Observational Study. *Radiology: Artificial Intelligence*, 4(4).
- Szyc, K. (2020). An Impact of Different Images Color Spaces on the Efficiency of Convolutional Neural Networks. In *Corporate Social Responsibility and Corporate Change*, pages 506–514.
- Tabrizchi, H., Javidi, M. M., and Amirzadeh, V. (2021). Estimates of Residential Building Energy Consumption using a Multi-Verse Optimizer-Based Support Vector Machine with K-Fold Cross-Validation. *Evolving Systems*, 12:755–767.
- Taufiq, Sari, A., and Jannah, R. (2021). Deteksi Rasa Kantuk pada Pengendara

- Kendaraan Bermotor Berbasis Pengolahan Citra Digital. *Jurnal Teknologi Terapan and Sains 4.0*, 2(1):83–92.
- Tekade, R. and Rajeswari, K. (2018). Lung Cancer Detection and Classification using Deep Learning. In *2018 Fourth International Conference on Computing Communication Control and Automation (ICCUBEA)*, pages 1–5.
- Thanh-Tung, H., Venkatesh, S., and Tran, T. (2019). Improving Generalization and Stability of Generative Adversarial Networks. *7th International Conference on Learning Representations, ICLR 2019*, pages 1–18.
- Tran, T. D., Kimura, M., and Nakashima, Y. (2020). Primary Visual Cortex Inspired Feature Extraction Hardware Model. In *2020 4th International Conference on Recent Advances in Signal Processing, Telecommunications and Computing, SigTelCom 2020*, pages 20–24.
- Trisiawan, I. K. and Yuliza, Y. (2022). Penerapan Multi-Label Image Classification Menggunakan Metode Convolutional Neural Network (CNN) Untuk Sortir Botol Minuman. *Jurnal Teknologi Elektro*, 13(1):48.
- Wan, S., Liang, Y., and Zhang, Y. (2018). Deep Convolutional Neural Networks for Diabetic Retinopathy Detection by Image Classification. *Computers and Electrical Engineering*, 72:274–282.
- Wang, Q., Ma, Y., Zhao, K., and Tian, Y. (2022). A Comprehensive Survey of Loss Functions in Machine Learning. *Annals of Data Science*, 9:187–212.
- Wang, W., Xu, Y., Gao, R., Lu, R., Han, K., Wu, G., and Tan, W. (2020). Detection of SARS-CoV-2 in Different Types of Clinical Specimens. *JAMA - Journal of the American Medical Association*, 323(11):1843–1844.

- Wen, L., Li, X., and Gao, L. (2020). A Transfer Convolutional Neural Network for Fault Diagnosis Based on ResNet-50. *Neural Computing and Applications*, 32:6111–6124.
- Weston, A. (2021). How to Segment a Pancreas CT.
- Widjaja, Y., Santoso, A. H., Wijaya, D. A., Satyanegara, W. G., Kurniawan, J., Herdiman, A., Hartono, V. A. B., Ranonto, S. V., and Lumintang, V. G. (2023). Peningkatan Kewaspadaan Masyarakat Terhadap Penyakit Ginjal Kronis dengan Edukasi Gaya Hidup dan Skrining Fungsi Ginjal. *Communnity Development Journal*, 4(6):12147–12153.
- Wildah, S. K., Agustiani, S., Mustopa, A., Wuryani, N., Nawawi, H. M., and Safitri, R. A. (2021). Pengenalan Wajah Menggunakan Pembelajaran Mesin Berdasarkan Ekstraksi Fitur Pada Gambar Wajah Berkualitas Rendah. *INFOTECH : Jurnal Informatika dan Teknologi*, 2(2):95–103.
- Wiliyanarti, P. F. and Muhith, A. (2019). Life Experience of Chronic Kidney Diseases Undergoing Hemodialysis Therapy. *Jurnal NurseLine*, 4(1):54–60.
- Winoto, A. A. and Roy, A. F. (2023). Model of Predicting the Rating of Bridge Conditions in Indonesia with Regression and K-Fold Cross Validation. *International Journal of Sustainable Construction Engineering and Technology*, 14(1):249–259.
- Wu, Y., Liu, L., Bae, J., Chow, K.-H., Iyengar, A., Pu, C., Wei, W., Yu, L., and Zhang, Q. (2019). Demystifying Learning Rate Policies for High Accuracy Training of Deep Neural Networks. *2019 IEEE International Conference on Big Data (Big Data)*, pages 1971–1980.

- Wulandari, I., Yasin, H., and Widiharih, T. (2020). Klasifikasi Citra Digital Bumbu Dan Rempah Dengan Algoritma Convolutional Neural Network (CNN). *Jurnal Gaussian*, 9(3):273–282.
- Xu, J., Zhang, Y., and Miao, D. (2020). Three-Way Confusion Matrix for Classification: A Measure Driven View. *Information Sciences*, 507:772–794.
- Yang, B., Zhang, Z., Yang, C. Q., Wang, Y., Orr, M. C., Wang, H., and Zhang, A. B. (2022). Identification of Species by Combining Molecular and Morphological Data Using Convolutional Neural Networks. *Systematic Biology*, 71(3):690–705.
- Yani, M., Irawan, B., and Setiningsih, C. (2019). Application of Transfer Learning Using Convolutional Neural Network Method for Early Detection of Terry’s Nail. In *Journal of Physics: Conference Series*, volume 1201, page 012052.
- Yildirim, K., Bozdog, P. G., Talo, M., Yildirim, O., Karabatak, M., and Acharya, U. R. (2021). Deep learning model for automated kidney stone detection using coronal CT images. *Computers in Biology and Medicine*, 135(April):104569.
- You, K., Long, M., Wang, J., and Jordan, M. I. (2019). How Does Learning Rate Decay Help Modern Neural Networks? *CoRR*.
- Zafar, A., Aamir, M., Nawi, N. M., Arshad, A., Riaz, S., Alruban, A., Dutta, A. K., and Almotairi, S. (2022). A Comparison of Pooling Methods for Convolutional Neural Networks. *Applied Sciences*, 12(17):1–21.
- Zan, B. W., Han, Z. H., Xu, C. Z., Liu, M. Q., and Wang, W. Z. (2022). High-Dimensional Aerodynamic Data Modeling Using a Machine Learning Method Based on a Convolutional Neural Network. *Advances in Aerodynamics*, 4(39):1–31.

Zebua, H. (2023). *Pembuatan Model Deteksi Gejala Awal Penyakit Mulut dan Kuku pada Sapi Berbasis Citra Menggunakan Metode Convolutional Neural Network*. PhD thesis, Universitas Jambi.

Zeng, G. (2020). On The Confusion Matrix in Credit Scoring and Its Analytical Properties. *Communications in Statistics - Theory and Methods*, 47(24):1–15.

Zhang, H., Zhang, L., and Jiang, Y. (2019). Overfitting and Underfitting Analysis for Deep Learning Based End-to-end Communication Systems. In *2019 11th International Conference on Wireless Communications and Signal Processing, WCSP 2019*, pages 1–6. IEEE.



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S U R A B A Y A