

**PENGEMBANGAN MODEL THERMAL RESIDUALNET-128
UNTUK DETEKSI OBJEK DALAM CITRA TERMAL
MENGUNAKAN PEMBELAJARAN RESIDUAL**



SKRIPSI

Diajukan untuk memenuhi sebagian syarat untuk memperoleh gelar Sarjana Teknik
pada Program Studi Sistem Telekomunikasi

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Pengembangan Model Thermal ResidualNet-128 untuk Deteksi Objek dalam Citra Termal menggunakan Pembelajaran Residual

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Sebuah skripsi yang diajukan untuk memenuhi salah satu syarat memperoleh gelar
Sarjana Teknik pada Program Studi Sistem Telekomunikasi

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PENGEMBANGAN MODEL THERMAL RESIDUALNET-128 UNTUK DETEKSI OBJEK DALAM CITRA TERMAL MENGGUNAKAN PEMBELAJARAN RESIDUAL

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ABSTRAK

Deteksi objek untuk kendaraan otonom menghadapi tantangan dalam kondisi pencahayaan rendah. Citra termal menawarkan solusi yang efektif, namun arsitektur *Convolutional Neural Network* (CNN) konvensional sering mengalami penurunan performa pada jaringan dalam. Penelitian ini bertujuan mengembangkan dan mengevaluasi model Thermal ResidualNet-128 (TRNet-128), arsitektur CNN dengan mekanisme pembelajaran residual untuk mendeteksi objek manusia dan pengendara sepeda motor. *Dataset* yang digunakan terdiri dari 1.008 gambar citra termal *grayscale*. Performa model TRNet-128 dievaluasi menggunakan metrik *accuracy*, *precision*, *recall*, *F1-score*, *confusion matrix*, dan *mean Average Precision* (mAP), serta dibandingkan dengan model CNN sederhana tanpa *residual block*. Hasil menunjukkan model TRNet-128, setelah augmentasi data untuk mengatasi *overfitting*, mencapai akurasi validasi 99,86% dan mAP 0,99977, jauh lebih tinggi dibandingkan model CNN sederhana yang hanya 0,99652. Kesimpulannya, integrasi *residual blocks* pada TRNet-128 secara efektif meningkatkan akurasi klasifikasi dan ketepatan deteksi objek dalam citra termal.

Kata Kunci: Deteksi Objek, Citra Termal, *Convolutional Neural Network*, Pembelajaran Residual, TRNet-128

ABSTRACT

Object detection for autonomous vehicles faces challenges in low-light conditions. Thermal imagery offers an effective solution; however, conventional Convolutional Neural Network (CNN) architectures often experience performance degradation in deep networks. This study aims to develop and evaluate the Thermal ResidualNet-128 (TRNet-128) model, a CNN architecture equipped with a residual learning mechanism to detect human and motorcycle rider objects. The dataset used consists of 1,008 grayscale thermal images. The performance of the TRNet-128 model is evaluated using metrics such as accuracy, precision, recall, F1-score, confusion matrix, and mean Average Precision (mAP), and compared to a simple CNN model without residual block. Results show that the TRNet-128 model, after data augmentation to mitigate overfitting, achieved a validation accuracy of 99.86% and an mAP of 0.99977, significantly outperforming the simple CNN model, which only reached 0.99652. In conclusion, the integration of residual blocks in the TRNet-128 architecture effectively enhances classification accuracy and object detection precision in thermal images.

Keywords: Object Detection, Thermal Imagery, Convolutional Neural Network, Residual Learning, TRNet-128.

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