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Thermal Image Enhancement using Generative Adversarial Network for Pedestrian



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Detection

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Introduction

Motivation:

- Infrared cameras often suffer from low-contrast, low-resolution and blurred details.
- These problems could limit the feasibility of different infrared imaging applications.

Objective: To enhance the visual quality of thermal images in order to improve the pedestrian detection performance.

State-of-the-art:

Traditional methods based on Histogram Equalization (HE) and Contrast Limited Adaptive(CLAHE) Deep Learning methods (VDSR, SRCNN, TEN, CDN-MRF)

KAIST dataset [3].

Total number of frames(RGB and thermal): 95k Training set : 7601 thermal images Testing set : 2252 thermal images





Fig.2: Examples of thermal images from KAIST datatset.

Experiments & Results

Dataset

1.Visual quality evaluation in terms of:

Adversarial Learning methods(SRGAN, DCGAN)

Proposed Thermal Image Enhancement Architecture

- New Thermal Enhancement-GAN (TE-GAN) architecture basically \bullet inspired from EnlightenGAN [1] and DnCNN[2].
- The proposed architecture is composed of two modules with a postprocessing step to cover different limitations of thermal images.
- Training TE-GAN architecture according to an overall loss function that combines perceptual, content, global and local losses



			HE	CLAHE	TE-GAN	
The Peak Signal to Noise	Ratio (PSNR)	PSNR	7.81	11.92	13.92	
The Structural Similarity In	dex Metric (SSIM)	SSIM	0.34	0.37	0.50	
	Ta	b.1: Compa	arison of to HE an	the TE-GAN d CLAHE.	architectu	ire
Original Image	HE	CLA	HE	Propos	ed TE-GAN	
	The second second				ST THE REAL	ROTIC T

Fig.3: Qualitative results of our proposed architecture TE-GAN compared to other enhancement methods.

2. Detection results with and without enhancement in terms:

	cond
he mean Average Precision (mAP)	
he Log Average Miss Rate (LAMR)	D

Testing conditions	Metric	Without enhancement	With enhancement
Day	mAP	0.61	0.63
	LAMR	0.41	0.40

Fig.1 :The proposed TE-GAN architecture composed of two modules:(a) contrast enhancement and (b) denoising.

1.Contrast Enhancement Module

To improve the contrast using U-NET generator, global and local discriminators.

Perceptual Loss

To compute the distance between the output image and the ground-truth based on high-level representations extracted from VGG pre-trained model.

2. Denoising Module

Night	mAP	0.66	0.73
	LAMR	0.26	0.20
AII	mAP	0.62	0.65
	LAMR	0.45	0.43

Tab.2: Comparison of the detection performance of YOLOv3 detector with and without TE-GAN enhancement.

Original Image

Detection without enhancement Detection with enhancement







Fig.4:Some results of pedestrian detection using YOLOv3 on thermal images from KAIST dataset with and without TE-GAN enhancement.

Conclusion & Future work

To remove the noise level by a CNN generator and a global discriminator. **Content Loss**

To minimize the low-level content errors between the noisy image and the denoised generated image using Pixel-wise MSE,.

3. Post-processing Module

To highlight the edges and decrease the visual blur effects by means of a convolutional edge enhancement filter

References

[1] Jiang, Yifan, et al. "Enlightengan: Deep light enhancement without paired supervision." arXiv preprint arXiv:1906.06972 (2019). [2] Zhang, Kai, et al. "Beyond a gaussian denoiser: Residual learning of deep cnn for image denoising." IEEE Transactions on Image Processing 26.7 (2017): 3142-3155. [3] Hwang, Soonmin, et al. "Multispectral pedestrian detection: Benchmark dataset and baseline." Proceedings of the IEEE conference on computer vision and pattern recognition. 2015.

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The detection results using YOLOv3 detector are improved by means of TE-GAN architecture with a significant margin.



Extension of the proposed TE-GAN architecture to incorporate a super-resolution module.



Extend the proposed approach to other applications such as people tracking and activity recognition.