

# **DEEP LEARNING-BASED OBJECT DETECTION IN LOW-LIGHTING CONDITIONS SYSTEM**

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# OBJECTIVES

- Introduction
- Literature survey
- Comparison with existing idea and Proposed idea
- Proposed system model
- Design diagram hardware & Software
- Flowchart
- Algorithm
- Results
- Discussion
- Reference

# INTRODUCTION

- Low light image enhancement is an important and challenging task in computer vision which aims to improve the visibility and restore corruptions hidden in the image.
- Low lighting conditions issues challenge not only human visual perception but also other vision tasks like nighttime object detection.
- And to correct this problem we propose our system which enhances and also detects the objects present in an image which has been corrupted by low lighting condition.



**LOW LIGHT IMAGE**

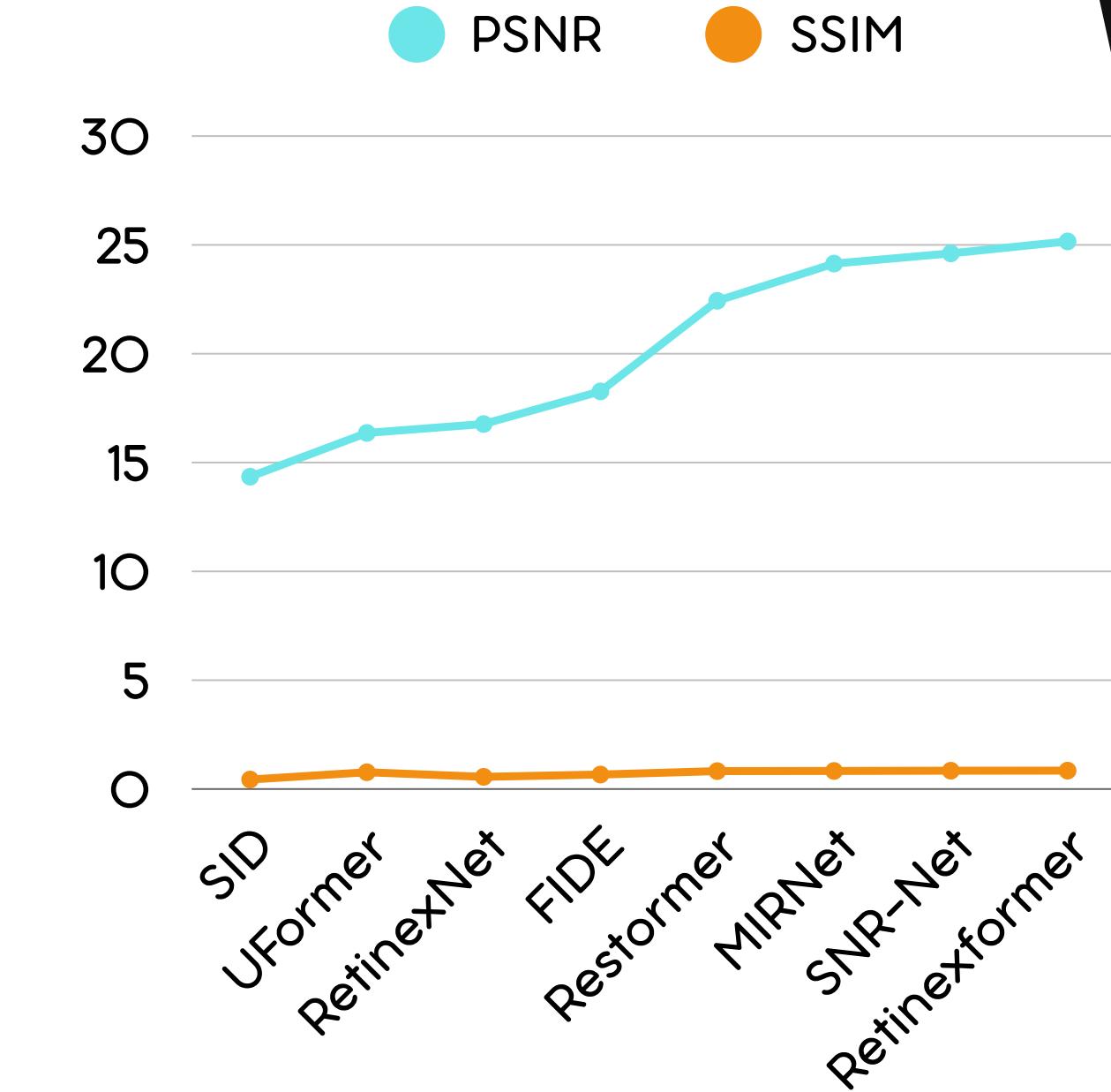


**ENHANCED IMAGE**

# LITERATURE SURVEY

## Image enhancement:-

- For image enhancement, histogram equalization is the most commonly used approach. However, it frequently produces under or over-enhanced images.
- Recently several deep learning models have applied retinex thoery [1] for image enhancement which have achieved state-of-the-art results.
- Also, CNNs have been successfully applied to low-light, image enhancement problems [2]. Notable works employ encoder-decoder networks [3,4], and GANs[5,6].



# LITERATURE SURVEY

## Object detection:-

- Key methods like the Viola-Jones framework (face detection)[7] and HOG (Histogram of Oriented Gradients) introduced by McConnell (1986) laid the foundation.
- Viola-Jones used Integral Images, Adaboost, and a Classifier Cascade, while HOG extracts pixel-based histograms as features, classified effectively using SVM by Dalal and Triggs[8].
- Overfeat (2013) utilized spatial convolutional features for detection. Due to the complexity and resource demands of two-stage methods, researchers now focus on single-stage detectors like YOLO.

# COMPARISON WITH EXISTING METHODS



a) Input



b) LIME



c) Retinex Net



d) EnlightenGAN



e) DRBN



f) Kind++



g) Kind



h) Zero-DCE



i) Retinex former



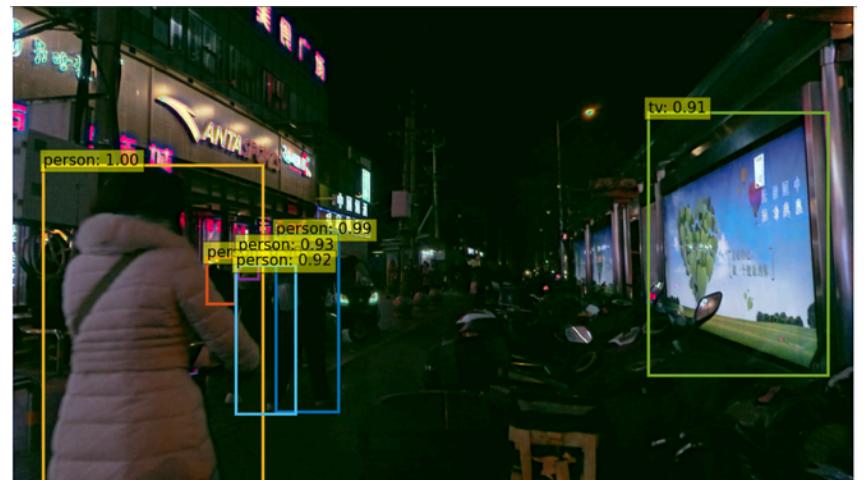
j) Reference

Figure: Visual comparison with state-of-the-art low-light image enhancement methods on LOL dataset.

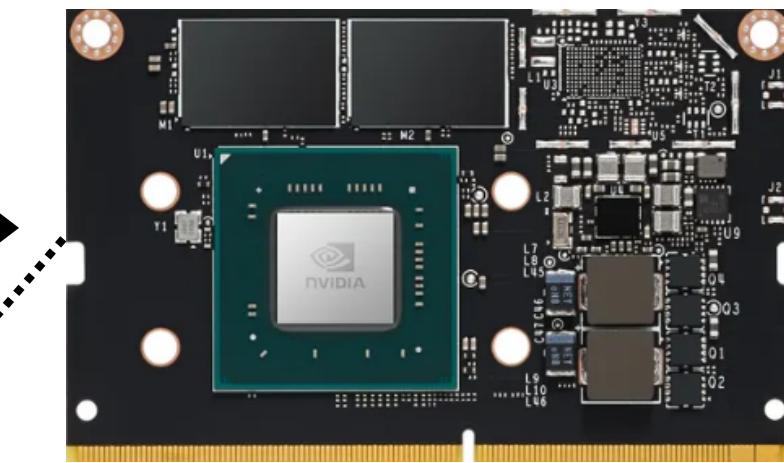
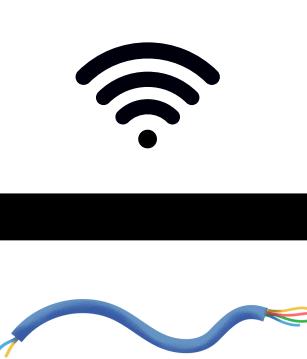
# DESIGN FLOW



LOW LIGHT INPUT IMAGE



OUTPUT DETECTED  
OBJECTS



NVIDIA JETSON  
NANO BOARD

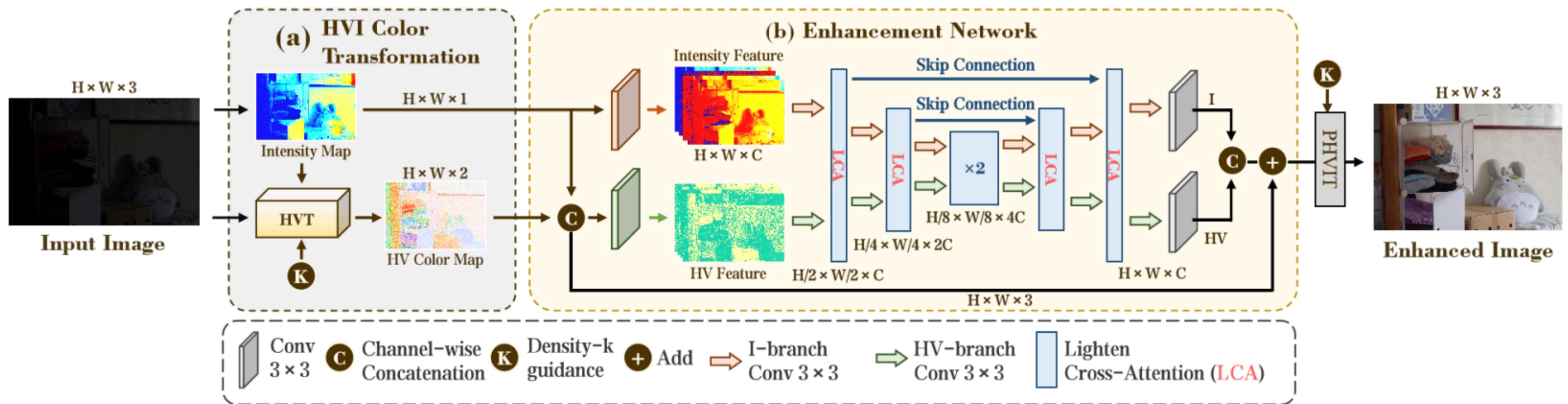
DEEP LEARNING  
BASED MODULES

Object Detector  
Model

Image Enhancer  
Model

INPUT  
LOW LIGHT

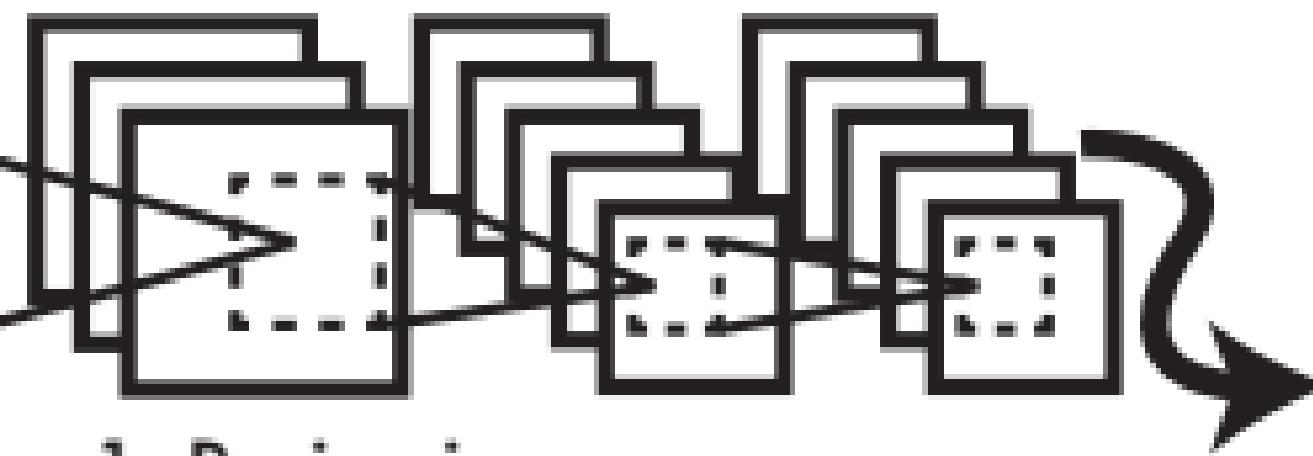
# IMAGE ENHANCEMENT MODEL



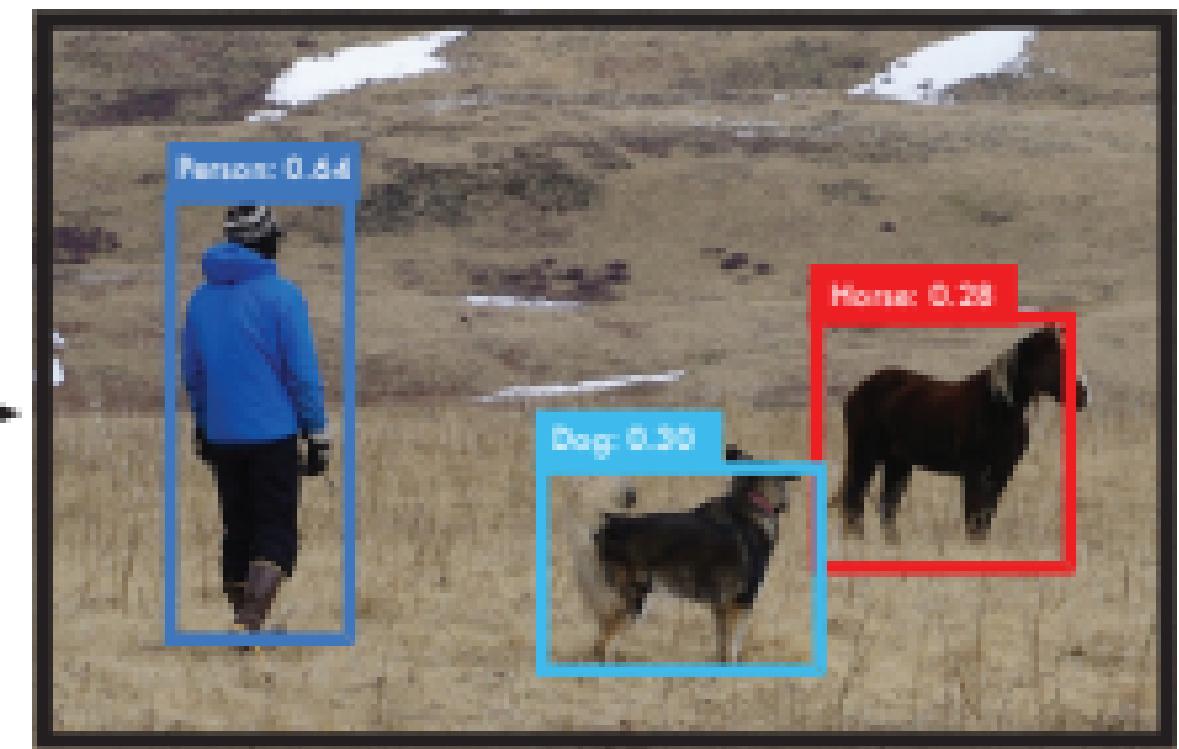
# OBJECT DETECTION MODEL



INPUT ENHANCED IMAGE



1. Resize image.
2. Run convolutional network.
3. Non-max suppression.



OUTPUT DETECTED OBJECTS

# RESULTS



FIG. ENHANCEMENT ON EXDARK DATASET



FIG. LOW LIGHT ENHANCEMENT WITH REAL TIME OBJECT DETECTION

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# THANK YOU