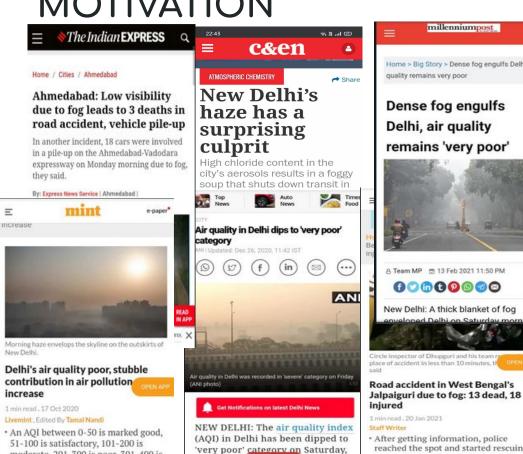
# Image Dehazing

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### MOTIVATION





reached the spot and started rescuing

the victims

according OPEN IN APP Air Quality





Air traffic affected due to poor visibility at Srinagar airport, 19 flights cancelled



Srinagar, Feb 3 (UNI) Air traffic was affected at Srinagar International airport, where 19

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incoming and outgoing flights were on Wednesday cancelled due to fresh snowfall and poor visibility.

'Four early morning incoming and outgoing flights were on Wednesday cancelled due to fresh snowfall and poor visibility at Srinagar airport,' officials said.

Tags: #Air Please log in to get detailed story. traffic affected due to poor visibility at Srinagar airport#19 flights cancelled

More News

**Punjab LB polls:** 

Livemint . Edited By Tamal Nandi

 An AOI between 0-50 is marked good, 51-100 is satisfactory, 101-200 is moderate, 201-300 is poor, 301-400 is very poor and 401-500 is considered

The vehicles which had a head-on collision near Thapal Chavadi in Ramanathapuram district on Sunday

#### Introduction

Image quality is of prime importance for many applications, but lot of factors contribute to make a image among which one of the keys is contrast. The project aims to understand the process of hazing through atmosphere scattering model, study dehazing algorithms, evaluate performance of different algorithms and propose an original architecture which would provide good and fast performance outdoors.

The are many approaches taken to solve the image hazing problem, such as polarization, contrast enhancement, Dark channel prior etc., but our focus will be to solve the problem using deep learning method.



Tel aviv univ.

#### Problem definition

- → To Combat the problem of loss of contrast and noise in images
- → Despite availability of many Dehazing algorithms the is no common comparison on performance on real haze images.
- → Integrate the model in commonly used machine vision pipeline to provide better performance, so as to combat the decision making dilemma faced by many SOTA machine vision problems.

RANK	MODEL	PSNR 1	SSIM
1	FFA-Net	35.77	0.9846
2	GCANet	30.23	0.98
3	GMAN	20.53	0.8081
4	Deep DCP	19.25	0.832

paperswithcode

# Literature review

## Datasets

### [1] RESIDE

- REalistic Single Image DEhazing
- Aimed at evaluating performance of single image Dehazing algorithms.
- The RESIDE:V0 provides about 313,000 synthetic outdoor hazy images
- Presents innovative set of evaluation techniques besides PSNR and SSIM, also employs no reference metrics such as BLIINDS-II and human subjective score for comparison (clearness & authenticity) fitted on Bradley-terry model.
- Provides a benchmark for single image dehazing
- Presents comparison and limitations of many SOTA, such as DehazeNet, AOD-Net, DCP etc

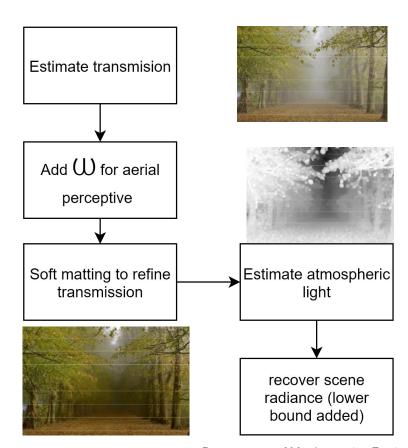
#### [6] O-haze

- Outdoor scenes database (named O-HAZE) composed of pairs of real hazy and corresponding haze-free images.
- O-HAZE is used to compare a representative set of state-of-the-art dehazing techniques, using traditional image quality metrics such as PSNR, SSIM and CIEDE2000.
- The few datasets that are currently considered, both for assessment and training of learning-based dehazing techniques, exclusively rely on synthetic hazy images.
- This reveals the limitations of current techniques, and questions some of their underlying assumptions.

# Models

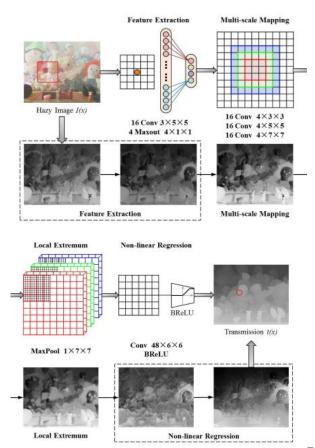
## [9] Dark Channel Prior

- DCP is a kind of statistic for haze free images, based on assumption that local patches in haze free images contain some pixels which have very low intensities in at least one color channel.
- Uses the atmosphere scattering model as a starting point I(x)=J(x)t(x)+A(1-t(x));  $t(x)=e^{-\beta d(x)}$
- DCP is based on digital imaging principle so it is simple and low cost approach.
- Has shortcomings such as colour distortion, detail loss and computational complexity

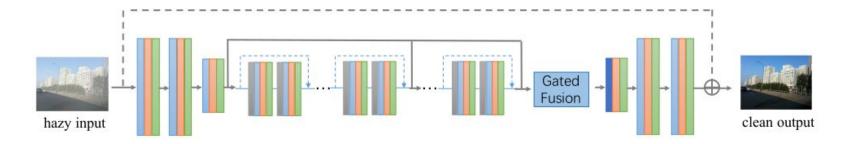


## [10] Dehaze Net

- DehazeNet adopts Convolutional Neural Networks
   (CNN) based deep architecture, whose layers are specially designed to embody the established assumptions/priors in image dehazing.
- Layers of Maxout units are used for feature extraction,
   which can generate almost all haze-relevant features.
- Uses a novel nonlinear activation function in DehazeNet, called Bilateral Rectified Linear Unit (BReLU), which is able to improve the quality of recovered haze-free image.
- Drawback dehazing sky region and the scene objects which are inherently similar to the atmospheric light (such as the fair-skinned complexion)



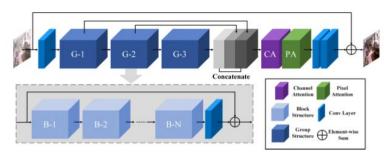
## [3] Gated Context Aggregation Network

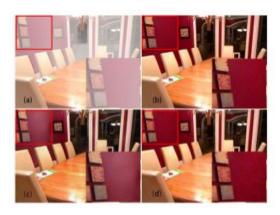


- Adopts smoothed dilated convolution to avoid loss of spatial resolution in subsampling
- Uses a gated fusion sub-network to fuse high level semantic feature maps at all scales
- Mean square error loss, adam optimizer, trained for 100 epochs on RESIDE dataset
- Obtains better performance in terms of PSNR & SSIM when compared to DCP, AOD-Net, Dehazenet.

## [4] Feature Fusion Attention Network

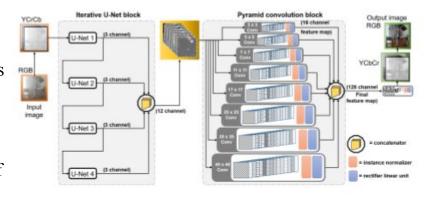
- A novel Feature Attention (FA) module combines
   Channel Attention with Pixel Attention mechanism.
- Giving more weight to important features. This structure can also retain the information of shallow layers and pass it into deep layers.
- Uses MSE loss function and RESIDE dataset.
- FFA network has a powerful advantage in the restoration of image detail and color fidelity.
- Shortcoming Complexity of network and cannot be used for real time application.





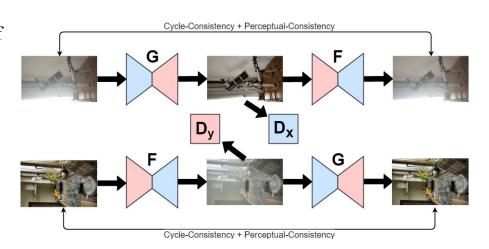
## [6] Back Projected Pyramid Network

- A novel technique named pyramid convolution is introduced for dehazing to obtain spatial features of multiple scales structural information.
- UNet block for image dehazing tasks to make the generator learn different and complex features of haze without the loss of local and global structural information.
- Extensive experimentation is done on four contemporary challenging datasets, namely I-Haze and O-Haze datasets of NTIRE 2018 challenge, Dense-haze dataset
- Uses a combination of MSE (L2 loss), adversarial loss Ladv, content loss Lcon, and structural similarity loss LSSIM.
- Drawback color preservation and seamless color cast



## [5] Cyclic gan

- CycleGAN combines cycle-consistency and perceptual losses in order to improve the quality of textural information recovery and generate visually better haze-free images.
- After obtaining low-resolution outputs from the network, we utilize the Laplacian pyramid to upscale the output images to the original resolution unlike other models which display the output in lower resolution.
- Trained with NYU-Depth, I-HAZE, and O-HAZE datasets



# Metrics

#### Need for different metrics

Traditional quality estimation metrics are:

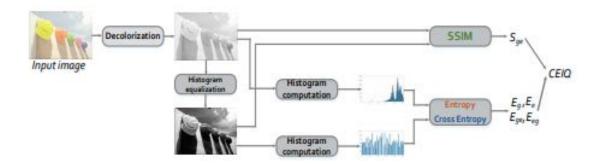
- PSNR- peak signal to noise ratio
- SSIM- structural similarity index measure

Both these need reference image and fail to provide the specificity required for dehazing problems

- The training of deep learning models generally requires ground truth images but the during prediction process i.e. evaluating a new image there is absence of ground truth image.
- NR IQA help in quantifying image quality when there is no reference to compare.

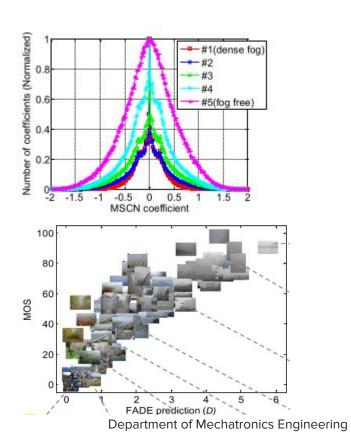
## [11] Contrast Distorted Image Quality

- A simple but effective metric for predicting quality of contrast-altered images based on the fact that a high-contrast image is often more similar to its contrast enhanced image.
- Experiments on four publicly available databases validate the superiority and efficiency of the proposed technique.
- Uses 5 features and learn a regression module to quantize the image quality structural-similarity index (SSIM), histogram based entropy and cross entropy for the original image and the enhanced one respectively.



## [2] Fog Aware Density evaluator

- Predicts visibility of a foggy scene from a single image without reference
- Based on NSS Model
- Fog Aware features such as variance, sharpness, colorfulness etc derived from a local patch.
- Features are fitted to a MVG, level of fog is determined by the Mahalanobis-like distance between hazy image and MVG model extracted from a corpus of 500 fog free images.
- FADE is compared to human subjective study



### Objectives

- Study literature on SOTA dehazing algorithms (7 Feb 10 Feb)
- Develop different SOTA models on tensorflow and pytorch (11 Feb 20 Feb)
- Quantize the quality of an image with reference and non-referential image. (21 Feb 24 Feb)
- Performance comparison of common metrics (25 Feb 28 Feb)
- Compare various models and study their shortcomings. (1Mar 5 Mar)
- Develop a novel model (5 Mar 20 Mar)
- Learning various preprocessing techniques and hyperparameter tuning to improve the overall model performance. (21 Mar - 31 Mar )
- Iterate over different model blocks to improve performance and justify need for each block. (1 Apr - 15 Apr)

## Methodology

- The current standard for image dehazing is studied and the algorithms are implemented using Python programming language, using deep learning frameworks such as TensorFlow and PyTorch.
- The algorithms are modified so that they output a common metric such as FADE (Fog Aware Density Evaluator) and compared.
- The learning from the present SOTA algorithms are put in developing a new architecture.
- The new model is trained on datasets such as RESIDE and O-Haze

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# Thank you