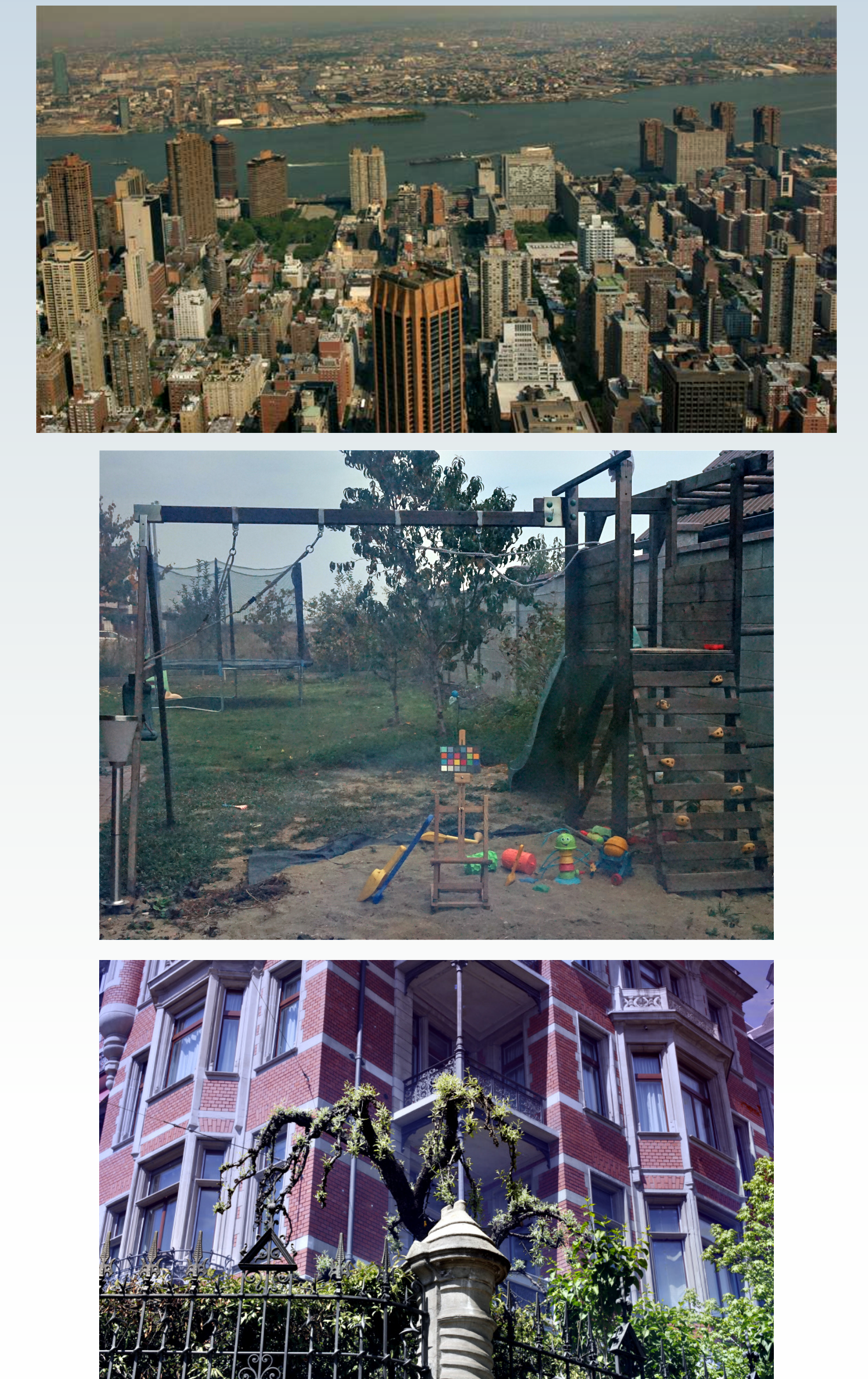
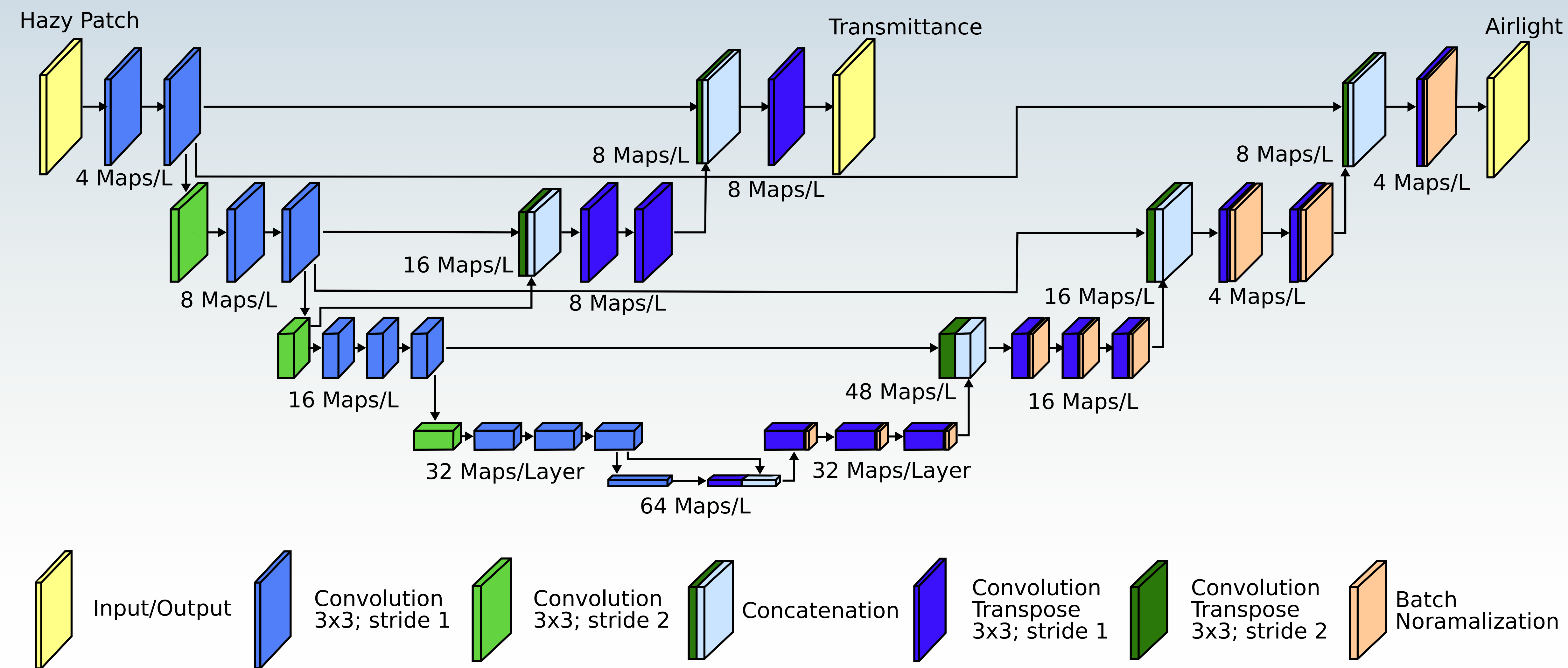
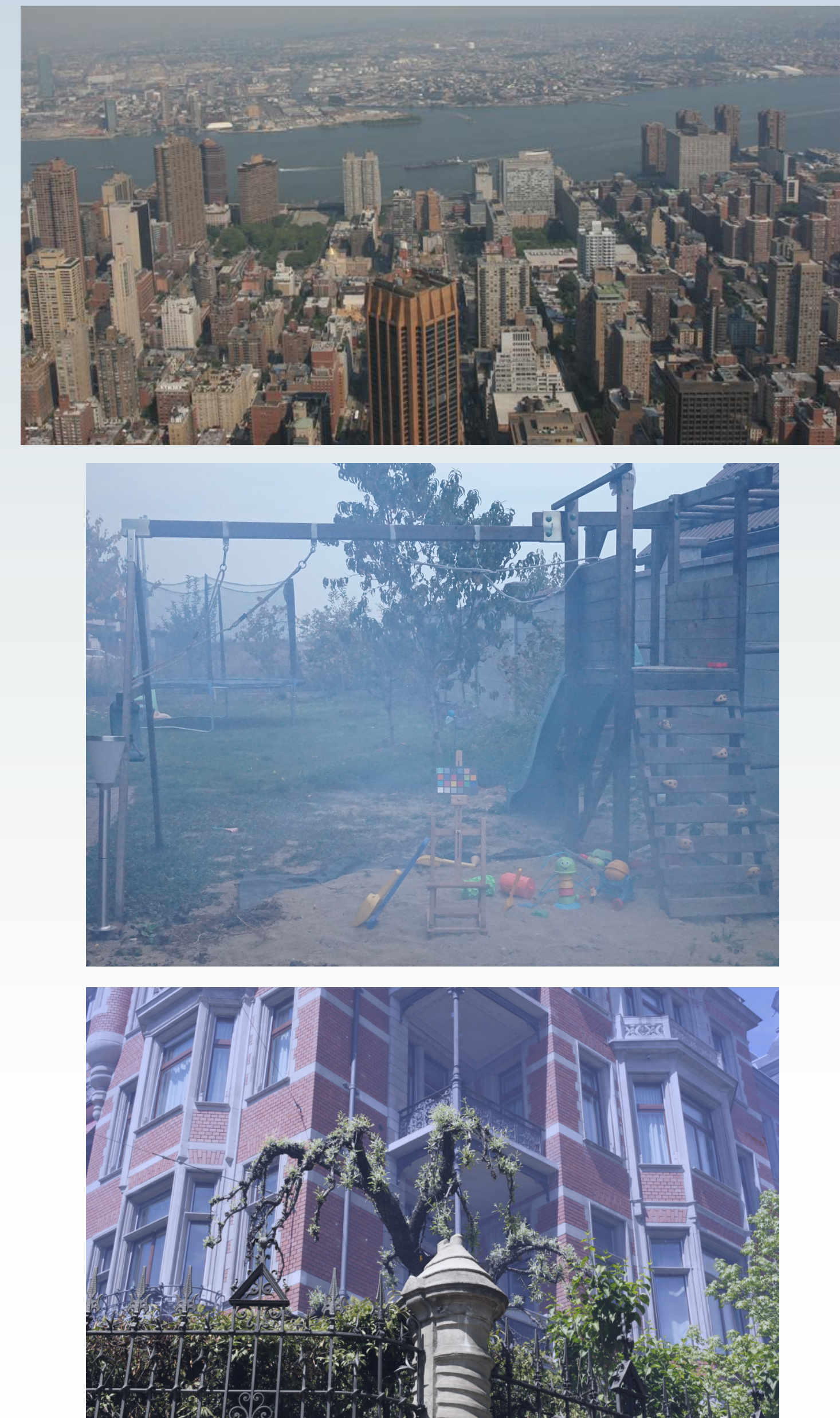


# Image Dehazing by Joint Estimation of Transmittance and Airlight using Bi-Directional Consistency Loss Minimized FCN

Ranjan Mondal, Sanchayan Santra, Bhabatosh Chanda

Electronics and Communication Sciences Unit, Indian Statistical Institute, Kolkata, India



## Image Formation Equation

Usual Model

$$I(\mathbf{x}) = J(\mathbf{x})t(\mathbf{x}) + (1 - t(\mathbf{x}))A$$

Relaxed to

$$I(\mathbf{x}) = J(\mathbf{x})t(\mathbf{x}) + (1 - t(\mathbf{x}))A(\mathbf{x}), \\ = J(\mathbf{x})t(\mathbf{x}) + K(\mathbf{x}).$$

## Bidirectional Consistency Loss

$$L = \frac{1}{N} \sum_{\mathbf{x}} (L_1(\mathbf{x}) + L_2(\mathbf{x}))$$

$$L_1(\mathbf{x}) = |I(\mathbf{x}) - J(\mathbf{x})t'(\mathbf{x}) - K'(\mathbf{x})| \rightarrow \text{The error of generating hazy image from the clear image.}$$

$$L_2(\mathbf{x}) = \left| J(\mathbf{x}) - \frac{I(\mathbf{x}) - K'(\mathbf{x})}{\max\{t'(\mathbf{x}), \epsilon\}} \right| \rightarrow \text{The error of getting clear image by dehazing the input.}$$

## Training Data Generation

- We extract patch hazy and haze free patch pairs from the training data in multiple levels.
- We start with a patch of size  $P \times P$ , where  $P = \min\{\text{Height}, \text{Width}\}$ .
- In the next level, we extract patches of size  $\frac{P}{2} \times \frac{P}{2}$  and  $\frac{P}{4} \times \frac{P}{4}$  in the next one. This halving process is repeated until the patch size falls below  $128 \times 28$ .
- All the extracted patches are resized to  $128 \times 128$  before they are used for training.

## $t(\mathbf{x})$ and $K(\mathbf{x})$ estimation

- Due to resource constrains, we first downscale the image.
- Estimate the  $t_i(\mathbf{x})$  and  $K_i(\mathbf{x})$  at three levels with patch sizes  $256 \times 256$ ,  $384 \times 384$  and  $512 \times 512$ .
- Feed the patches to the network to get  $t(\mathbf{x})$  and  $K(\mathbf{x})$  maps for the patches.
- In each level, aggregate the patches by averaging to get full size  $t(\mathbf{x})$  and  $K(\mathbf{x})$ -maps.

## Aggregation of $t(\mathbf{x})$ and $K(\mathbf{x})$

$$t(\mathbf{x}) = \frac{\sum_{i=1}^l w_i^{(t)} t_i(\mathbf{x})}{\sum_{i=1}^l w_i^{(t)}} \\ K(\mathbf{x}) = \frac{\sum_{i=1}^l w_i^{(K)} K_i(\mathbf{x})}{\sum_{i=1}^l w_i^{(K)}}$$

## Regularization using Guided Filter

- Due to the patch based processing, the  $t(\mathbf{x})$  and  $K(\mathbf{x})$  maps that we obtain, contain halos at the border of the patches.
- For this reason, we used Guided Filter[1] to smooth the maps.

[1] K. He, J. Sun, and X. Tang. Guided Image Filtering. IEEE Trans.on Pattern Anal. and Mach. Intell., 2013.

## Recovery of haze-free image

- The smooth transmittance map and airlight map is resized back to the original image size i.e.  $H \times W$ .
- The dehazed image is obtained as follows,
 
$$J'(\mathbf{x}) = \frac{I(\mathbf{x}) - K(\mathbf{x})}{\max\{t(\mathbf{x}), \epsilon\}}$$
- The output is clipped between 0 and 1 so that the output stays within the valid image intensity range.