

**Proper guidance image generation  
based on saliency factor  
for better transmission refinement  
in image dehazing**

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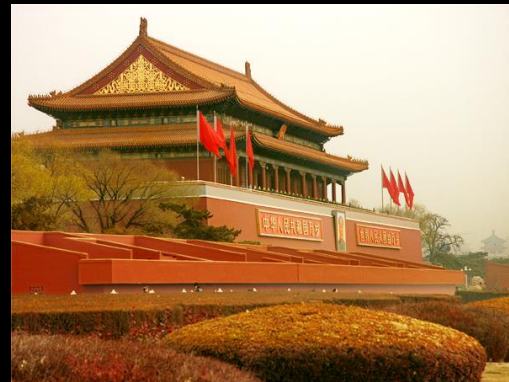
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# Introduction

## Image Dehazing:

- Restore haze-free images from corresponding hazy images



- The first step before other higher computer vision tasks  
Object detection, Pedestrian detection, OCR...

# Introduction

## Model of hazy image formation

Airlight (eg. (0.9, 0.85, 0.88))

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

Hazy image



Haze-free image



transmission



$$t(\mathbf{x}) = e^{-\beta d(\mathbf{x})}$$

# Introduction

Model of hazy image formation

$$I(x) = J(x) * t(x) + A(1 - t(x))$$

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

A: Global airlight, relatively easy (Eg. largest luminance in each channel)

t(x): Pixel mapping, hard to estimate ✓

# Introduction

Existing Method:

- Using addition information (multiple image/depth... )
- Single image haze removal ✓

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

underdetermined equation

Need prior or supposition

# Introduction

Single image haze removal:

- Dark Channel Prior [He et al. CVPR'08]
- Color Attenuation Prior [Zhu et al. TIP' 15, 24(11)]
- Maximum Reflectance Prior [Zhang CVPR'17]

region / neighbor  
based prior

Eg. Dark channel Prior

$$\min_{x \in \Omega(y)} \left( \min_{c \in \{r, g, b\}} J^c(y) \right) \rightarrow 0$$

$$\tilde{t}(x) = 1 - \min_{x \in \Omega(y)} \left( \min_{c \in \{r, g, b\}} \frac{I^c(y)}{A^c} \right)$$



# Introduction

Single image haze removal – learning based method:

- Random forest [Tang et al. CVPR'14]
- DehazeNet [Cai et al. TIP'16 25(11)]



Local maximum/minimum



blocky artifacts

How to solve blocky artifacts?

Guided image filtering

a filter transferring the structure of guidance image to filtering output

# Introduction

Traditional usage of guided image filtering in image dehazing:



Input p  
Rough estimated  $t(x)$



Guide I  
Input hazy image



Output q  
Refined transmission  $t(x)$

has exactly the same structures and details

Problem: Contrary to  $t(x) = e^{-\beta d(x)}$



# Introduction

Traditional usage of guided image filtering in image dehazing:

- Refined transmission and input hazy image have exactly the same structures and details
- Problem: Contrary to  $t(\mathbf{x}) = e^{-\beta d(\mathbf{x})}$

Natural idea: better guidance image

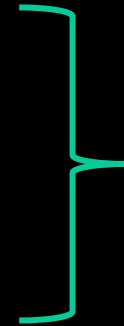
texture details  
are blurred

depth change regions  
are remained

# Methodology

texture details  
are blurred

depth change regions  
are remained



Capture depth change region

# Methodology

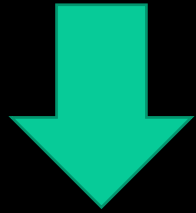


Feature of hazy image:

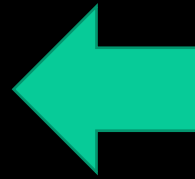
- **Neighboring pixels** at same depth
  - Disparity are degraded by haze
  - similarity
  - Majority
- **Neighboring pixels** at different depth
  - Disparity are enhanced by haze
  - discrepancy
  - Rare

# Methodology

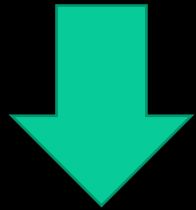
Depth change region



Rarely appeared pixel pairs



co-occurrence histogram  
based saliency detection



Higher saliency value

# Methodology

co-occurrence  
histogram based  
saliency detection

Rarely appeared pixel pairs

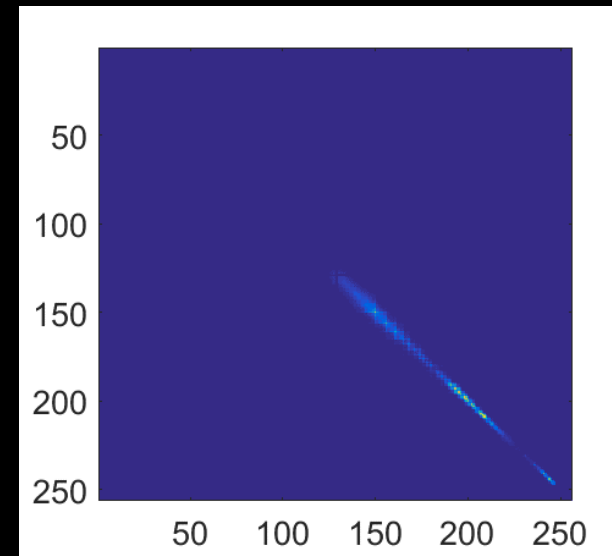
Higher saliency value

Input single channel image:

$$I_m = \min_{c \in \{r, g, b\}} (I)$$

1. co-occurrence histogram COH

$$\text{COH} = [\text{coh}(m, n)]$$



# Methodology

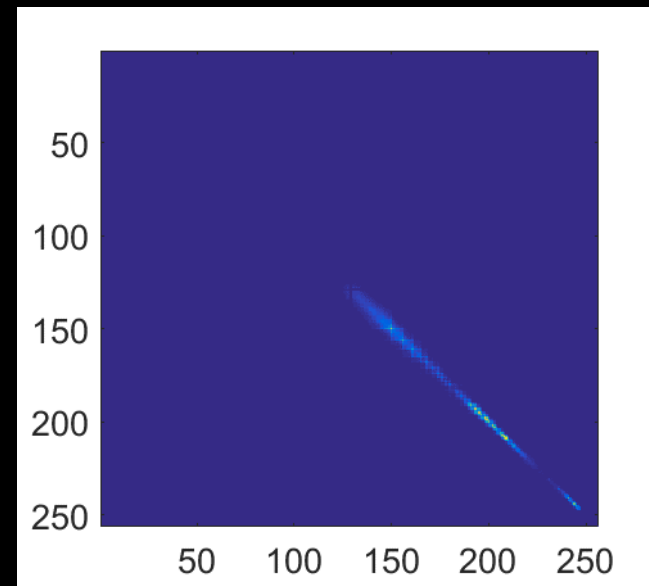
co-occurrence  
histogram based  
saliency detection

Rarely appeared pixel pairs

Higher saliency value

2. normalized co-occurrence  
histogram NCOH

$$\text{NCOH} = \frac{\text{COH}}{\max(\text{COH})}$$



# Methodology

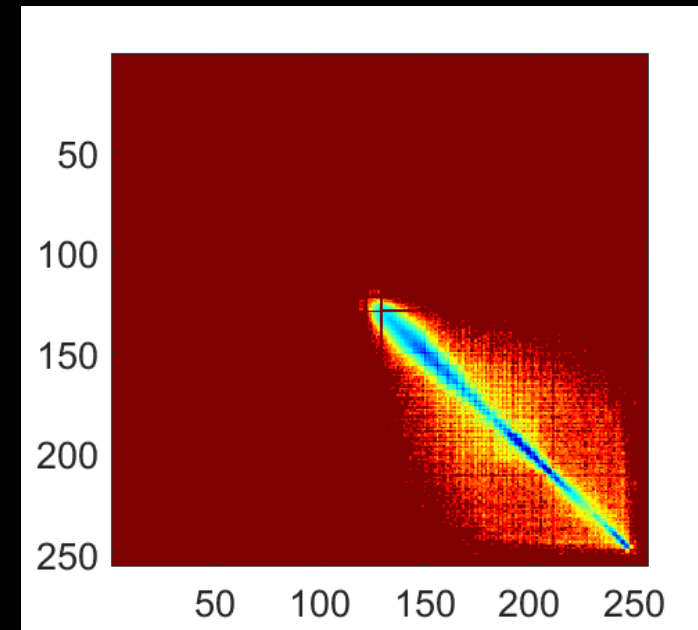
co-occurrence  
histogram based  
saliency detection

Rarely appeared pixel pairs

Higher saliency

3. saliency value  
of each intensity pair

$$s_p(m, n) = -\ln(\text{NCOH})$$



# Methodology

co-occurrence  
histogram based  
saliency detection

Rarely appeared pixel pairs

Higher saliency

3. saliency value  
of each intensity pair

$$s_p(m, n) = -\ln(\text{NCOH})$$

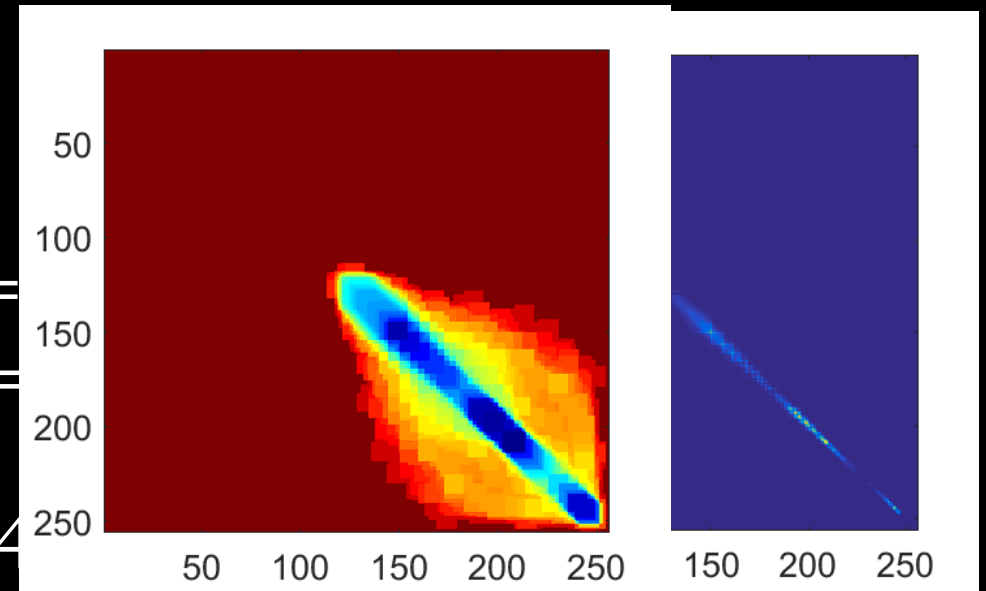
$$s(m, n) = \min_{(x, y) \in \Omega} s_p(m, n)$$

COH  
Suppose:

$$(48, 102) =$$

$$(49, 100) =$$

$$(47, 101) (4$$





# Methodology

co-occurrence  
histogram based  
saliency detection

Rarely appeared pixel pairs

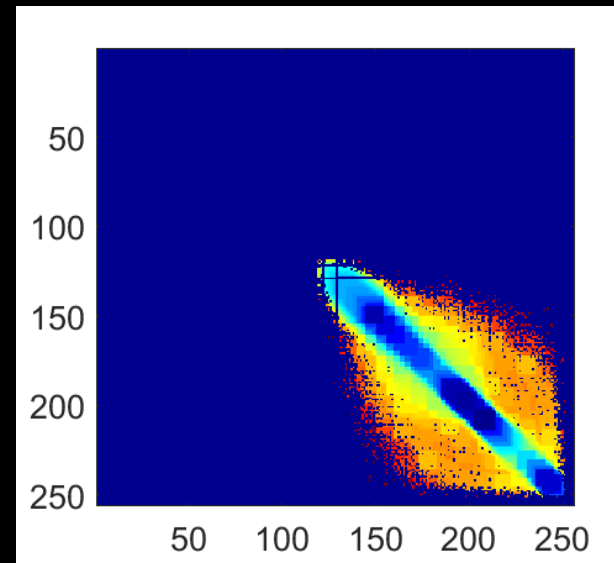
Higher saliency

3. saliency value  
of each intensity pair

$$s_p(m, n) = -\ln(\text{NCOH})$$

$$s(m, n) = \min_{(x, y) \in \Omega} s_p(m, n)$$

$$s(\text{COH} = 0) = 0$$



# Methodology

co-occurrence  
histogram based  
saliency detection

4. saliency map  
of each intensity pair

$$sa(i, j) = \sum_{i'=i-r}^{i+r} \sum_{j'=j-r}^{j+r} s(\text{Im}(i, j), \text{Im}(i', j'))$$

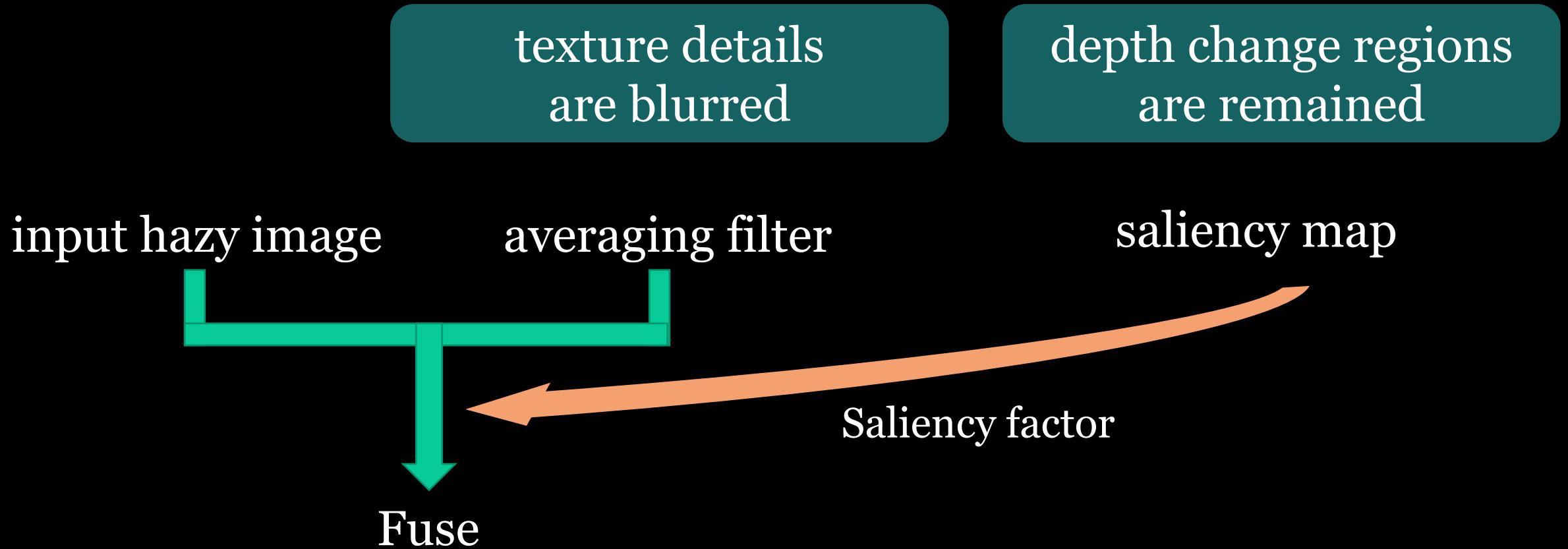
Rarely appeared pixel pairs

Higher saliency



# Methodology

Proper guidance image generation



# Methodology

## Saliency based guidance image generation



only marks pixels on the edge  
Not include pixels around them

$$d(x,y) = \frac{E(x,y)}{\sigma}$$

$$F(x) = \max_{y \in \Psi(x)} \frac{sa(y)}{d(x,y) + 1}$$

$$G = F * G_1 + (1 - F)G_2$$

# Methodology



# Methodology



# Result



$e = 1.2548$     $r = 1.0960$



$e = 1.5978$     $r = 1.5588$

# Result



$e = 0.0708$   $r = 1.3000$



$e = 0.0741$   $r = 1.5084$



# Result



$e = 0.4310$     $r = 1.4301$



$e = 0.4890$     $r = 1.9263$

# Result



$$e = 0.2520 \quad r = 1.2365$$



$$e = 0.3178 \quad r = 1.4736$$

**Thanks!**