

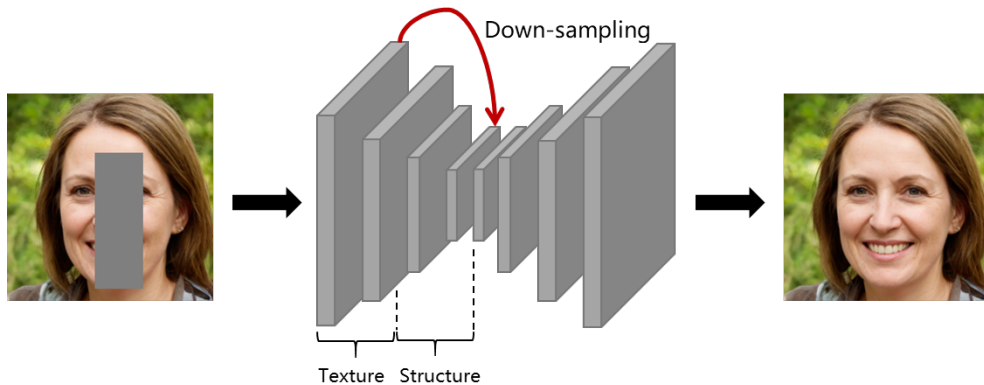
# Parallel Multi-Resolution Fusion Network for Image Inpainting

Wentao Wang, Jianfu Zhang, Li Niu, Haoyu Ling, Xue Yang, Liqing Zhang

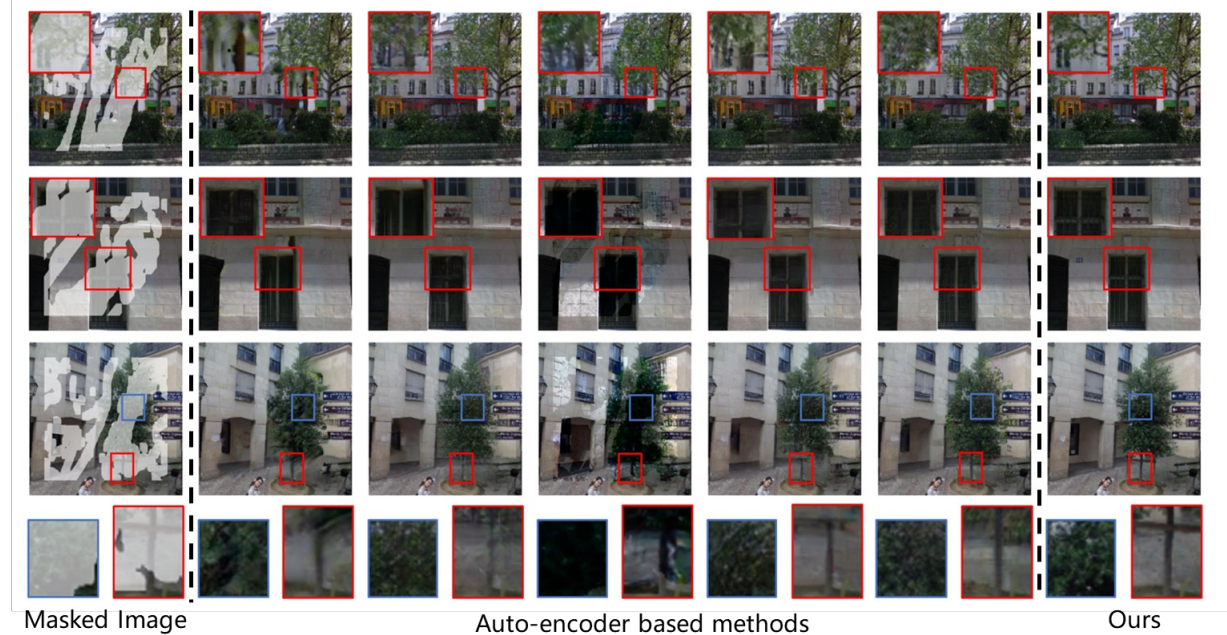
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# Challenges

- **Issue 1.** Most deep image inpainting methods are based on auto-encoder architecture, in which the spatial details of images will be lost in the down-sampling process.



Auto-encoder based inpainting methods



Qualitative comparison on Paris Street View

- **Issue 2.** Texture information and structure information can not be well integrated into a serial inpainting network like auto-encoder.

# Our Solutions

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## ◆ For **Issue 1**.

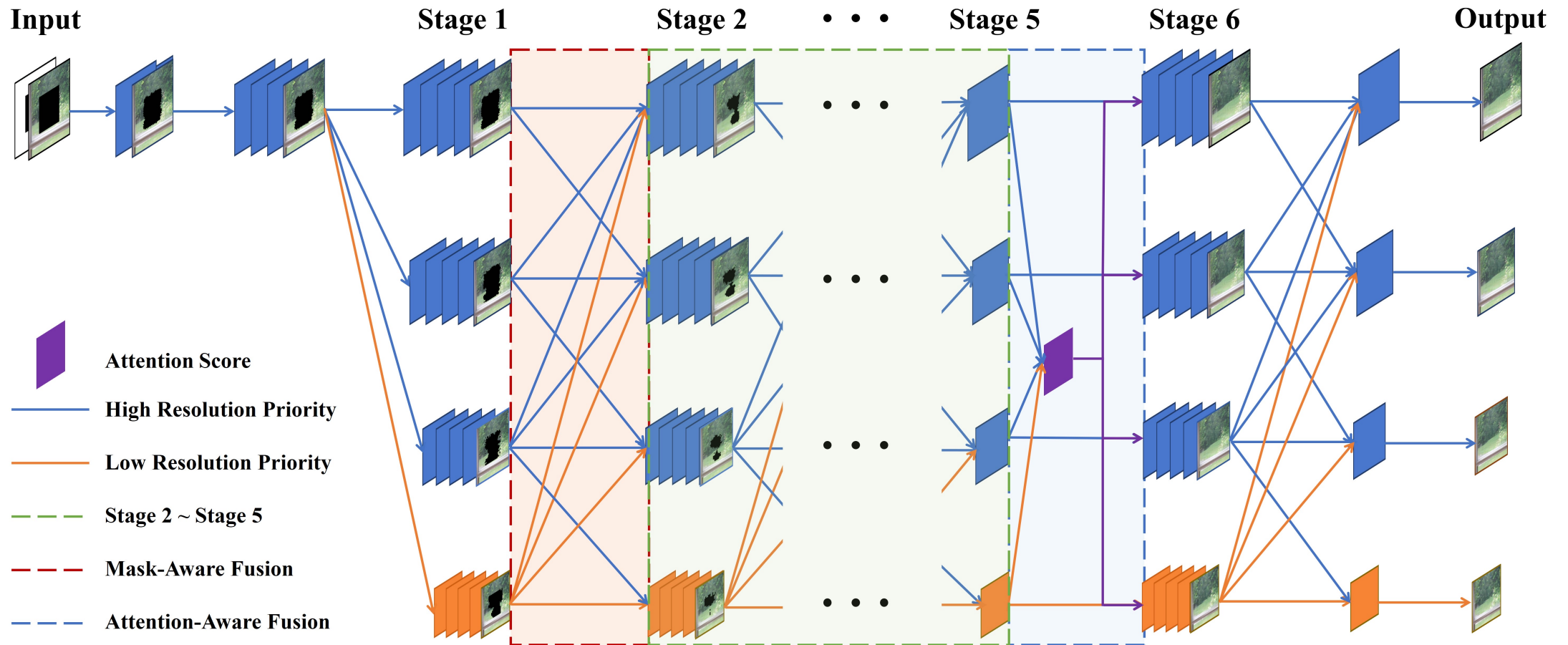
- Parallel multi-resolution architecture always maintains high resolution representation.

## ◆ For **Issue 2**.

- Multiple multi-resolution feature fusions make a better integration of structure and texture.
- Two fusion methods:
  - Mask-aware representation fusion.
  - Attention-guided representation fusion.

# Proposed Method

- Parallel Multi-Resolution Inpainting Network



# Proposed Method

- Inpainting Priority

## Step 1: Mask update

$$m' = \begin{cases} 1, & \text{if } \text{sum}(\mathbf{M}_p) > 0 \\ 0, & \text{otherwise} \end{cases}$$

## Step 2: Feature update

$$x_p = \begin{cases} \frac{\Omega_p}{\text{sum}(M_p)} W \cdot (X_p \odot M_p) + b, & \text{if } m' = 1 \\ 0, & \text{otherwise} \end{cases}$$

**Partial Conv<sup>[1]</sup>**

## Step 1: Mask update

$$m' = \begin{cases} 1, & \text{if } m = 1 \text{ or } q \geq \delta \cdot q^{\max} \\ 0, & \text{otherwise} \end{cases}$$

The calculate of the pixel  $x$  priority  $q$ :

$$q = \text{sum}(\mathbf{M}_p) \cdot \rho^l(x)$$

Common priority term :  $\text{sum}(\mathbf{M}_p)$

Low-resolution priority ( $l = 0$ ):

$$\rho^l(x) = |n_p \cdot \nabla X_p^\perp|$$

High-resolution priority ( $l = 1, 2, 3$ ):

$$\rho^l(x) = |n_p \cdot \nabla (X_p - X_{p\uparrow\downarrow})|$$

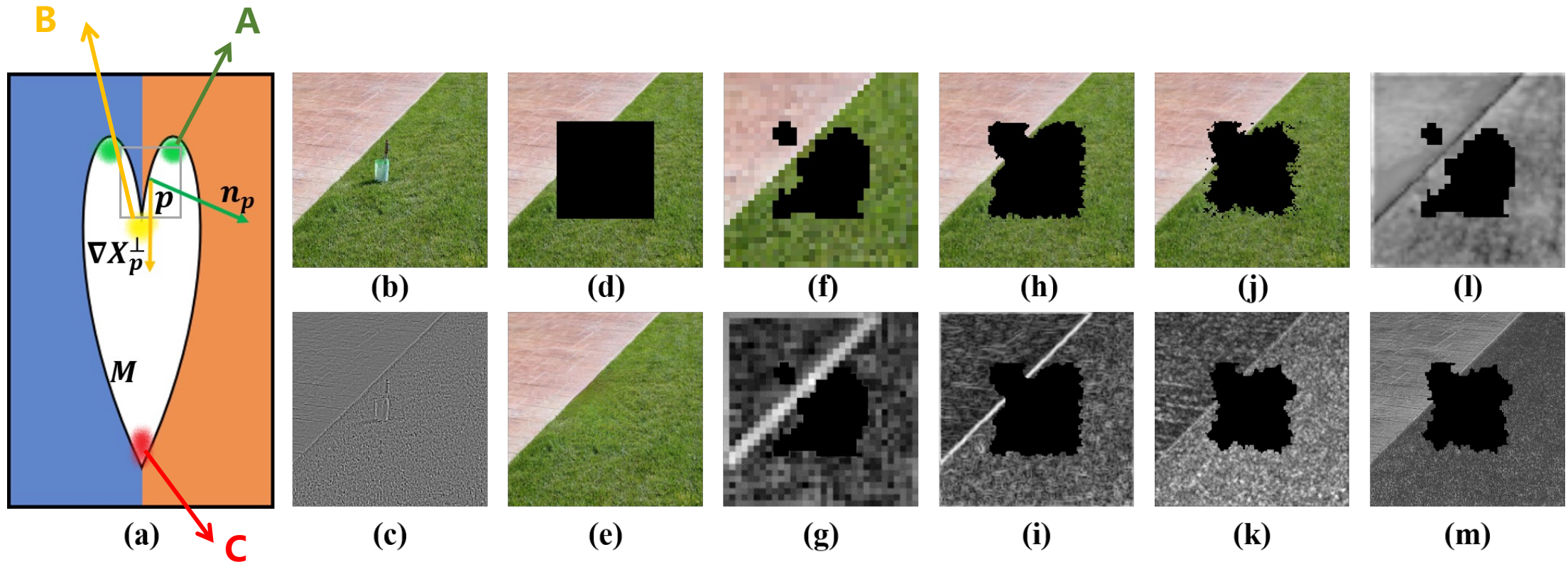
## Step 2: Feature update

$$x_p = \begin{cases} \frac{\Omega_p}{\text{sum}(M_p)} W \cdot (X_p \odot M_p) + b, & \text{if } m' = 1 \\ 0, & \text{otherwise} \end{cases}$$

**Ours**

# Proposed Method

- Inpainting Priority



**A:** High common priority but low contour striking priority.

**B:** High contour striking priority but low common priority.

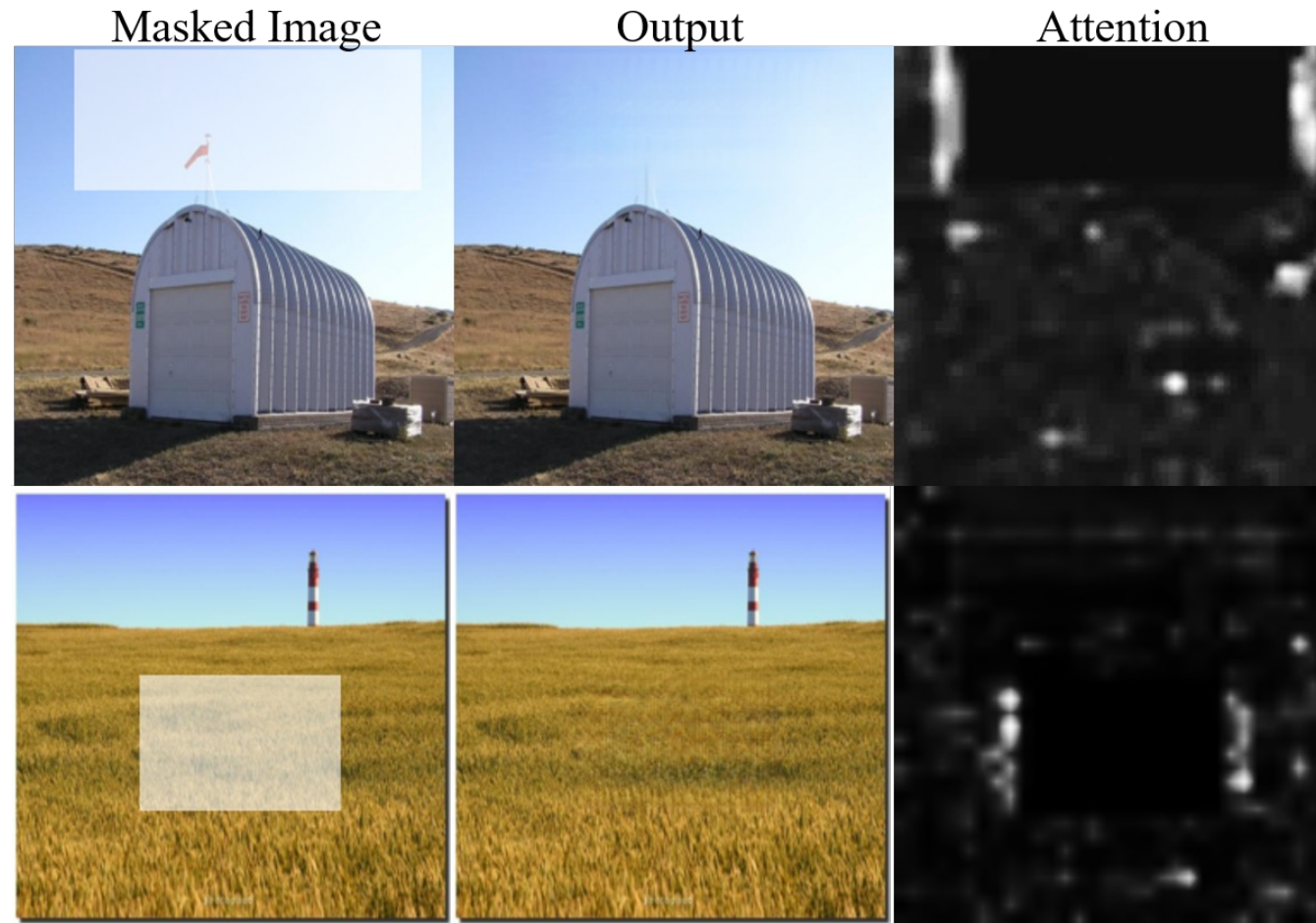
**C:** Both high common priority and high contour striking priority.



# Proposed Method

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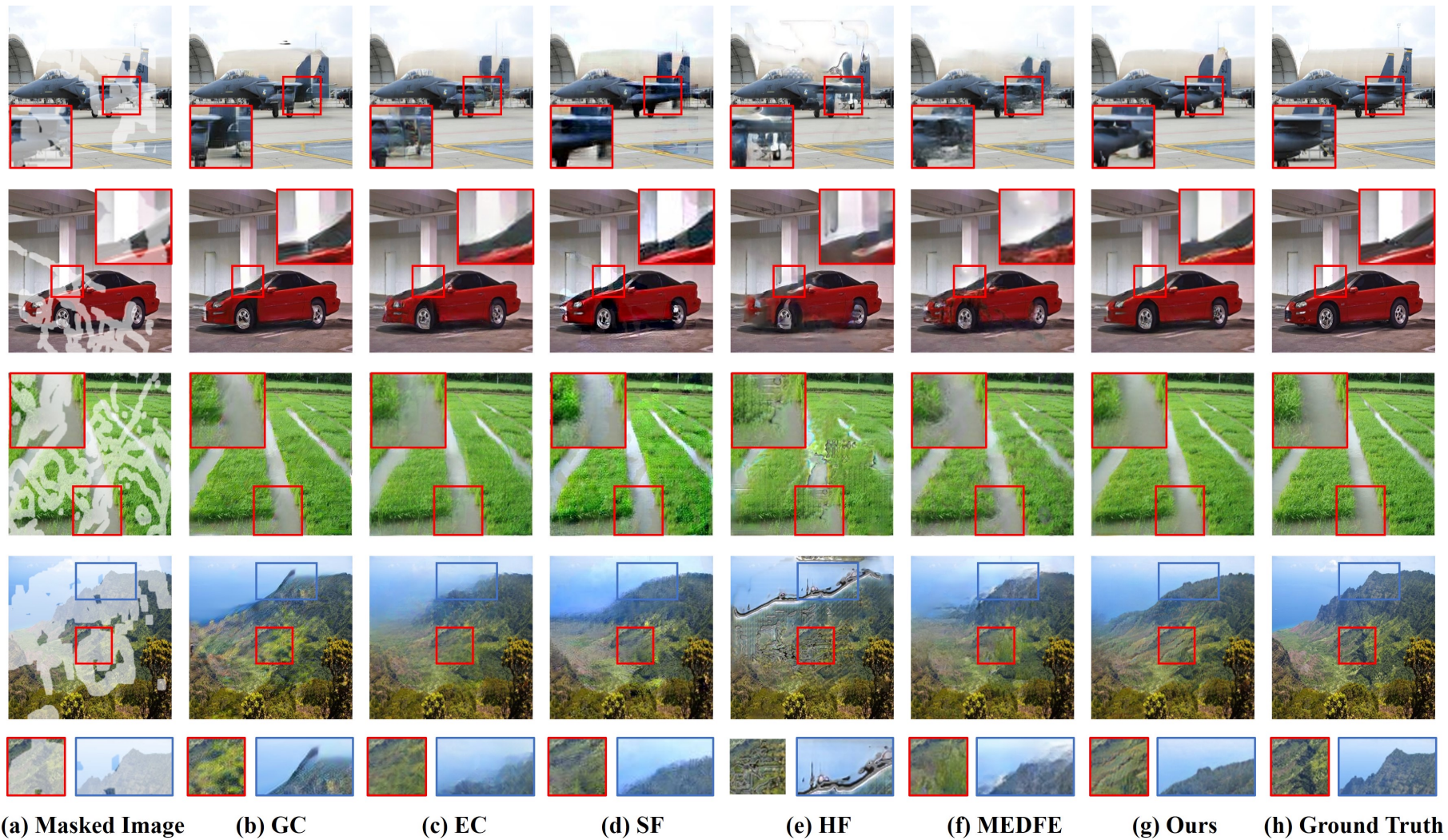
- Attention-Guided Representation Fusion



Masked region attends relevant information from unmasked region. E.g., wheat land (resp., sky) for wheat land (resp., sky).

# Experimental Results

- Qualitative comparison on Places2





Thanks for watching!

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