# **Deep Joint Image Filtering**

Yijun Li<sup>1</sup>, Jia-Bin Huang<sup>2</sup>, Narendra Ahuja<sup>2</sup>, Ming-Hsuan Yang<sup>1</sup>

<sup>1</sup>University of California, Merced

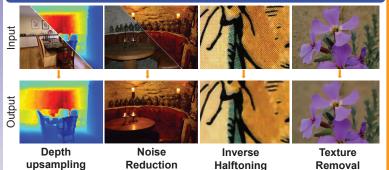
<sup>2</sup>University of Illinois, Urbana-Champaign



Project webpage: http://bit.ly/deepjointfiltering



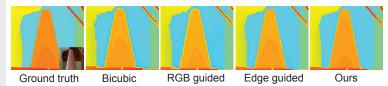
# Joint image filtering



# Limitations of existing methods

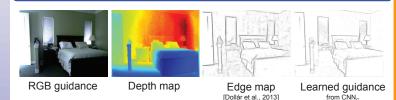
- 1. Fail to consider mutual structures.
- 2. Hand-crafted objective functions.
- 3. Inefficiency for optimization-based methods.

# **Design rationale**

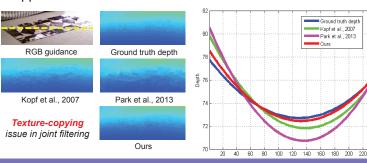


- Directly stack the RGB guidance image and target depth image, and feed them through a generic network → poor performance
- Replace the RGB guidance image with its edge map [Dollár et al., 2013] → good performance
- End-to-End: Extract structural features from both the target and guidance image. Then, combine them and reconstruct.

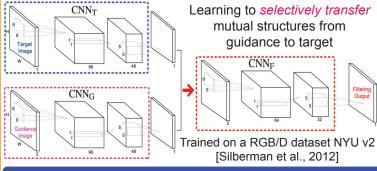
# What has the network learned?



- · The learned guidance appears like an edge map.
- CNN<sub>T</sub> and CNN<sub>G</sub> show strong responses to edges from the target and guidance image respectively.
- CNN<sub>E</sub> re-organizes the extracted structural features and suppresses inconsistent details.



#### **Network architecture**

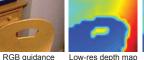


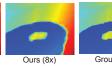
## **Contributions**

- 1. A learning-based approach for joint image filters.
- 2. State-of-the-art performance on depth upsampling.
- 3. A generic filter to handle image data in a variety of domains.

## **Experiments**

1. Depth map upsampling





2. Chromaticity map upsampling









3 times faster than the direct solution on the highresolution intensity images

Bicubic

Ours

3. Saliency map upsampling





GF [He et al., 2010]



RGB guidance

4. Inverse halftoning





Our result 5. Cross-modal noise reduction







NIR/RGB



Non-Flash/Flash Our denoised result

RMSE comparisons for depth map upsampling.						
	Middlebury (#30, [0,255])			NYU v2 (#449, cm)		
Method	4x	8x	16x	4x	8x	16x
He et al., 2010	4.01	7.22	11.70	7.32	13.62	22.03
Ferstl et al., 2013	3.39	5.41	12.03	6.98	11.23	28.13
Ham et al., 2015	3.14	5.03	8.83	5.27	12.31	19.24
Ours	2.14	3.77	6.12	3.54	6.20	10.21