# NAS-DIP: Learning Deep Image Prior with Neural Architecture Search



Yun-Chun Chen

#### Chen Gao

#### Esther Robb

Jia-Bin Huang



### Learning-based Methods



#### Deep Image Prior (DIP) [Ulyanov et al. CVPR 2018]



## NAS-DIP (Ours)



Neural Architecture Search (NAS)

#### **Overview of NAS-DIP**



# **Model Training**



#### **NAS** Training



## Testing



## Search Space

• Upsampling cell

- Cross-scale residual connections
  - Shared cross-level patterns
  - Progressive upsampling



1 x 1

1

None

Bilinear

2D convolution

# Search Space for the Upsampling Cell



## **Cross-scale Residual Connections**



#### Decomposition



Direct upsampling

Progressive upsampling (Ours)

## Weight Sharing



Progressive upsampling (Ours)



# **Quantitative Results**

Mathad	$\mathbf{Set5}$			$\operatorname{Set} 14$		
Metnod	$2 \times$	$4 \times$	$8 \times$	2  imes	$4 \times$	$8 \times$
Bicubic	33.66	28.44	24.37	30.24	26.05	23.09
Glasner et al.	-	28.84	-	-	26.46	-
TV prior	-	28.85	24.87	-	26.42	23.48
RED	-	30.23	25.56	-	27.36	23.89
$\operatorname{DeepRED}$	-	30.72	26.04	-	27.63	24.28
$\mathbf{SelfExSR}$	<b>36.60</b>	30.34	25.49	<b>32.2</b> 4	27.41	23.92
DIP	33.19	29.89	25.88	29.80	27.00	24.15
Ours	35.32	30.81	<b>26.4</b> 1	31.58	<b>27.84</b>	<b>24.59</b>

Method	Inpainting	Denoising
Papyan et al.	31.19	-
DIP	33.48	30.43
SGLD	34.51	30.81
Ours	<b>34.72</b>	<b>31.42</b>

# **Visual Comparisons**





Bicubic



DIP





Ours



# **Model Transferability**



Hazy image





Ground truth



DoubleDIP (U-Net)



DoubleDIP (Ours)

#### Winter $\rightarrow$ Summer



Input



CycleGAN (U-Net)



CycleGAN (Ours)



Input





CycleGAN (U-Net)



 ${f CycleGAN} \ (Ours)$ 

# Summary



- Search for neural architectures for inverse image problems.
- Search spaces for the upsampling layer and cross-level residual connections.
- State-of-the-art results on image restoration tasks.
- Model transferability.



Super-Resolution Denoising

Inpainting

Dehazing

Translation