

## **IDEA**:

### IMDN[1] uses:



#### Problems !!!



Requires a lot of parameters

Marginal Contributions (Manga 109) IIC (0.04dB) < PRM (0.24dB)

IMDB uses standard 2D Convolutions



#### **Proposed Solutions !!!**

IIC PRM Replace with mechanism feature for aggregation, **Global Progresive Refinement Module, GPRM** 

Network-in-Network[3] type structure

Remove Channel Contrast Aware Attention at the local lave and add Block Non-Local Attention in the main branch

IMDB's standard 2D Replace Conv with Modified GBlocks of 2021 Mobile SISR AI winner [2], XLSR grouped using convolutions. Grouped Distilling Blocks, Information GIDB

# IMDeception: Grouped Information Distilling Super-Resolution Network Mustafa Ayazoglu

(mayazoglu@aselsan.com.tr)

Aselsan Research, Ankara, Turkey

# IMDeception



# **Global Progresive Refinement Module, GPRM**



# **Grouped Information Distilling Block, GIDB**



# **Modified GBlock**



# **RESULTS:**

Original Bicubic









#### Parameters

IMDeception	core = 16 + NLA	core = 12	core = 8	cc
#Params	316K	198K	113K	
FLOPS[G]	20.7	12.9	7.4	
Act.[M]	206	149	103	
Runtime[ms]*	60	45	31	
#Conv2d	62	58	58	
#Div2K Val (PSNR)	29.02	28.82	28 70	



# Summary/Conclusion

Very efficient SR network only with **316K** parameters and **20G** FLOPS Many high performing variants of it exist with little sacrifice of the performance Can run on edge devices in real-time with inputs up to 512x256

# References

[1] Zheng Hui, Xinbo Gao, Yunchu Yang, and Xiumei Wang. Lightweight image super-resolution with information multidistillation network. In Proceedings of the 27th ACM International Conference on Multimedia. ACM, oct 2019 [2] Mustafa Ayazoglu. Extremely lightweight quantization robust real-time single-image super resolution for mobile devices. In IEEE Conf. Comput. Vis.

Pattern Recog. Workshops, 2021

[3] Min Lin, Qiang Chen, and Shuicheng Yan. Network in network, 2014

Conf. Comput. Vis.2018



CARN[4]

### IMDN

#### IMDeception









# $bre = 4 + NLA \mid core = 4$

## Edge Device Comparison

Model Hardware	RTX 2080	Jetson Xavier	
Model	TensorRT (ms)	TensorRT (ms)	
IMDeception	177	88.9	
core=16+NLA	17.7		
IMDeception*	10	08.0	
core=12	19	96.9	
IMDeception	0.0	51.1	
core=8	9.9		
IMDeception	0.8	44.7	
core=4+NLA	9.0	44.7	
IMDeception	0.2	41.0	
core=4	9.2	41.7	

#### Performance Comparison

	S. 93			8 Å	5	
Model	Set5	Set14	BSD100	Urban100	Manga109	Div2K (Val)
Bicubic	28.42	26.00	25.96	23.14	24.89	26.66
LapSRN (812K)	31.54	28.19	27.32	25.21	29.09	28.75
IMDN (779K)	32.21	28.58	27.56	26.04	30.45	28.94
EDSR (43M)	32.46	28.80	27.71	26.64	31.02	29.25
CARN [2] (1.5M)	32.14	28.61	27.58	26.07	30.46	28.96
E-RFDN (433K)	32.16	28.65	27.60	26.15	30.59	29.04
IMDeception (316K)	20.14	29.64	27.00	26.20	20.67	20.02*
core=16 + NLA	32.14	28.64	27.60	26.20	30.67	29.02*
IMDeception(198K)	21.02	22 29 17	27.50	25.02	20.29	20 02
core=12	51.65	20.47	27.50	23.82	30.28	20.05
IMDeception(113K)	21.60	1 60 29 25	27.42	25.60	20.80	28.60
core=8	51.09	26.55	27.42	23.00	29.89	28.09
IMDeception(57K)	31 35	28.12	27.27	25.23	20.20	28.45
core=4	51.55	20.12	21.21	23.23	29.20	20.45
IMDeception(57K)	31 42	28.18	27.26	25.26	20.28	28.48
core=4 + NLA	51.42	20.10	27.20	23.20	29.20	20.40

[4] Namhyuk Ahn, Byungkon Kang, and Kyung-Ah Sohn. Fast, accurate, and lightweight super-resolution with cascading residual network. In Eur.