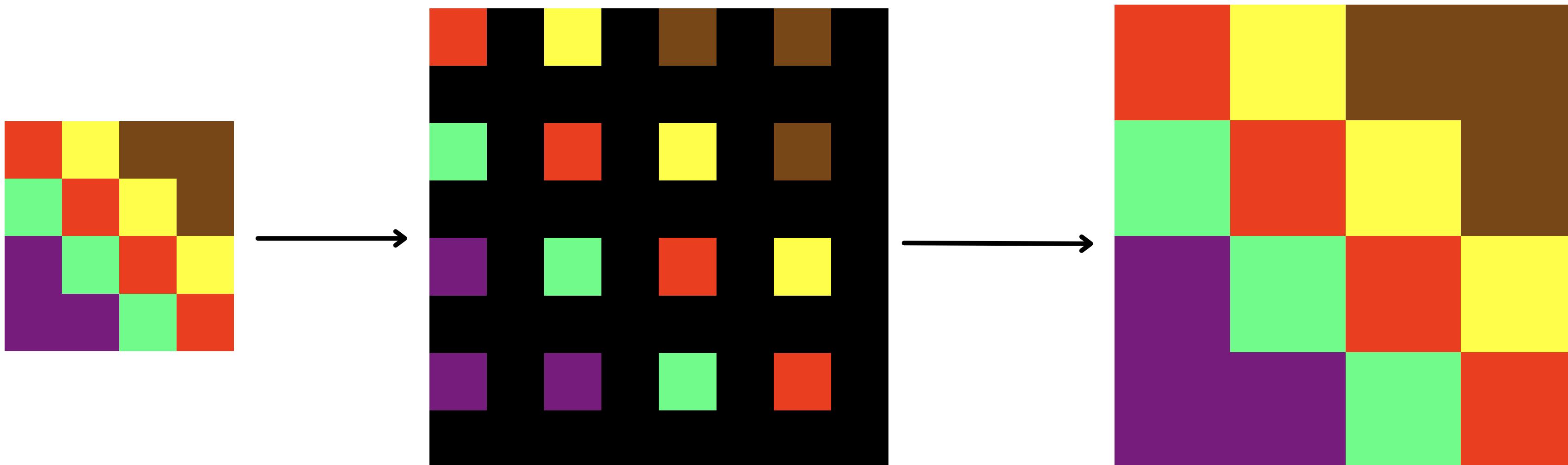


IMAGE SUPER-RESOLUTION

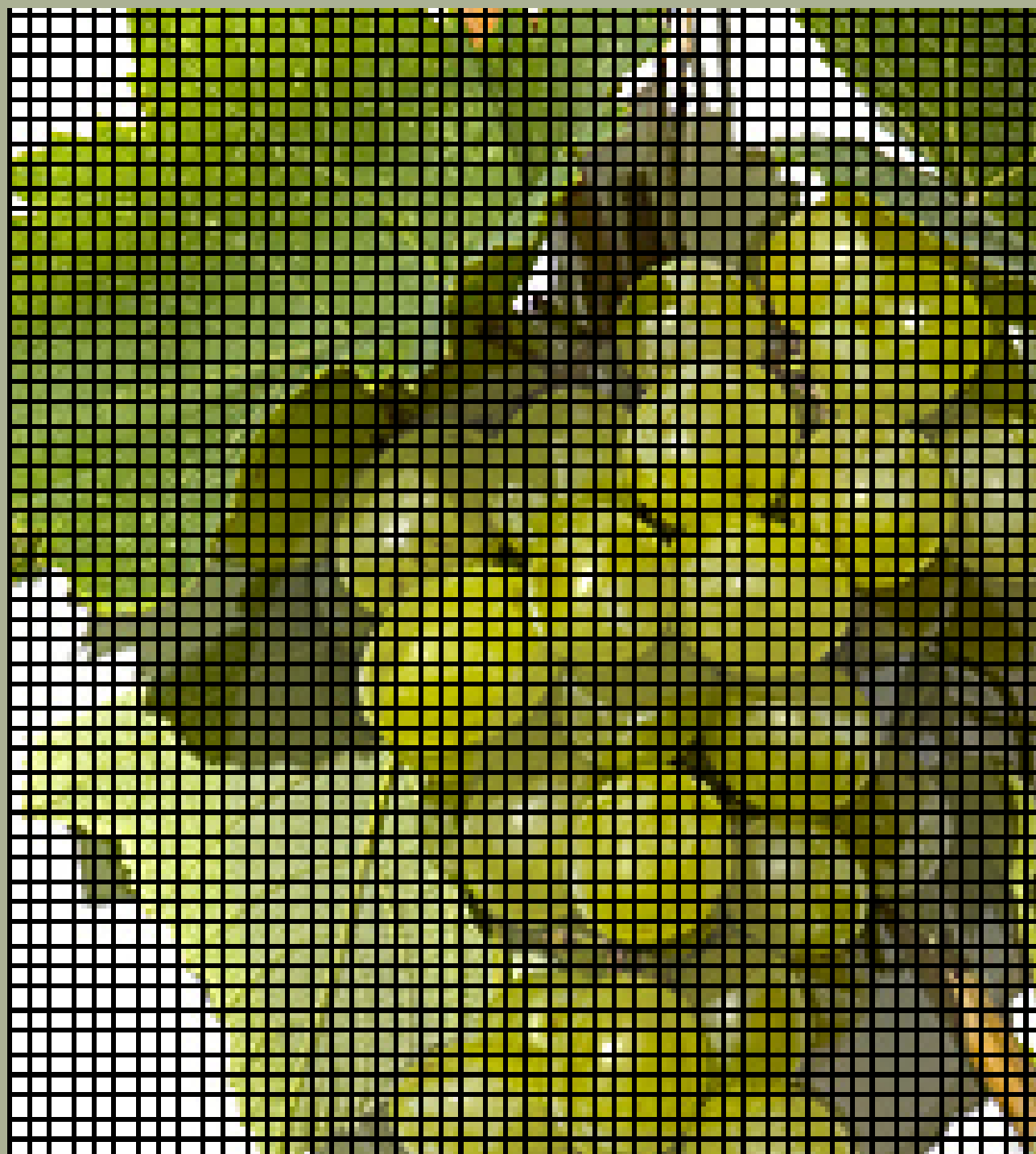
Или как увеличить разрешение изображения



IMAGE SUPER-RESOLUTION



NEAREST NEIGHBOR



NEAREST NEIGHBOR



BICUBIC

Data Processing Inequality

Processing data cannot add information content

$$X \rightarrow Y \rightarrow Z$$

$$I(X; Y) \geq I(X; Z)$$

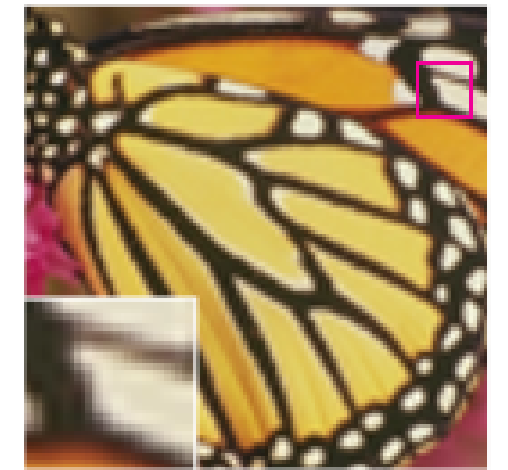
Image Super-Resolution Using Deep Convolutional Networks

Chao Dong, Chen Change Loy, *Member, IEEE*, Kaiming He, *Member, IEEE*,
and Xiaoou Tang, *Fellow, IEEE*

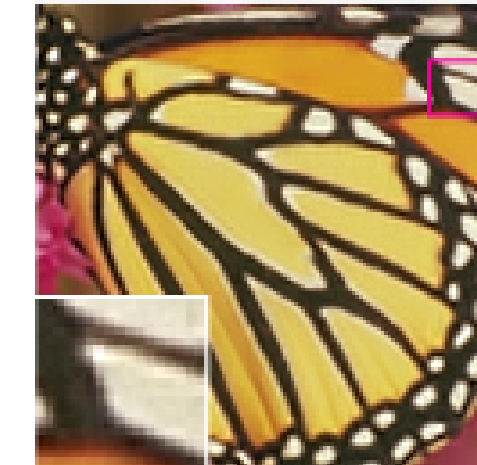
Abstract—We propose a deep learning method for single image super-resolution (SR). Our method directly learns an end-to-end mapping between the low/high-resolution images. The mapping is represented as a deep convolutional neural network (CNN) that takes the low-resolution image as the input and outputs the high-resolution one. We further show that traditional sparse-coding-based SR methods can also be viewed as a deep convolutional network. But unlike traditional methods that handle each component separately, our method jointly optimizes all layers. Our deep CNN has a lightweight structure, yet demonstrates state-of-the-art restoration quality, and achieves fast speed for practical on-line usage. We explore different network structures and parameter settings to achieve trade-offs between performance and speed. Moreover, we extend our network to cope with three color channels simultaneously, and show better overall reconstruction quality.



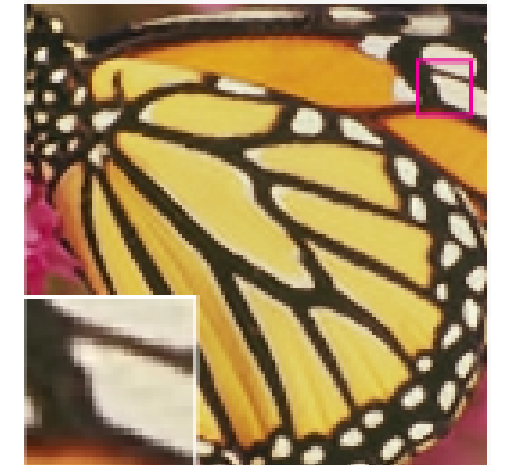
Original / PSNR



Bicubic / 24.04 dB



SC / 25.58 dB

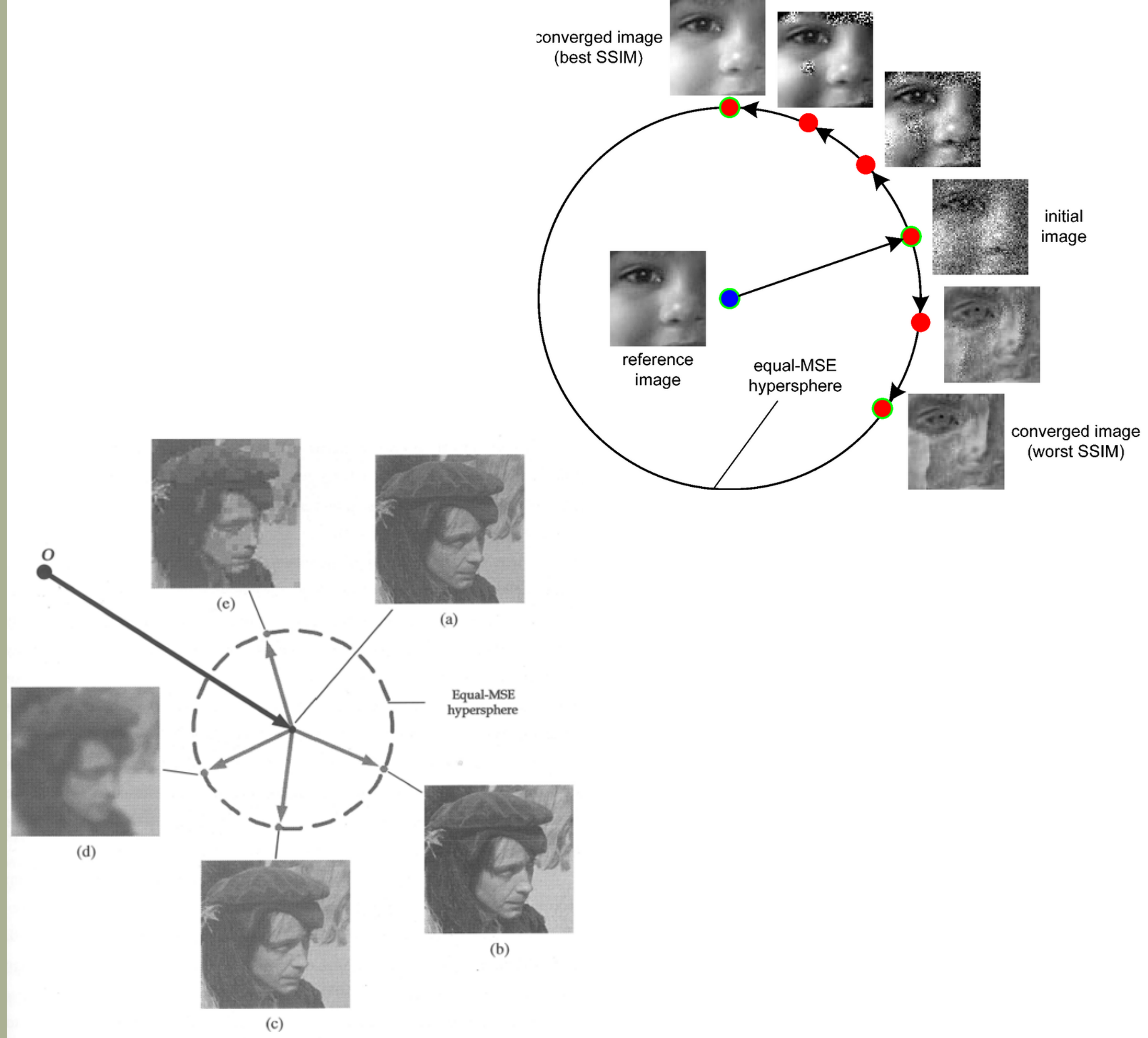


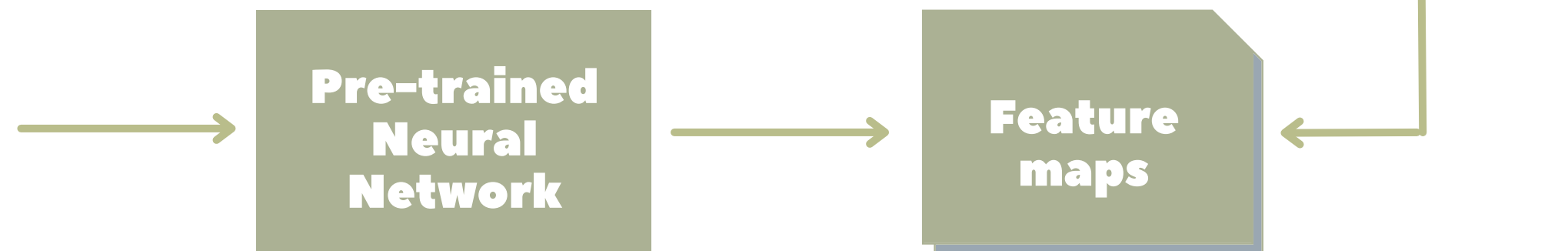
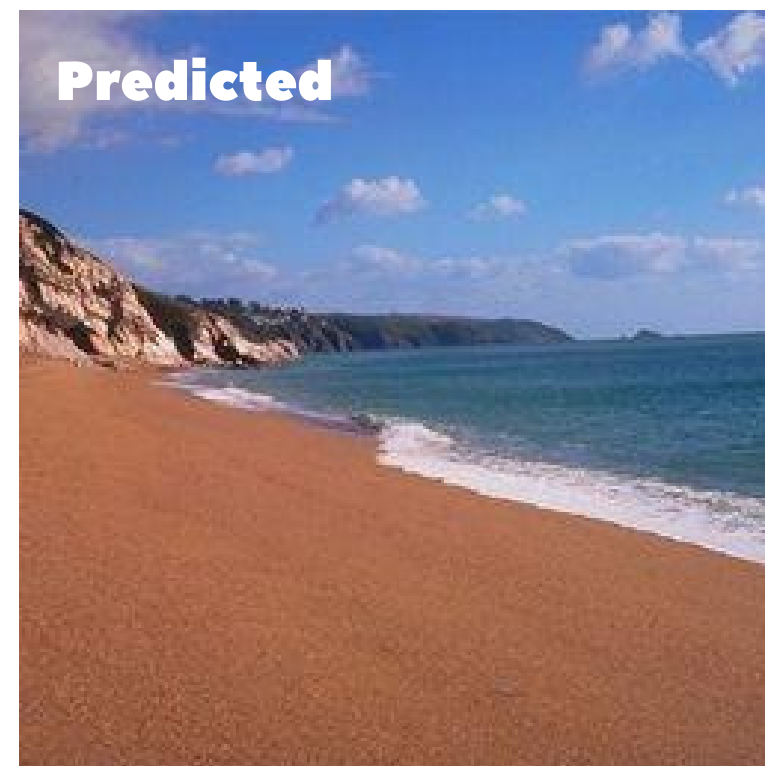
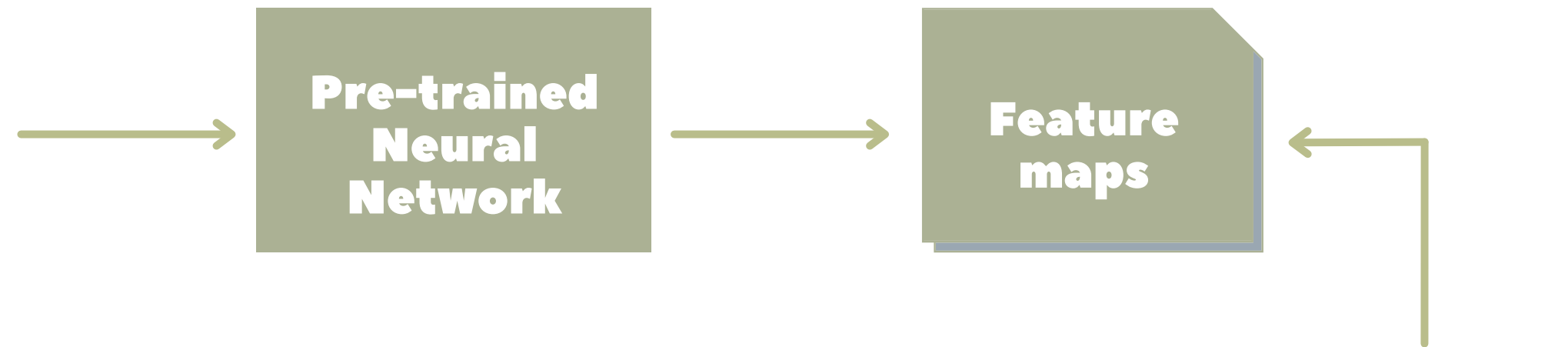
SRCNN / 27.95 dB

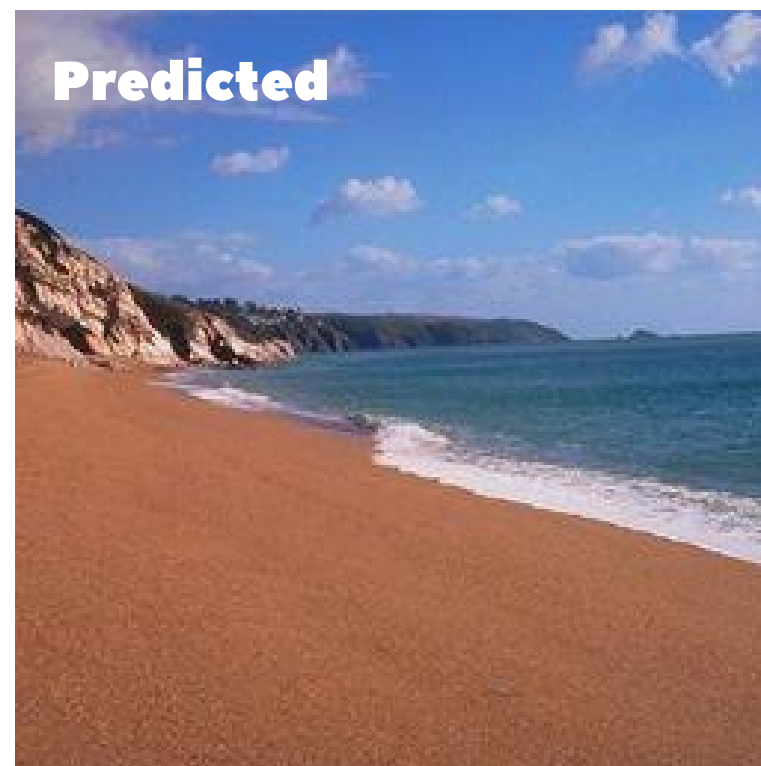
$$E[(\text{Output} - \text{Ground Truth})^2]$$

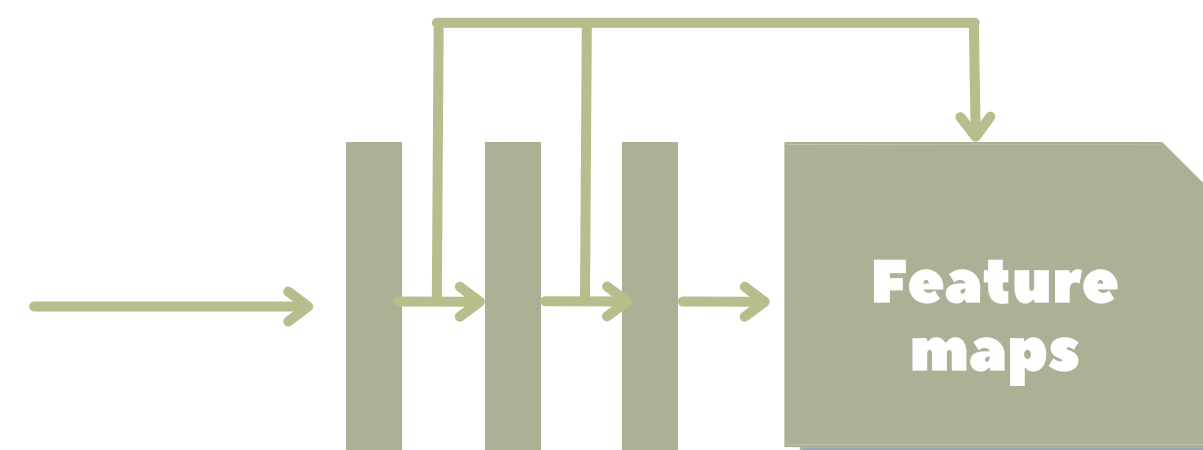
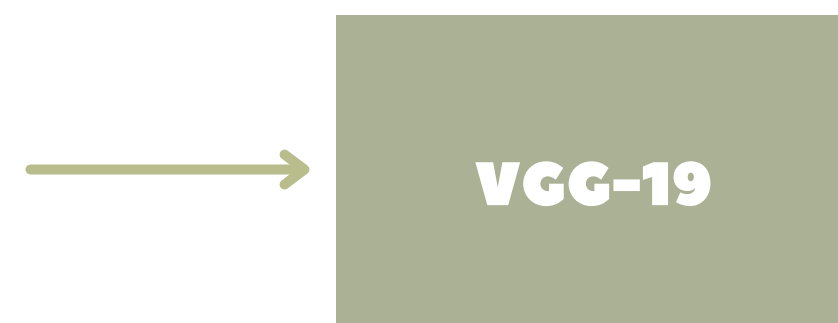
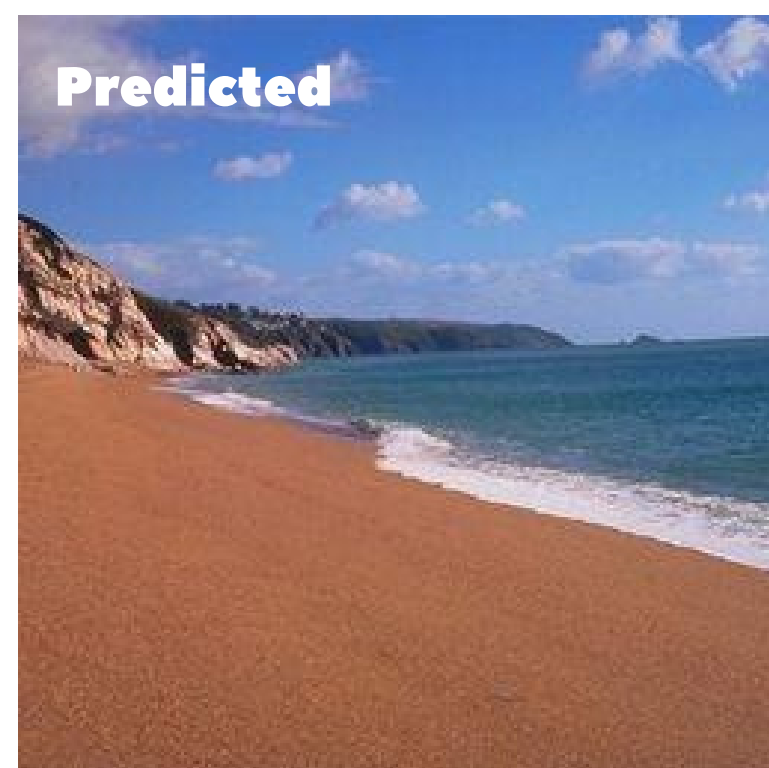
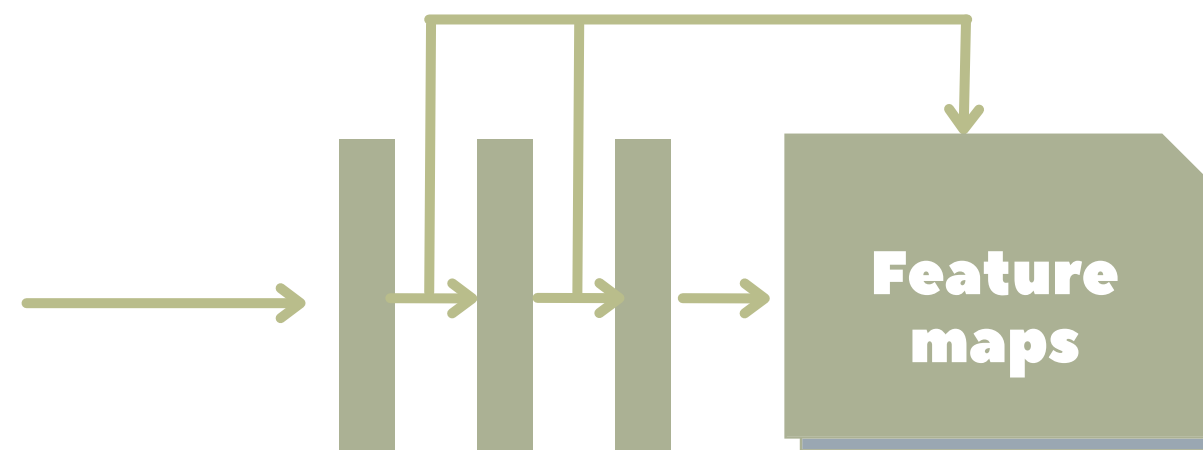
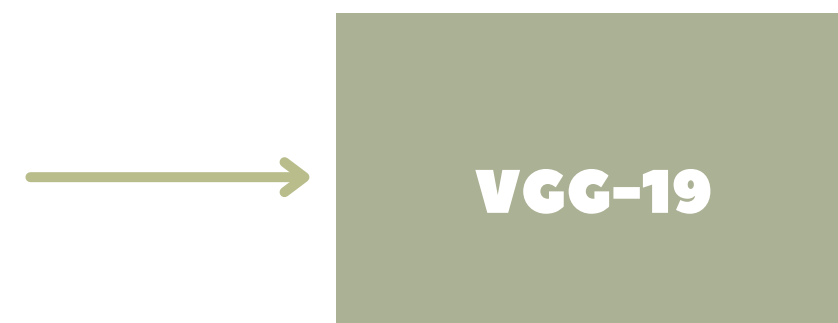
МИНИМИЗАЦИЯ КВАДРАТИЧНОЙ
РАЗНИЦЫ

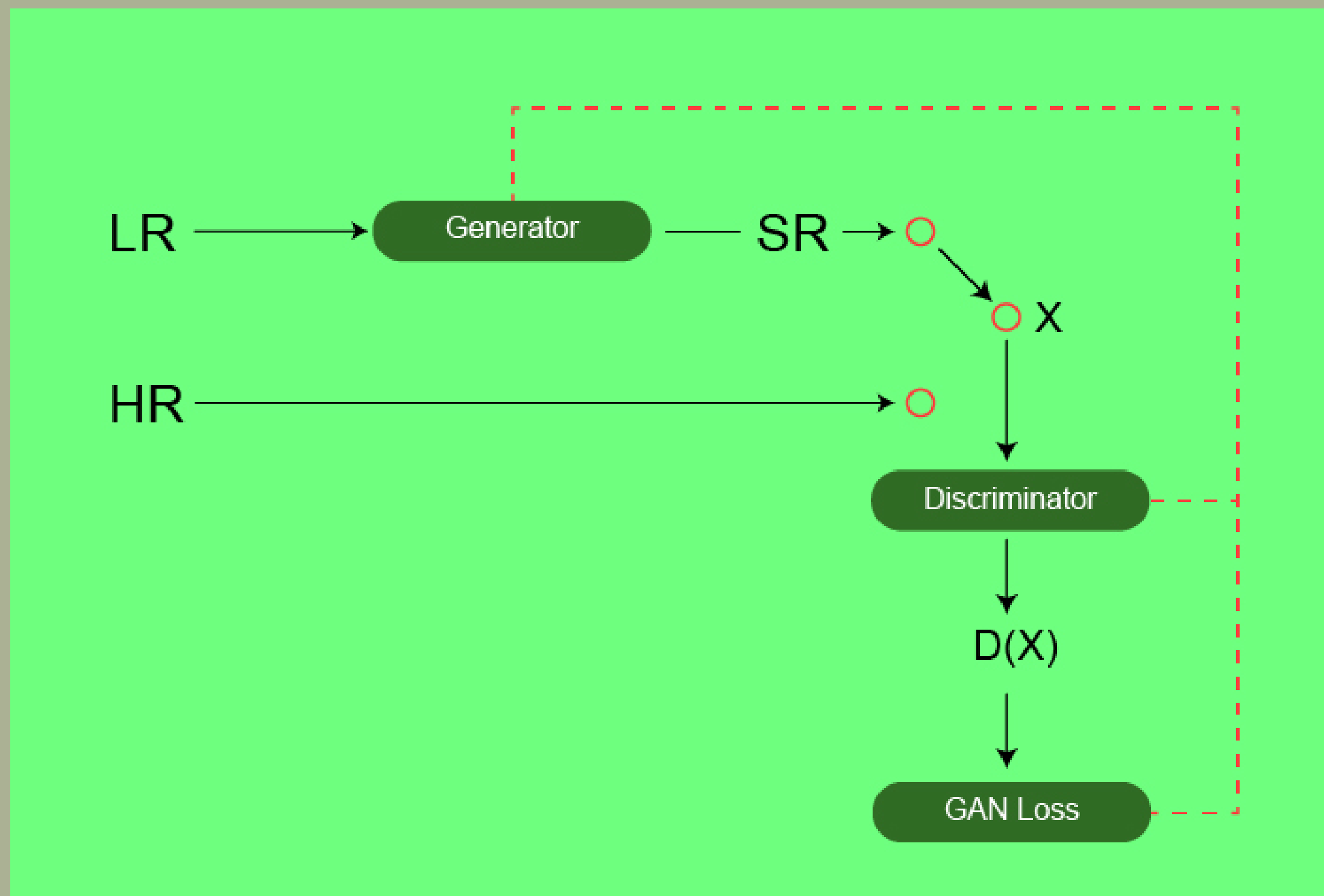
MEAN SQUARED ERROR









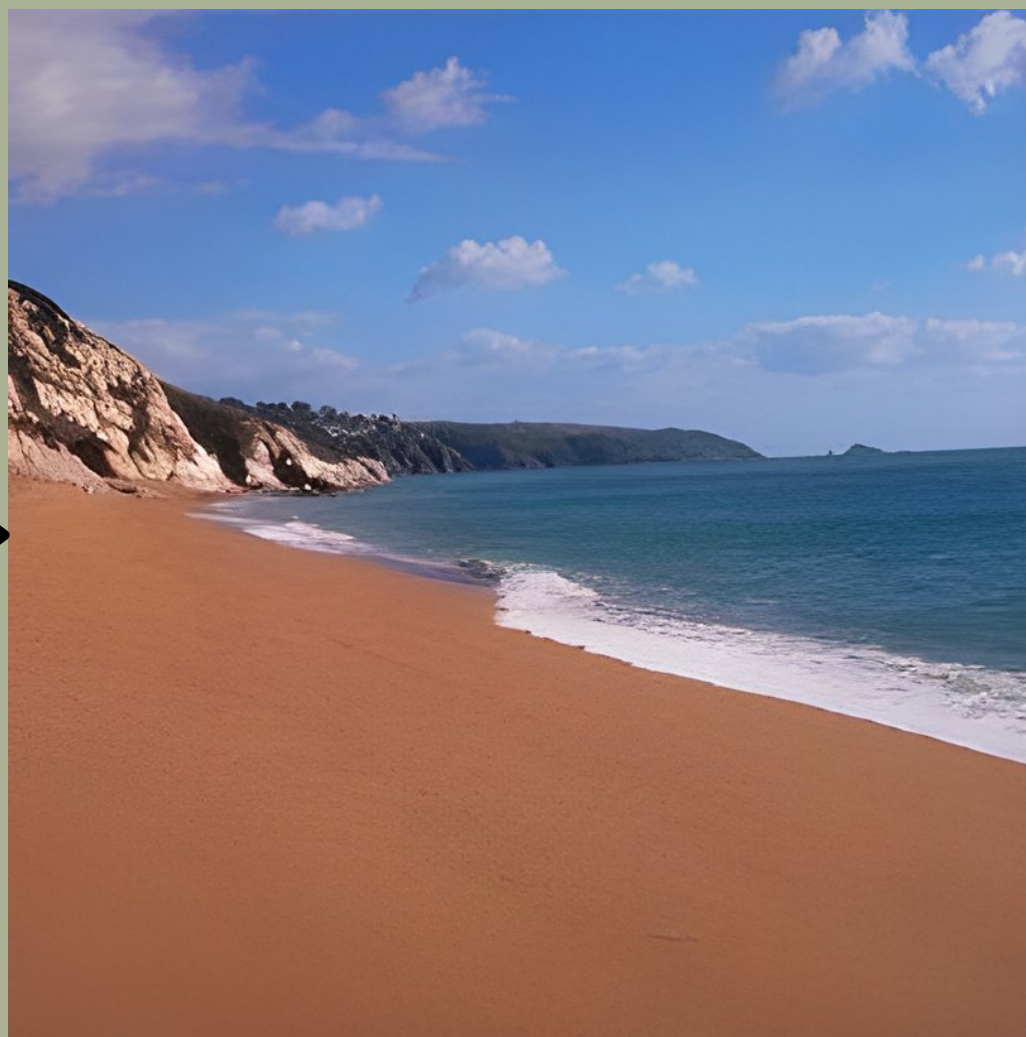


SRGAN



HIGH
RESOLUTION?

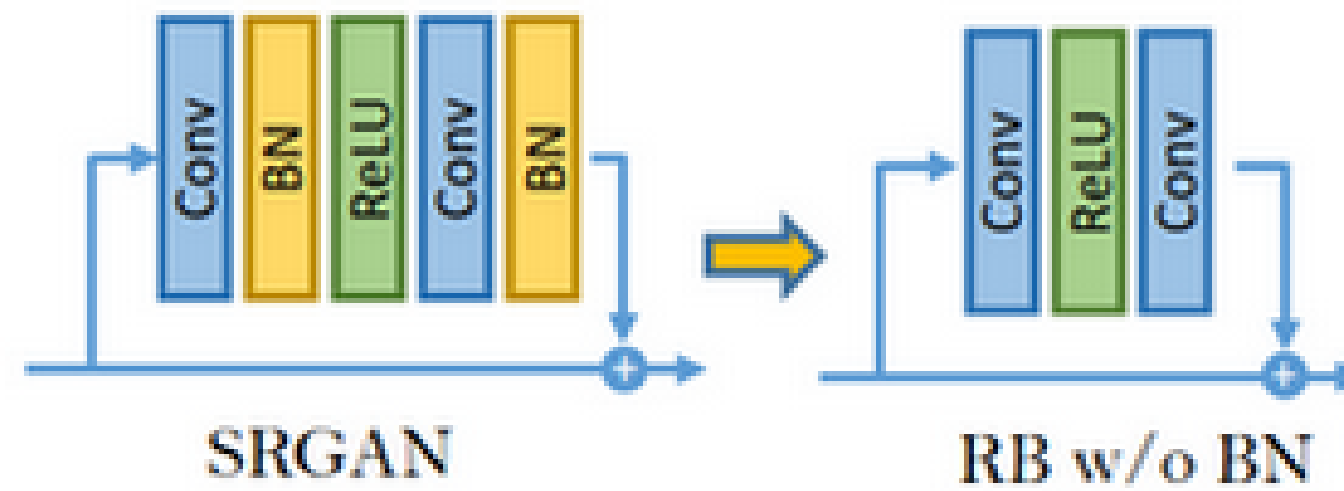
GAN



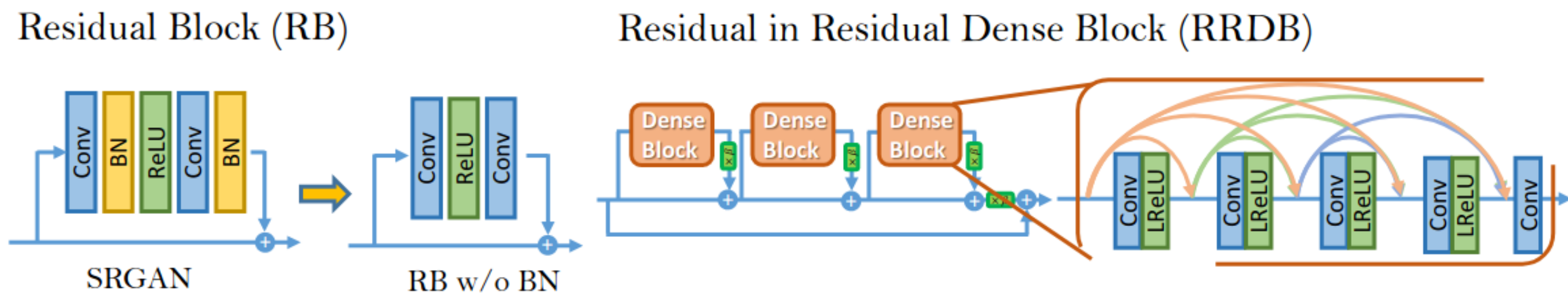
SRGAN

NO BATCH NORMALIZATION

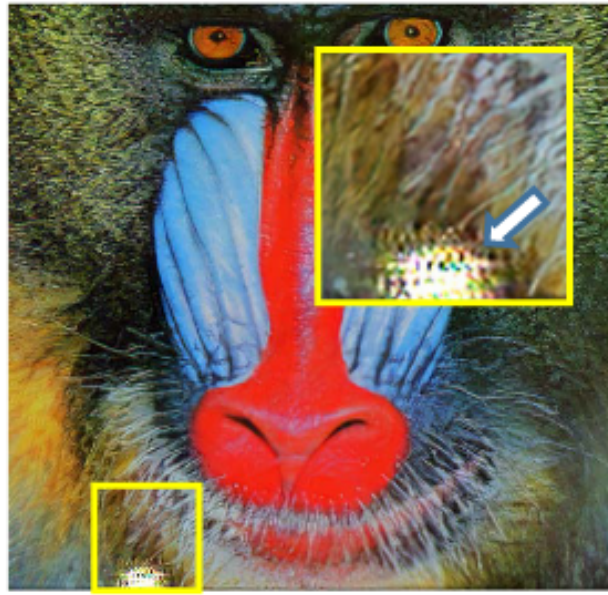
Residual Block (RB)



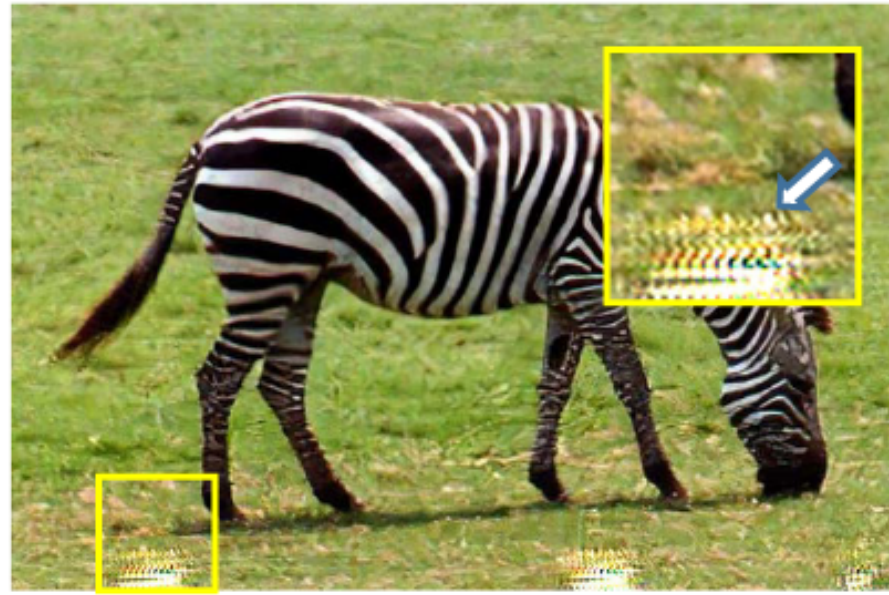
ESRGAN



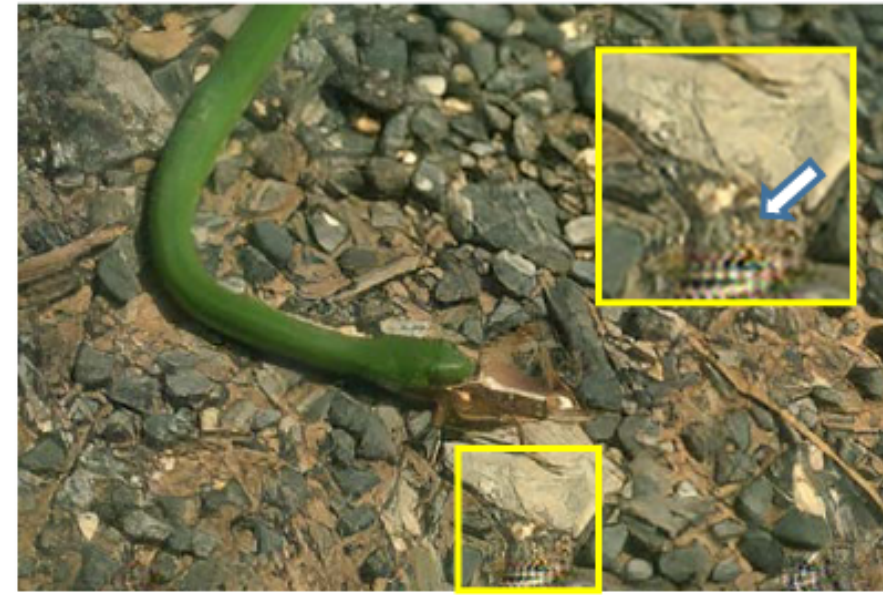
ESRGAN



baboon from Set14



zebra from Set14

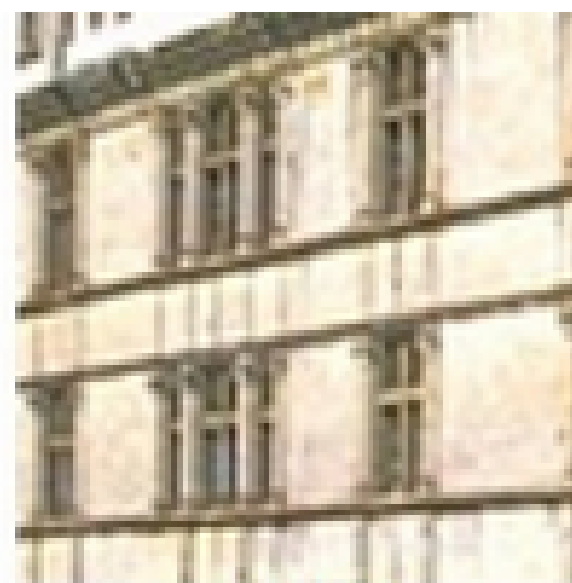


175043 from BSD100

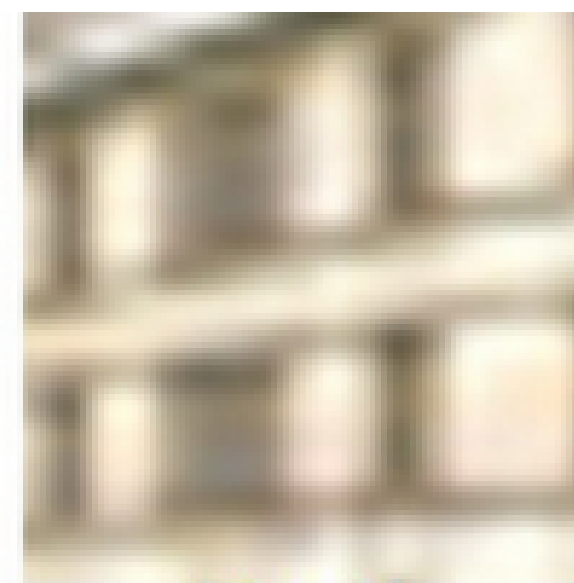
SRGAN ARTIFACTS



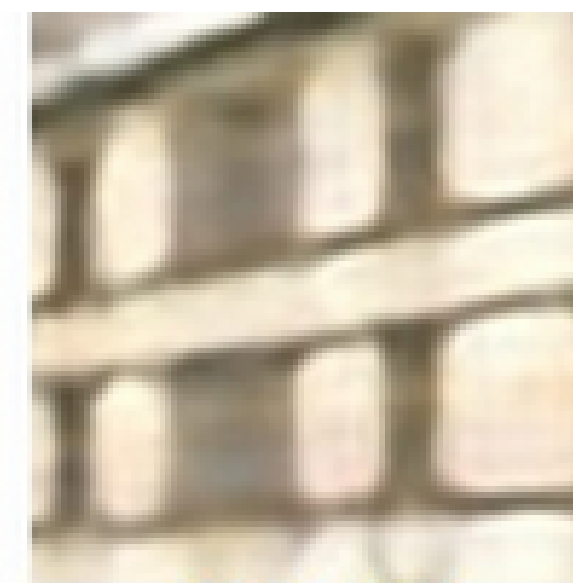
102061 from BSD100
(PSNR / Perceptual Index)



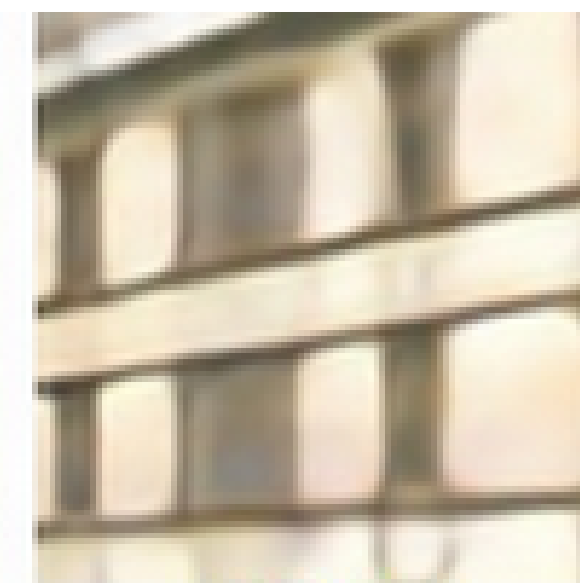
HR
(∞ / 2.12)



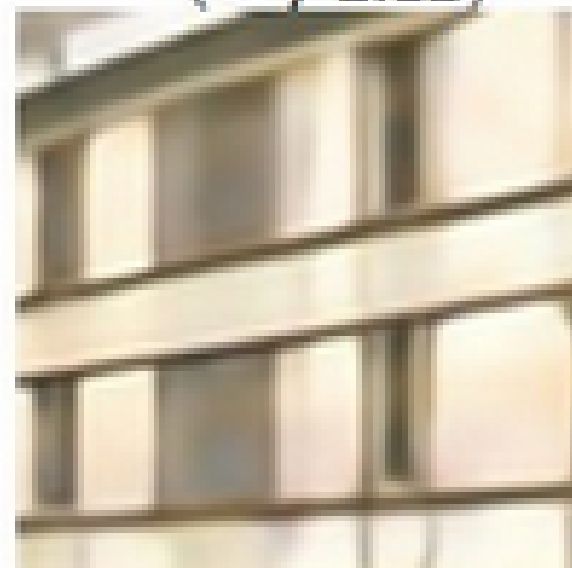
Bicubic
(25.12 / 6.84)



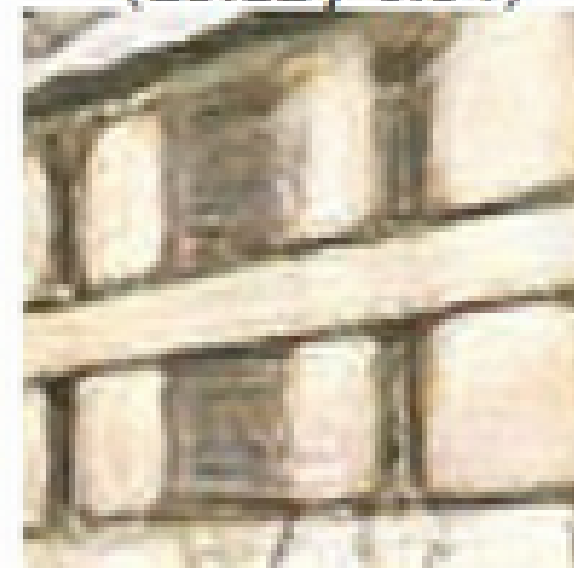
SRCNN
(25.83 / 5.93)



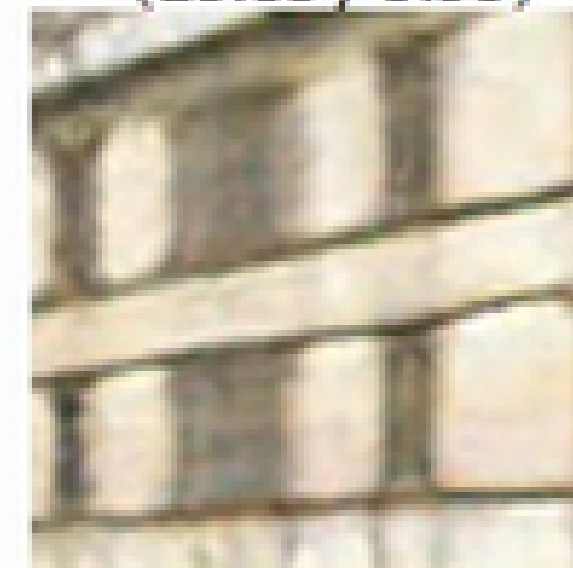
EDSR
(26.62 / 5.22)



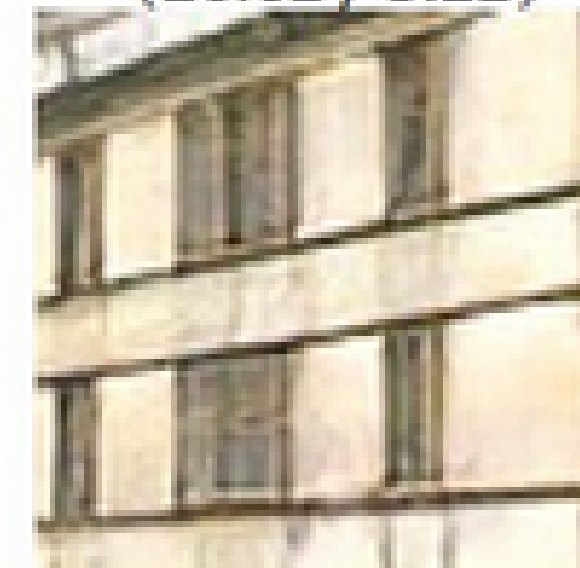
RCAN
(26.86 / 4.43)



EnhanceNet
(24.73 / 2.06)



SRGAN
(25.28 / 1.93)



ESRGAN(ours)
(24.83 / 1.96)

СРАВНЕНИЕ АЛГОРИТМОВ



ESRGAN

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NEIGHBOR-IMAGE-SCALING/](https://tech-algorithm.com/articles/nearest-neighbor-image-scaling/)
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- 3) [HTTPS://EN.WIKIPEDIA.ORG/WIKI/DATA_PR
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