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Decode barcode image using YOLOv8 & REAL-ESRGAN

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Most approaches (such as ESRGAN) assume an ideal Bicubic down-sampling kernel, which is different from real degradations.

- ESRGAN methods can't restore the real-life image.
 - Can't remove unknown blurs, complicated noises, and common compression artifacts
- > Real-ESRGAN is the extension of ESRGAN to practical restoration application



Challenges

- Unknown and complex degradations:
 - Usually, **paired training data** with similar degradations to real scenarios is required to train the networks.

Capture paired data with specific cameras followed by alignments

e.g., RealSR^[1]

Directly learn degradation distributions and then synthesize paired training data

e.g., Cycle-in-Cycle GAN^[2]

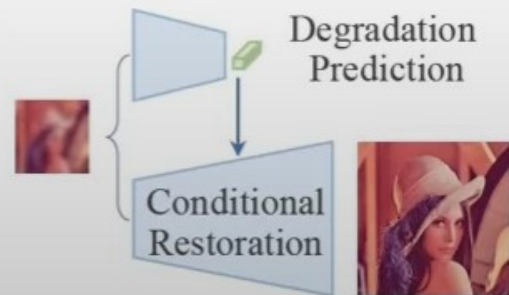
Synthesize paired data with classical operators and generalize trained models to real degradations

As close to real data as possible

- Deal with diverse degraded images in one unified network.

Typical blind SR methods:
two-branch network

e.g., DAN^[3]



Process all degraded images in one network



[1] Cai J, Zeng H, et al. Toward real-world single-image super-resolution. A new benchmark and a new model. ICCV 2019.

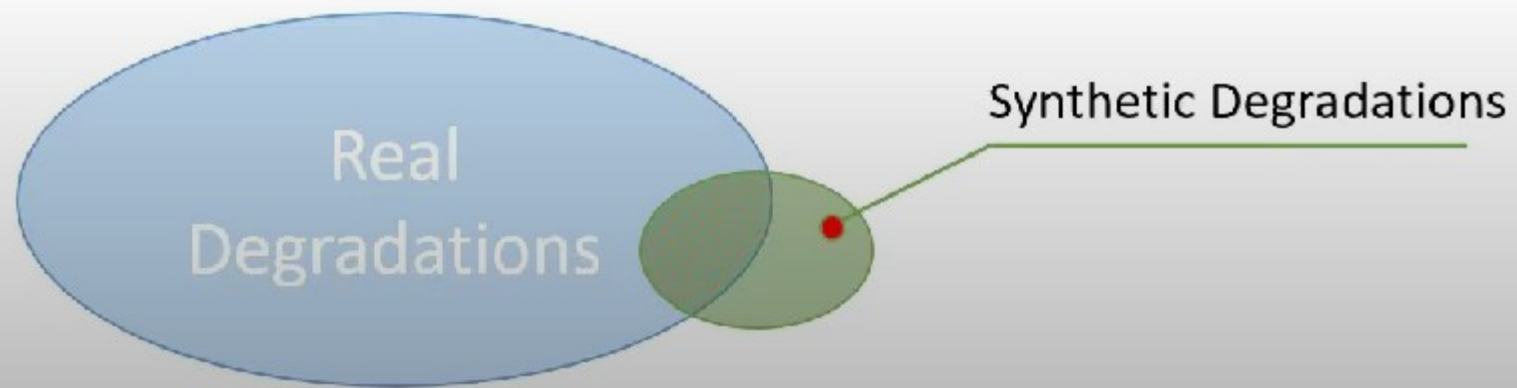
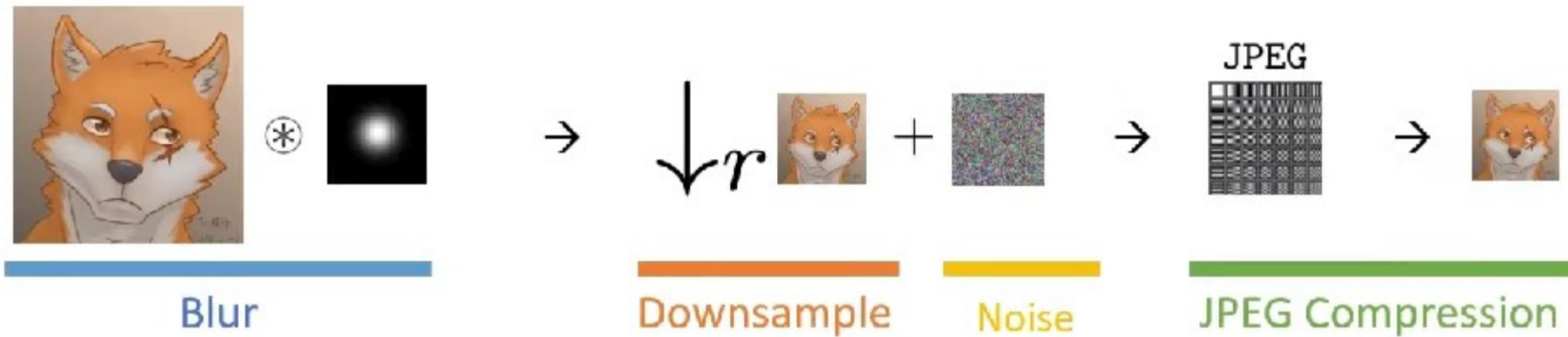
[2] Yuan Y, Liu S, Zhang J, et al. Unsupervised image super-resolution using cycle-incycle GAN. CVPRW 2018.

[3] Luo J, Huang Y, Li S, et al. Unfolding the alternating optimization for blind super-resolution. NeurIPS 2020.

[4] Xie L, Wang X, Dong C, et al. Finding Discriminative Filters for Specific Degradations in Blind Super-Resolution. NeurIPS 2021.

Finding Discriminative Filters for Specific Degradations in Blind Super-Resolution
Published on [NeurIPS 2021](#)

$$x = \underbrace{[(y \circledast k_\sigma)]}_{\text{Blur}} \underbrace{\downarrow_r}_{\text{Downsample}} \underbrace{+ n_\delta}_{\text{Noise}} \underbrace{] \text{JPEG}_q}_{\text{JPEG Compression}}$$

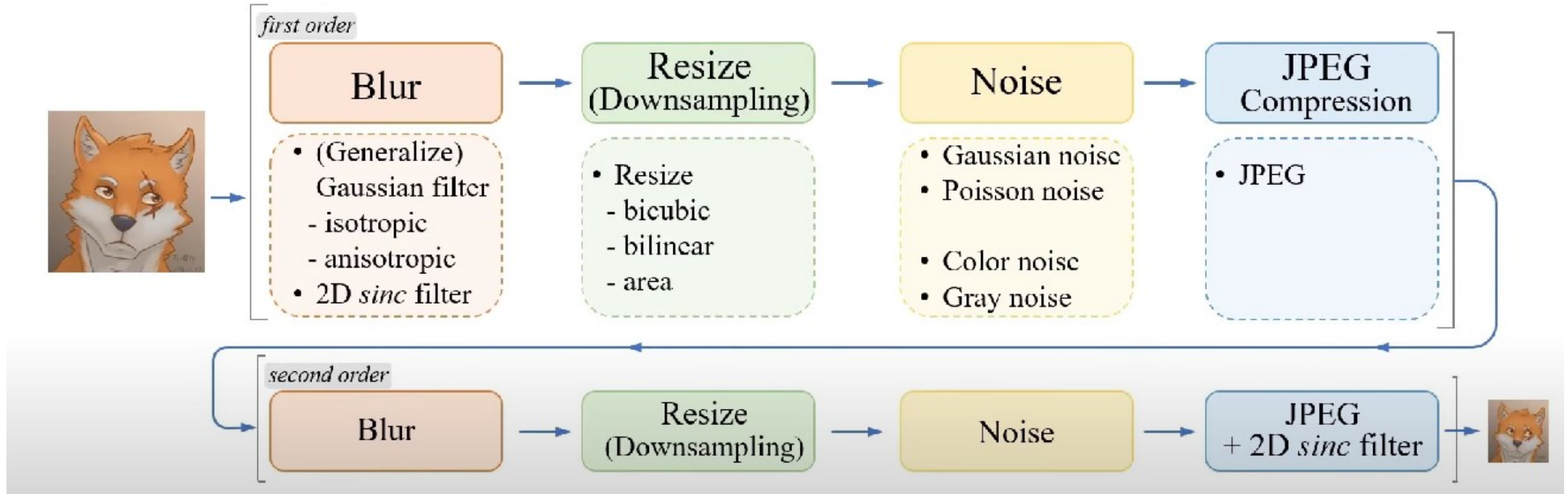


Complicated Combinations of Degradation Processes

The real complex degradations usually come from complicated combinations of different degradation processes, such as imaging system of cameras, image editing, and internet transmission.

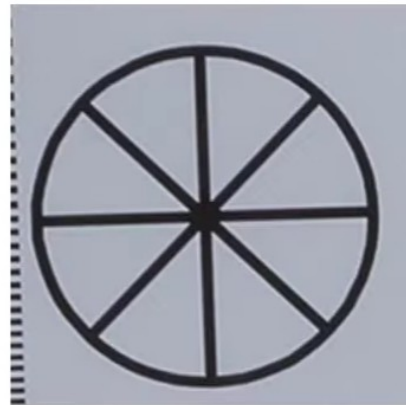
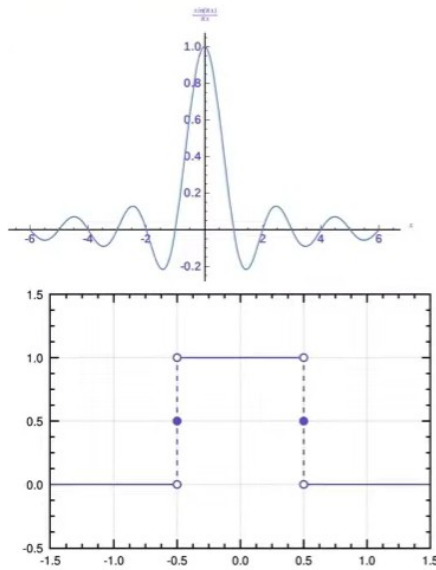
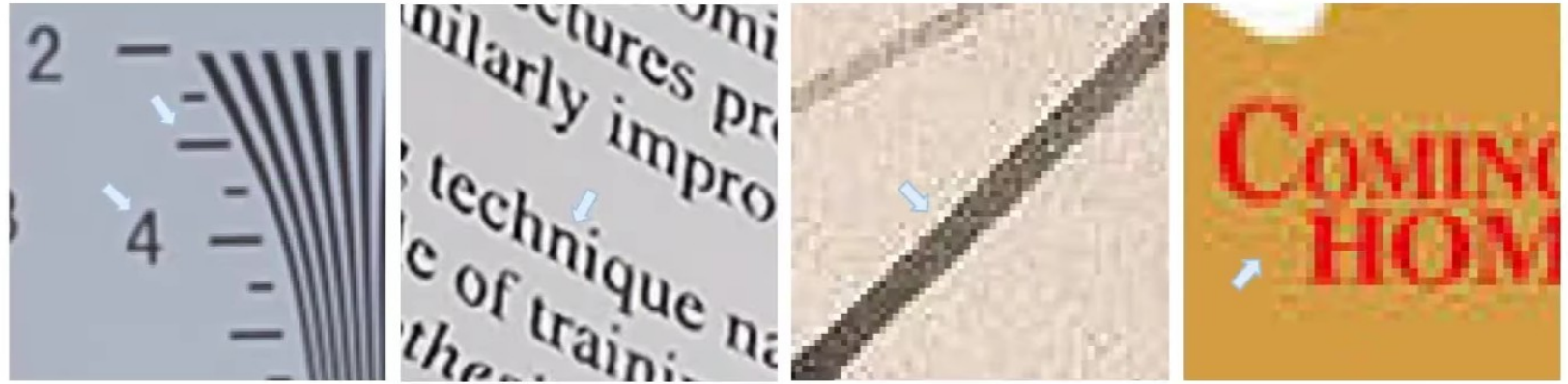


High-Order Degradation Process

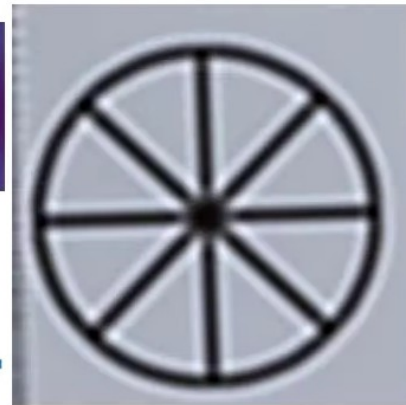
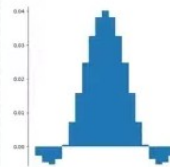
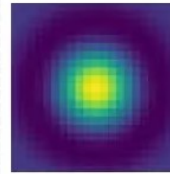


* The “high-order” here is different from that used in mathematical functions. Mainly refers to the implementation time of the same operation.

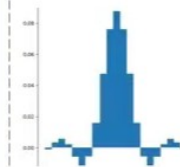
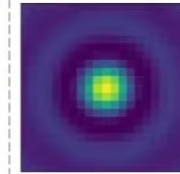
Real Samples



Input



$\omega_c = \pi/5$

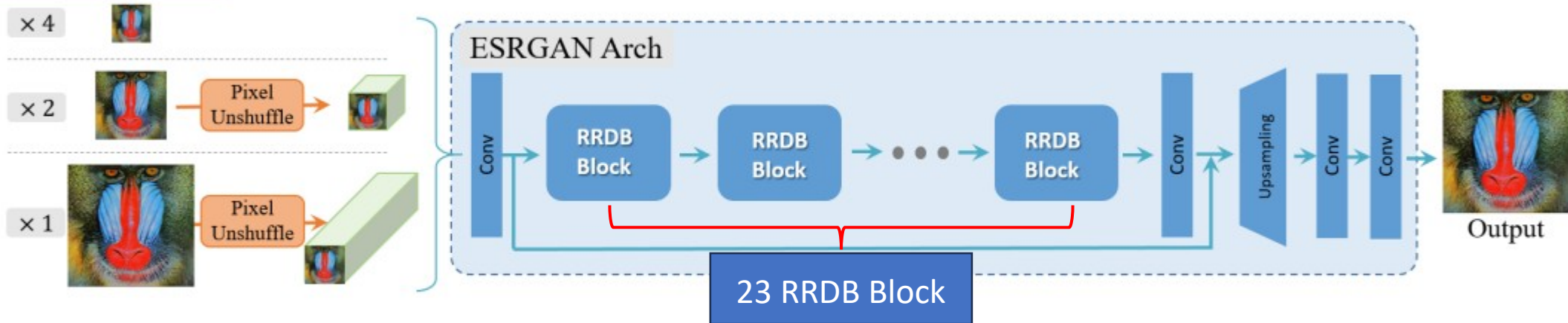


$\omega_c = \pi/3$

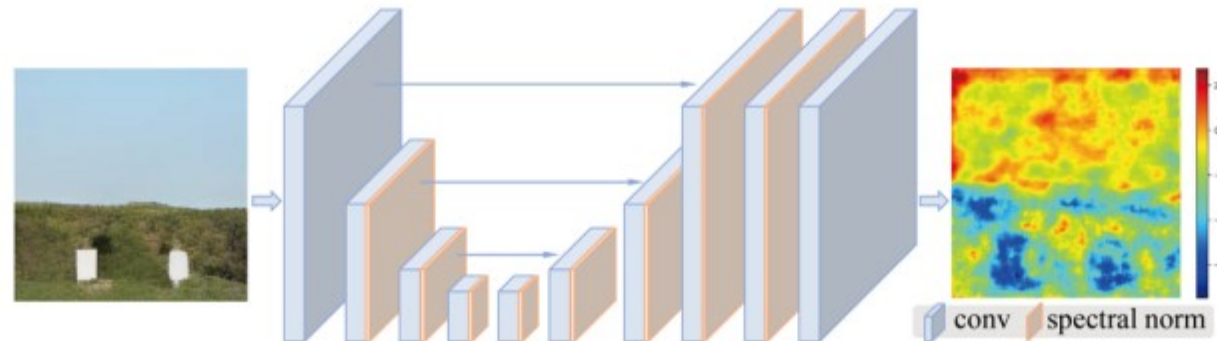
➤ Network Architecture

- We employ the same generator architecture as ESRGAN
- U-Net discriminator with spectral normalization is used to increase discriminator capability and stabilize the training dynamics

Generator



Discriminator



➤ Qualitative Comparisons



Bicubic



ESRGAN



DAN



CDC



RealSR



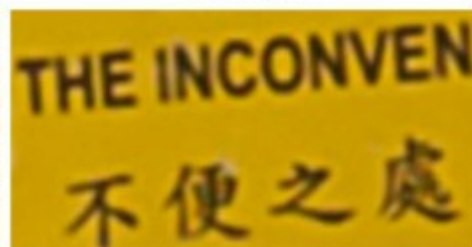
BSRGAN



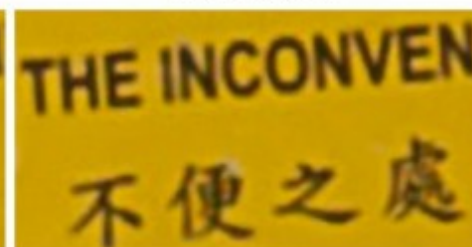
Real-ESRGAN



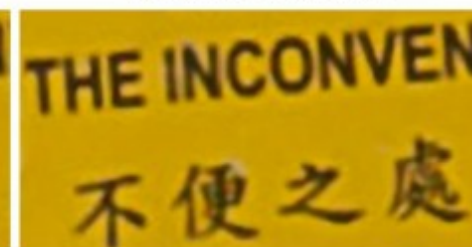
Real-ESRGAN+



Bicubic



ESRGAN



DAN



CDC



RealSR



BSRGAN

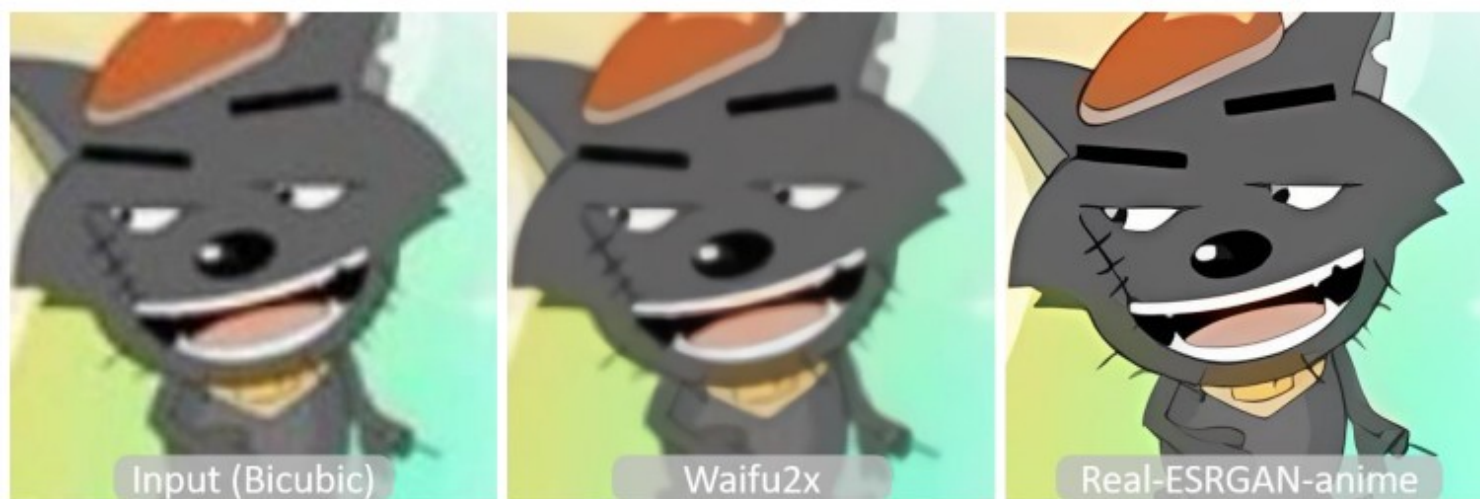


Real-ESRGAN



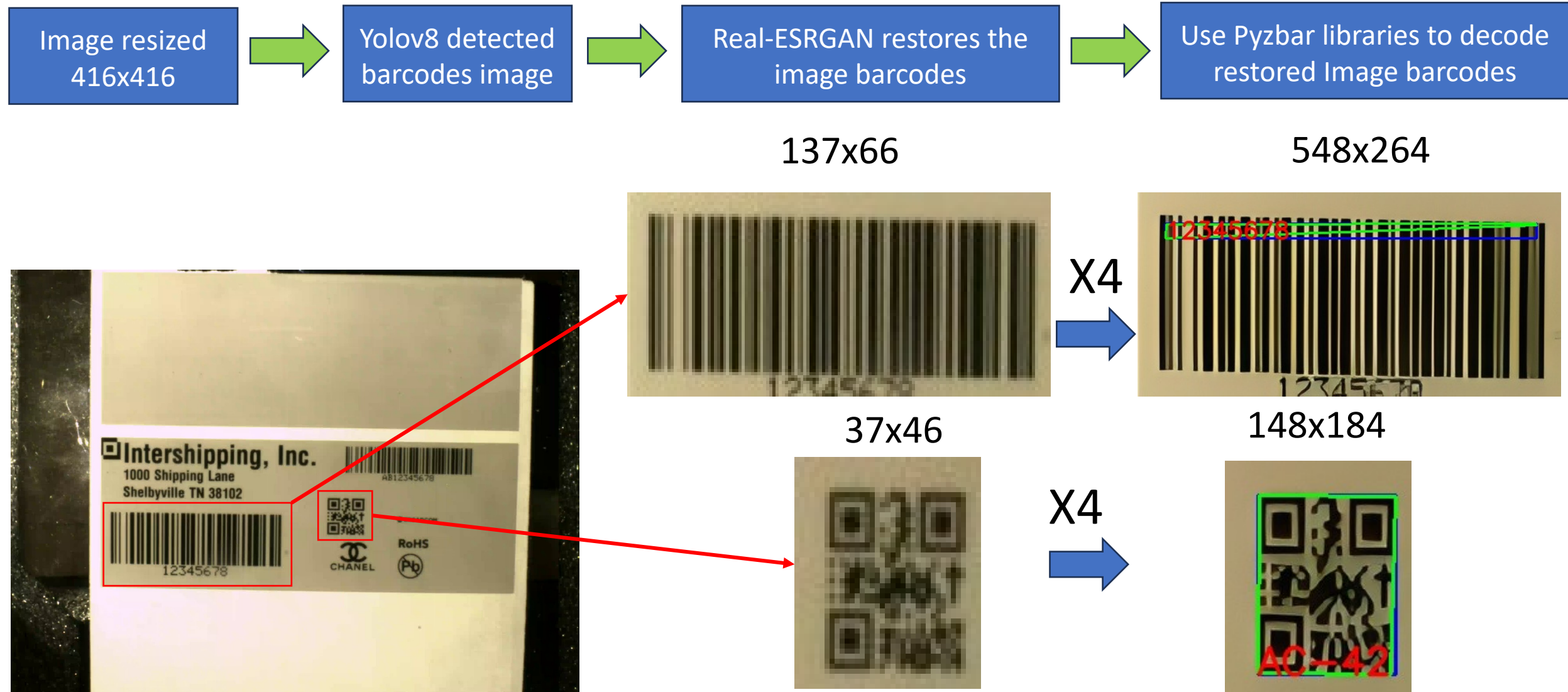
Real-ESRGAN+

➤ Optimize for Anime Images



Deploy on YOLOv8 and REAL-ESRGAN on a practical data

System workflow



weights

- RealESRGAN_x4.pth 9/6/2023 11:48 AM, 67
- RealESRGAN_x8.pth 9/8/2023 9:16 AM, 67
- RealESRGAN_x2.pth 9/8/2023 9:18 AM, 67

Model_YOLOv8

Reports

Save_Input

- 2023-09-10_17-47-24-804.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-24-994.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-25-187.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-49-174.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-49-365.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-49-558.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-52-858.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-53-065.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-53-251.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-58-948.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-59-156.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-59-344.jpg 9/11/2023 1:47 AM

Save_Output

- 2023-09-10_17-47-24-804.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-24-994.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-25-187.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-49-174.jpg 9/11/2023 1:47 AM
- 2023-09-10_17-47-49-365.jpg 9/11/2023 1:47 AM

main

0: 416x416 2 Barcodes, 1 QR Code, 10.6ms

Speed: 1.4ms preprocess, 10.6ms inference, 3.0ms postprocess per image at shape (1, 3, 416, 416)

The time of Enhance Super Resolution Image is: 441.24412536621094 ms

The time of Enhance Super Resolution Image is: 198.43697547912598 ms

The time of Enhance Super Resolution Image is: 186.51342391967773 ms

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File Settings Config Help About

START

STOP

...

OK: 0

NG: 0

Barcode 0.84

Barcode 0.75

QR_code 0.54

Intershipping, Inc.

1000 Shipping Lane

12345678

CHANEL

RoHS

Pb

REAL-ESRGAN reference time



GitHub

- My project is now available on Github 🔥: [Link](#)
- YOLOv8 🚀: [link](#)
- Real-ESRGAN 🚀: [link](#)
- Fast deployment Real-ESRGAN 🔥: [link](#)
- For easy integrated Real-ESRGAN on your project 📁: [link](#)

Thank you for listening