



## Introduction

### Challenges in Real-World Blind Super-Resolution

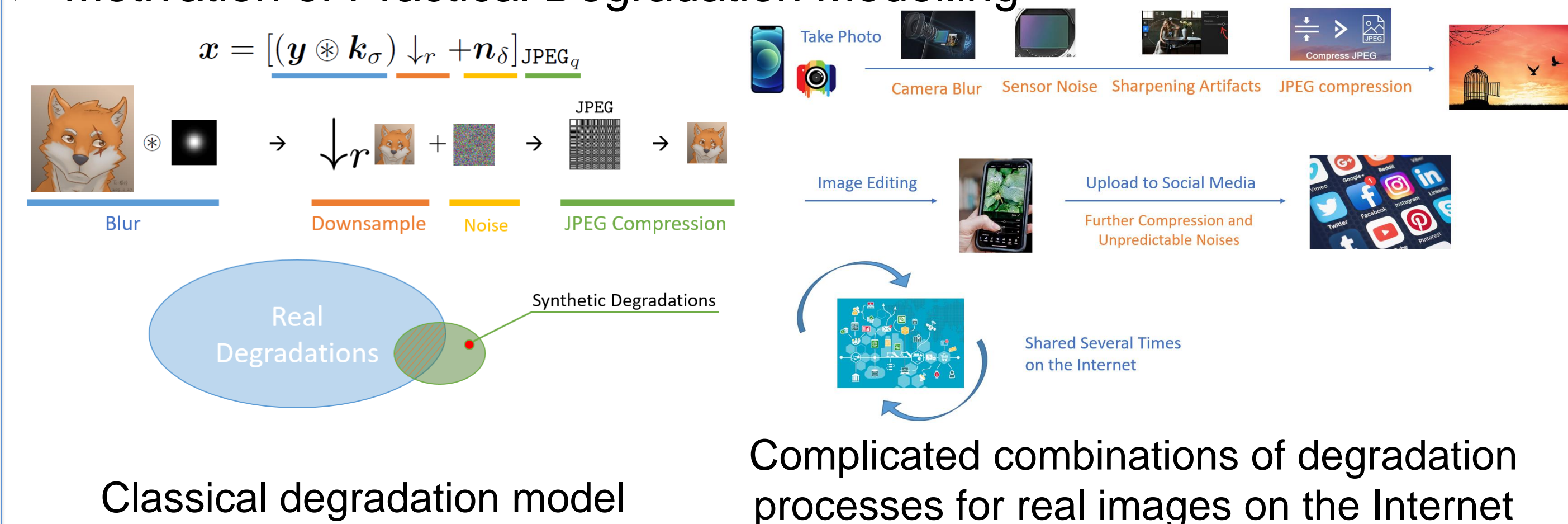
- Unknown and complex degradations
- Different and various contents
- Deal with them in one unified network



### Contributions

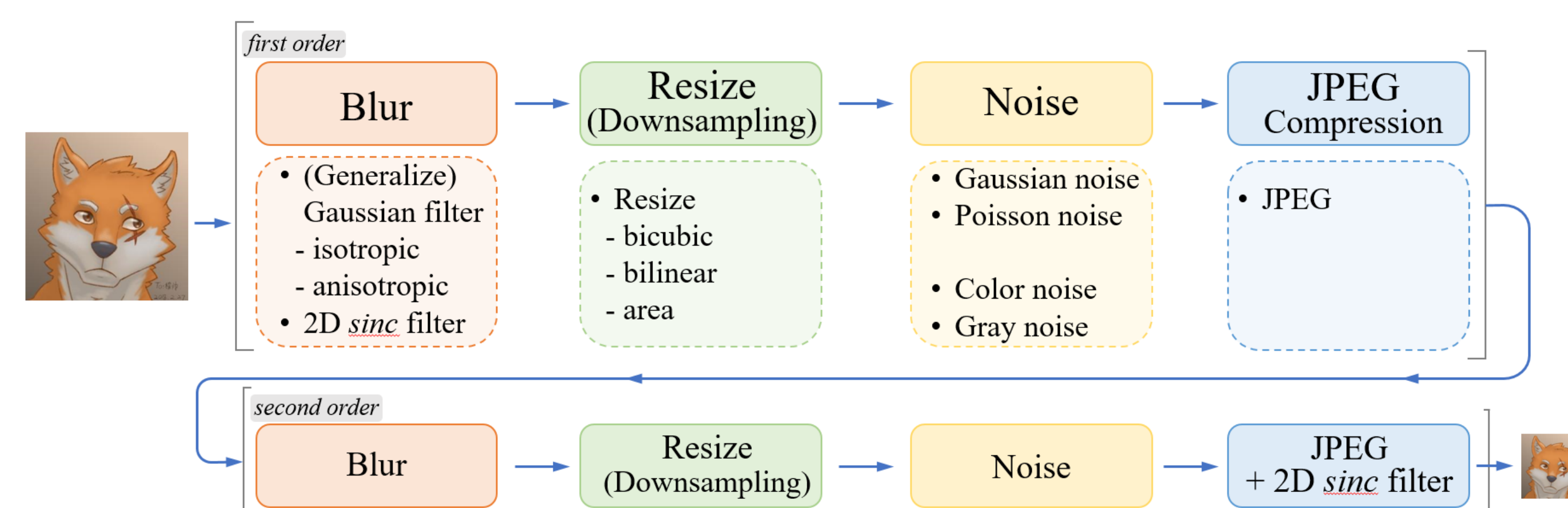
- Propose a **high-order degradation** process to model practical degradations, and utilize *sinc* filters to model common **ringing and overshoot artifacts**.
- Employ **several essential modifications** (e.g., U-Net discriminator with spectral normalization) to increase discriminator capability and stabilize the training dynamics.
- Real-ESRGAN trained with pure synthetic data is able to restore most real-world images and achieve better visual performance than previous works, making it more **practical in real-world applications**.

### Motivation of Practical Degradation Modelling

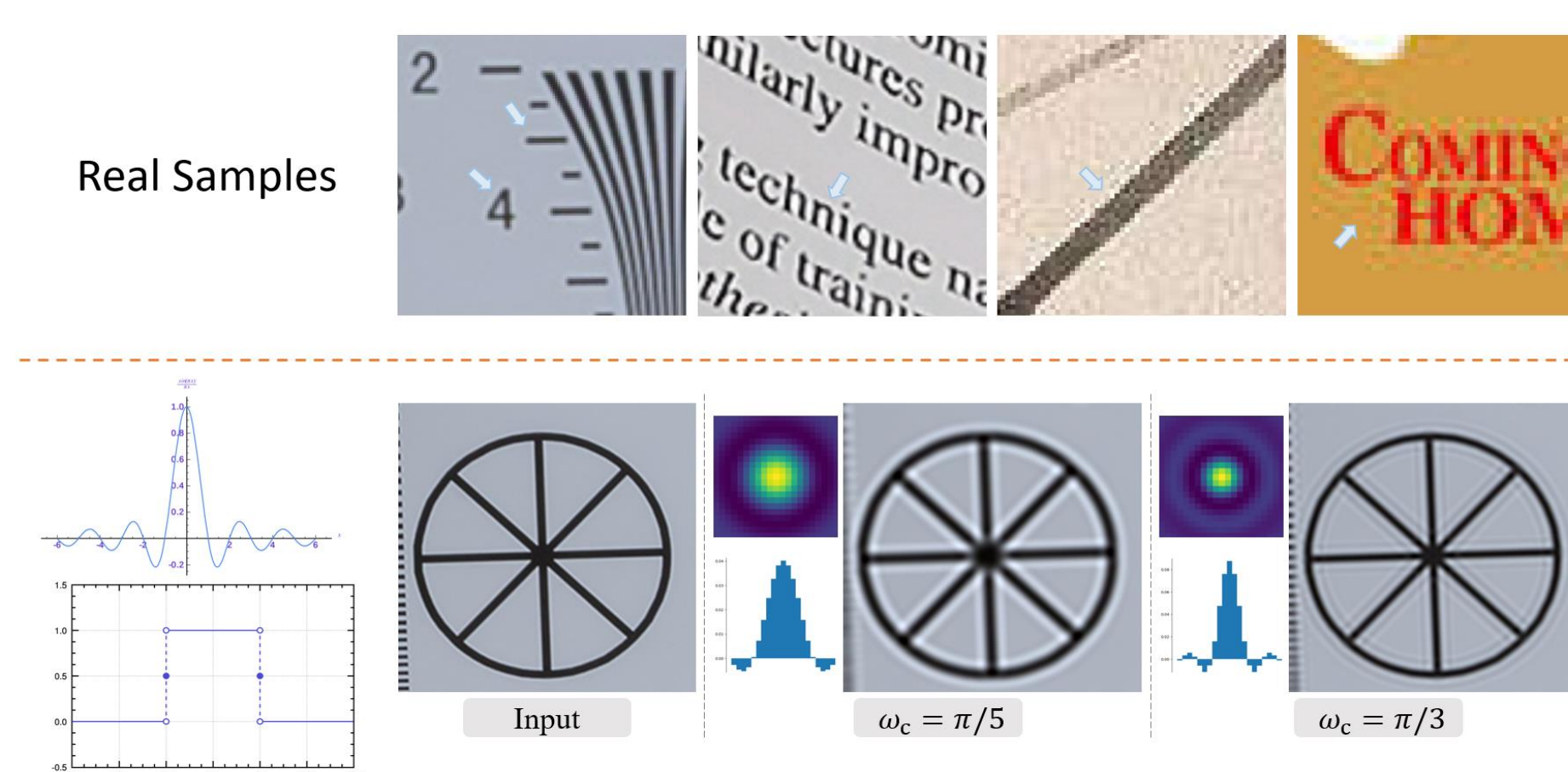


## Real-ESRGAN

### High-Order Degradation Process

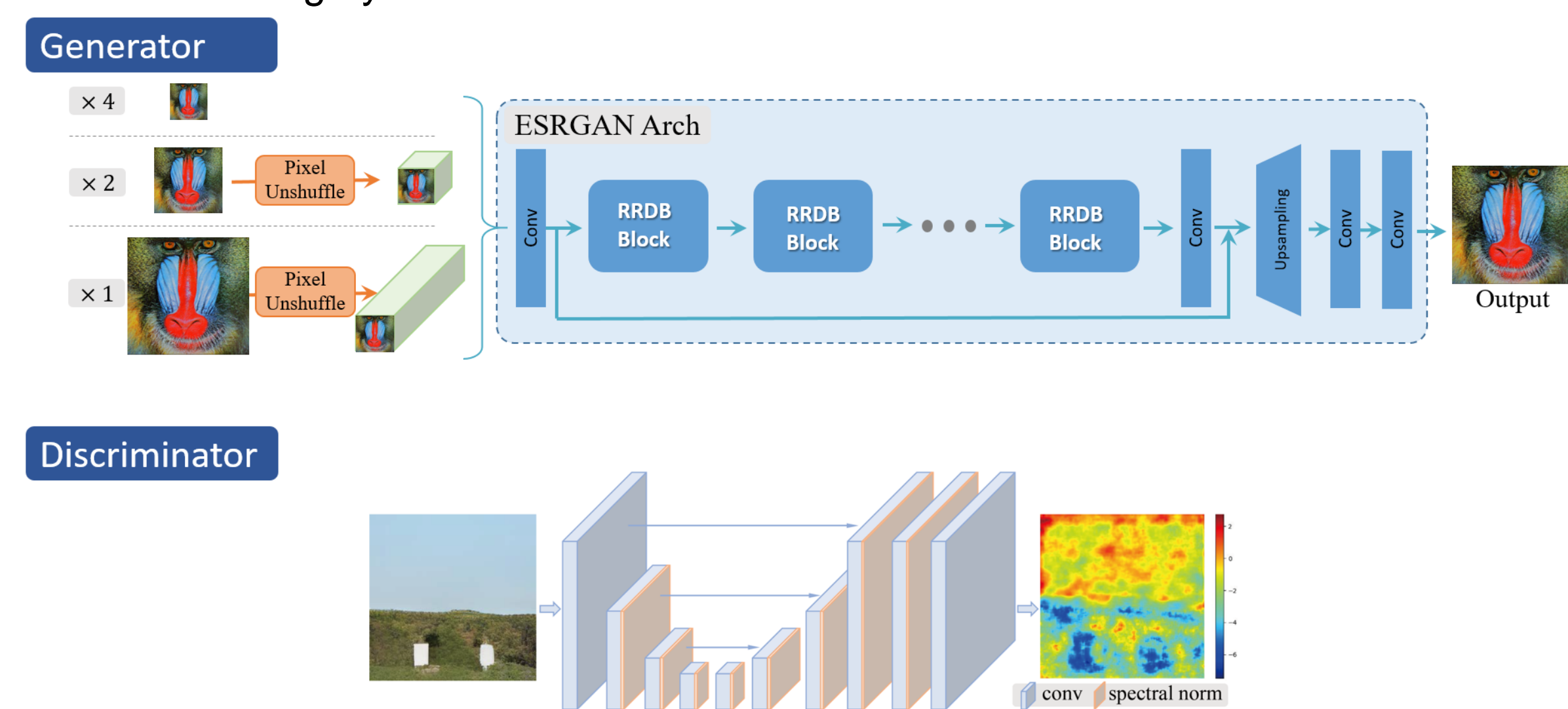


### The sinc Filters to Model Common Ringing and Overshoot Artifacts.



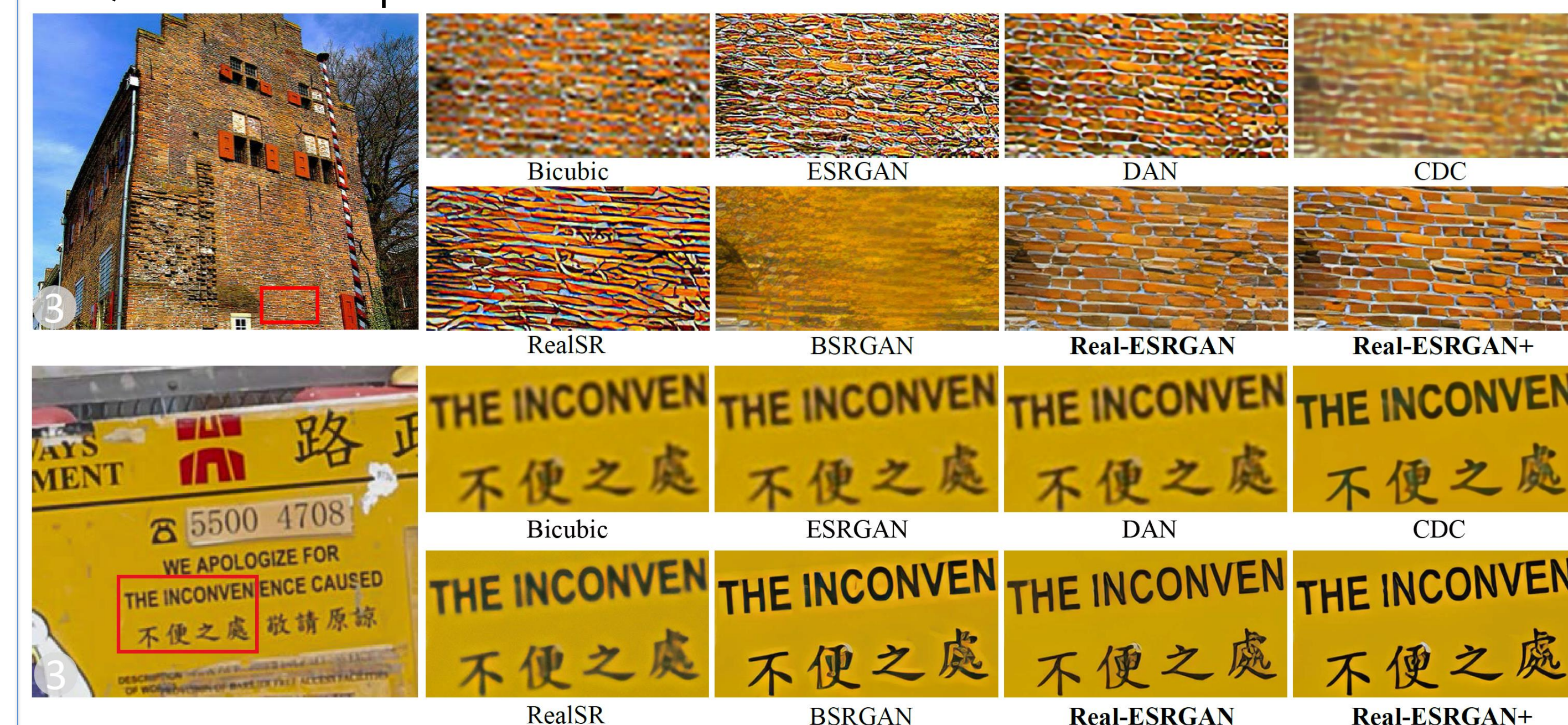
### Network Architecture

- 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

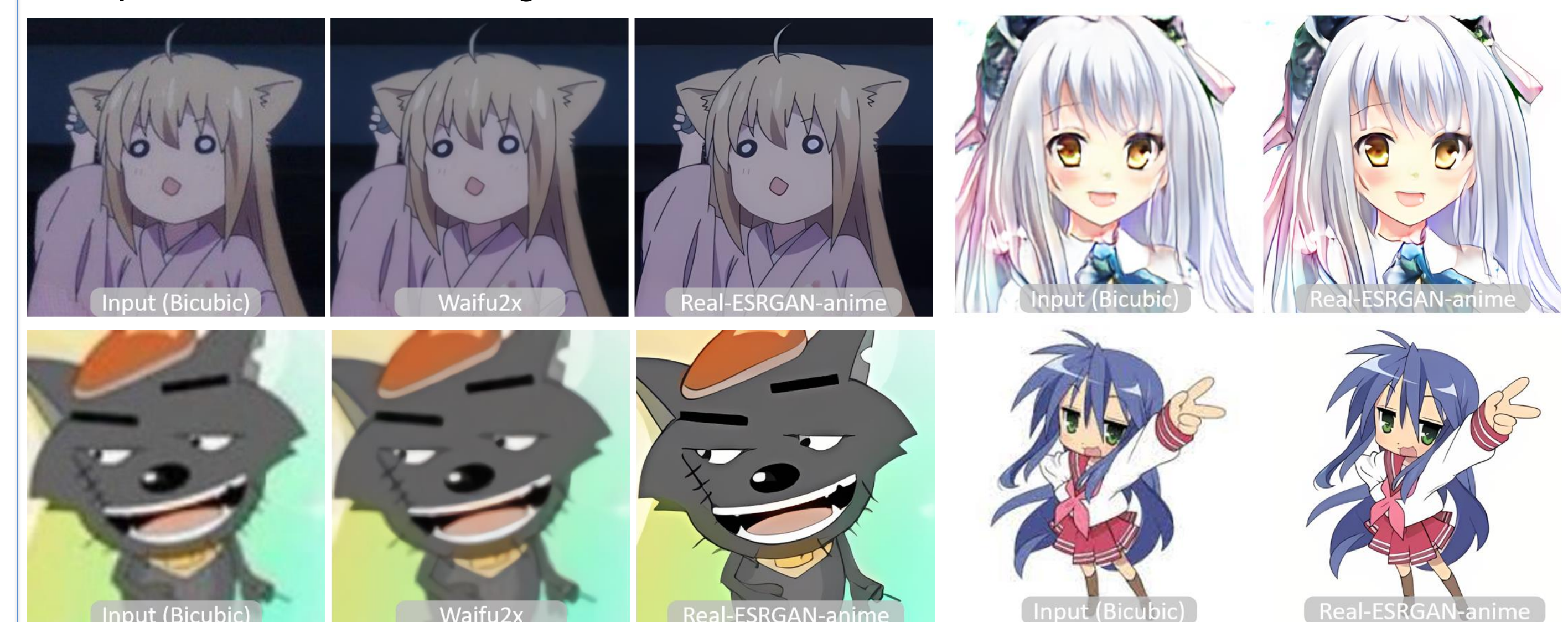


## Results and Open Source

### Qualitative Comparisons



### Optimize for Anime Images



### Open Source

In the GitHub, we provide:

- Full **training and testing** codes
- **Colab Demo** for Real-ESRGAN [Open in Colab](#)
- Portable **Windows / Linux / MacOS executable files** for Intel/AMD/Nvidia GPU, which is based on *Tencent ncnn*

We also incorporate the face restoration method – **GFPGAN**, to improve the face performance.



BasicSR



Codes & Models