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**Image Super-Resolution (ISR)** : A well-established low-level vision task whose objective is to generate a High-Resolution (HR) image (X) from the given corresponding Low-Resolution (LR) observation (y).

- The current state of the art methods maintain ISR accuracy while being computationally complex and heavily data driven..

**Goal of this paper:** An efficient and computationally lightweight Image Super-Resolution (ISR) model suitable for practical applications.

#### Related work:

- Zero-shot approaches using Self-supervised learning

#### Key feature of this work:

- The task of learning the ISR problem from image data is transformed to learning the "identity ( $\delta$ )" between degradation model ( $K$ ) and its inverse ( $K^{-1}$ ), based on the linear system concepts.
- This work detours typical degradation kernel estimation step in ISR framework.

#### Loss function - Deep Identity Learning

$$\text{Loss (L)} = \| K * K^{-1} - \delta \|_2 + R$$

Where

$$R = \lambda_1 \times L_{\text{ConvArea}} + \lambda_2 \times L_{\text{Center}}$$

$$L_{\text{ConvArea}} = |1 - \sum_{i,j} K^{-1}_{i,j}|$$

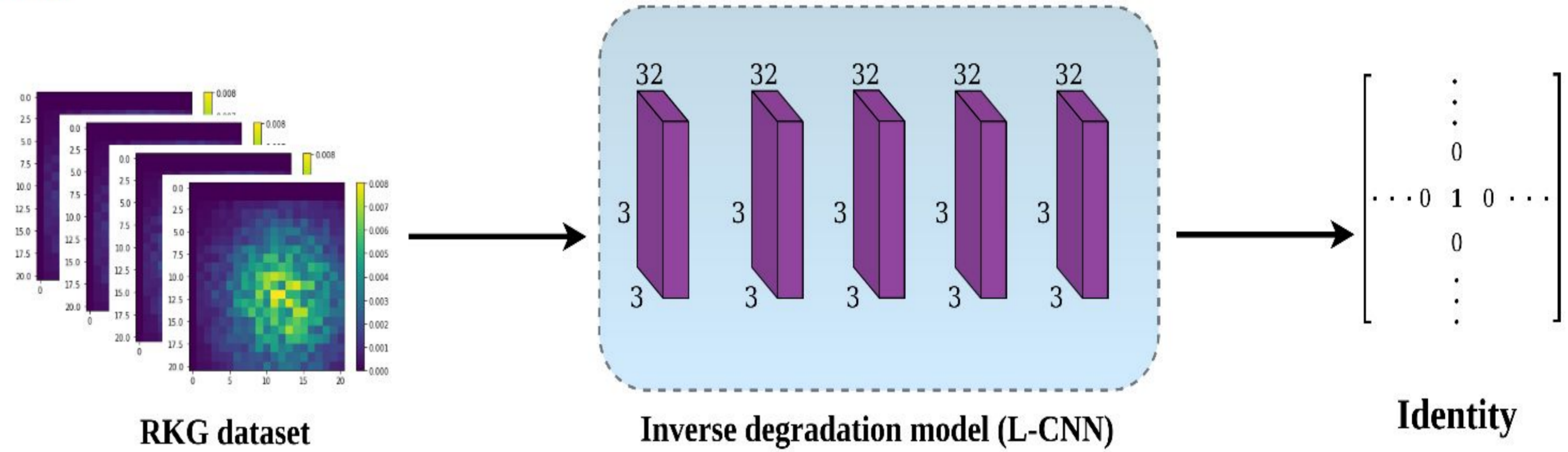
$$L_{\text{Center}} = |1 - K \cdot (K^{-1})^T|$$

$$\text{Degradation model: } y = (X * K) \downarrow_s$$

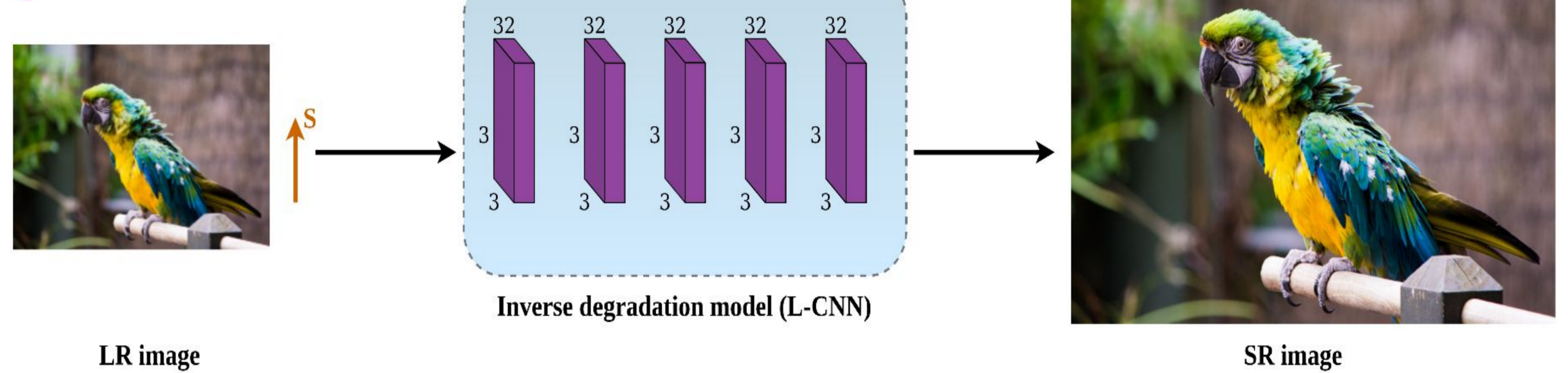
$$\text{Inverse Degradation model: } X = y \uparrow^s * K^{-1}$$

$$\text{Identity relation: } K * K^{-1} = \delta$$

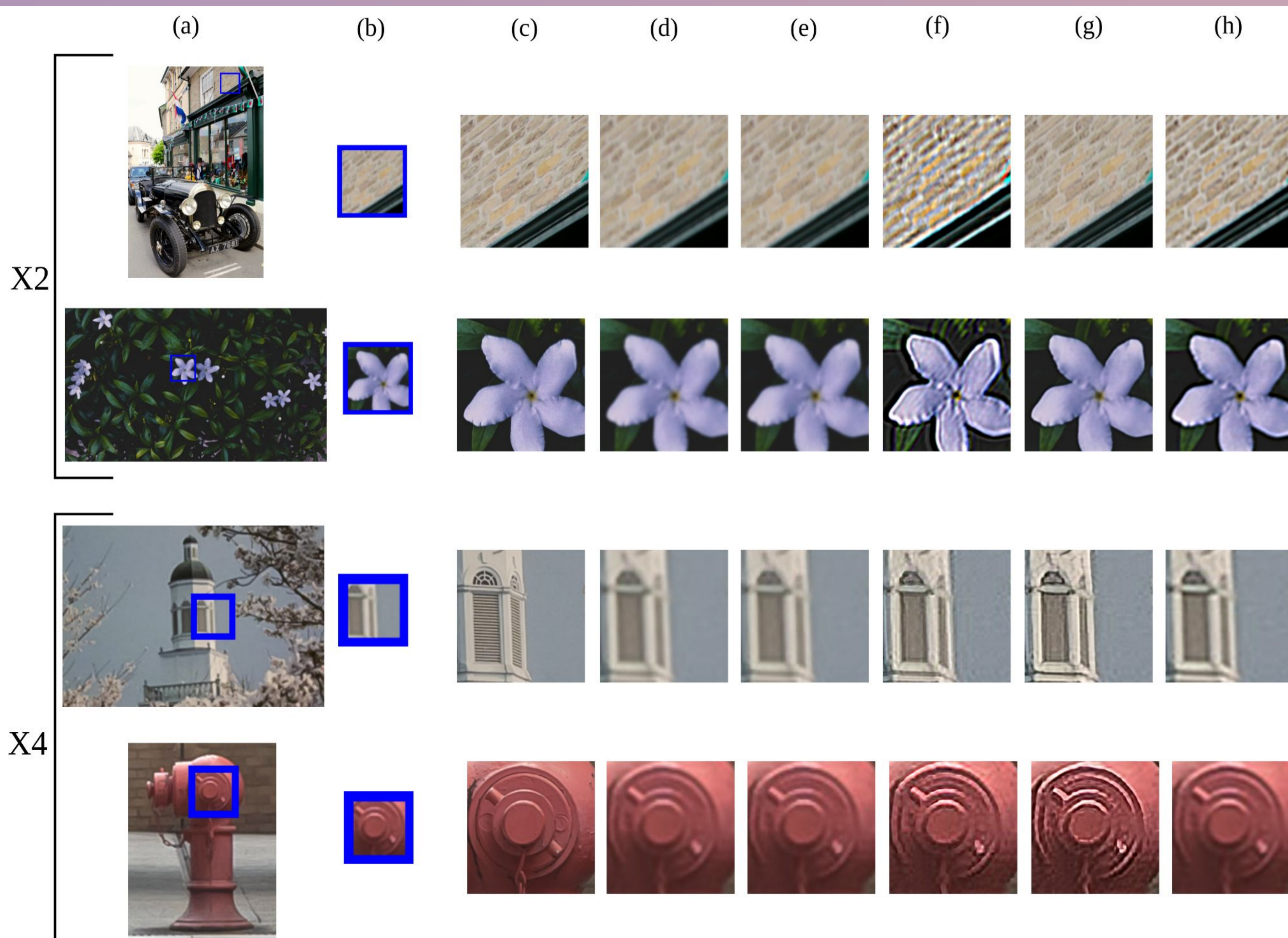
#### Training:



#### Testing:



ISR Method (DIV2KRRK dataset)	#Parameters (Millions) / Inference time (Minutes)	X2	X4
ZSSR	0.29 / >= 10	4.111/0.7925/27.51	4.156/0.6550/24.05
KernelGAN + ZSSR	0.151 + 0.29 / >= 13	4.071/0.8379/28.24	4.089/0.6799/24.76
DBPI	0.5 / >= 1	4.049/0.8684/30.77	4.146/0.7368/26.86
DualSR	0.45 / >= 3.5	5.005/0.8538/29.38	-/-/-
NSSR-DIL (Ours)	0.028 / 0.005	4.161/0.8644/26.02	4.170/0.7926/23.58



#### Conclusions:

- The first image data independent DL-based ISR model.
- Addresses the dataset availability, a challenge that is common to DL frameworks.
- A computationally efficient ISR model that super-resolves given LR image, on par with SotA methods, in fraction of second. Hence more suitable for practical applications.

[ZSSR] Assaf Shocher et al. "zero-shot" super-resolution using deep internal learning". In CVPR 2018.

[KernelGAN] Sefi Bell-Kligler et al. "Blind super-resolution kernel estimation using an internal-gan", in NeurIPS 2019.

[DBPI] Jonghee Kim et al. "Dual back-projection-based internal learning for blind super-resolution". IEEE Signal Processing Letters 2020.

[DualSR] Mohammad Emad et al. "Dualsr: Zero-shot dual learning for real-world super-resolution" in WACV 2021.

#### More details in our paper

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