



# UniDemoiré: Towards Universal Image Demoiréing with Data Generation and Synthesis

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## Motivation

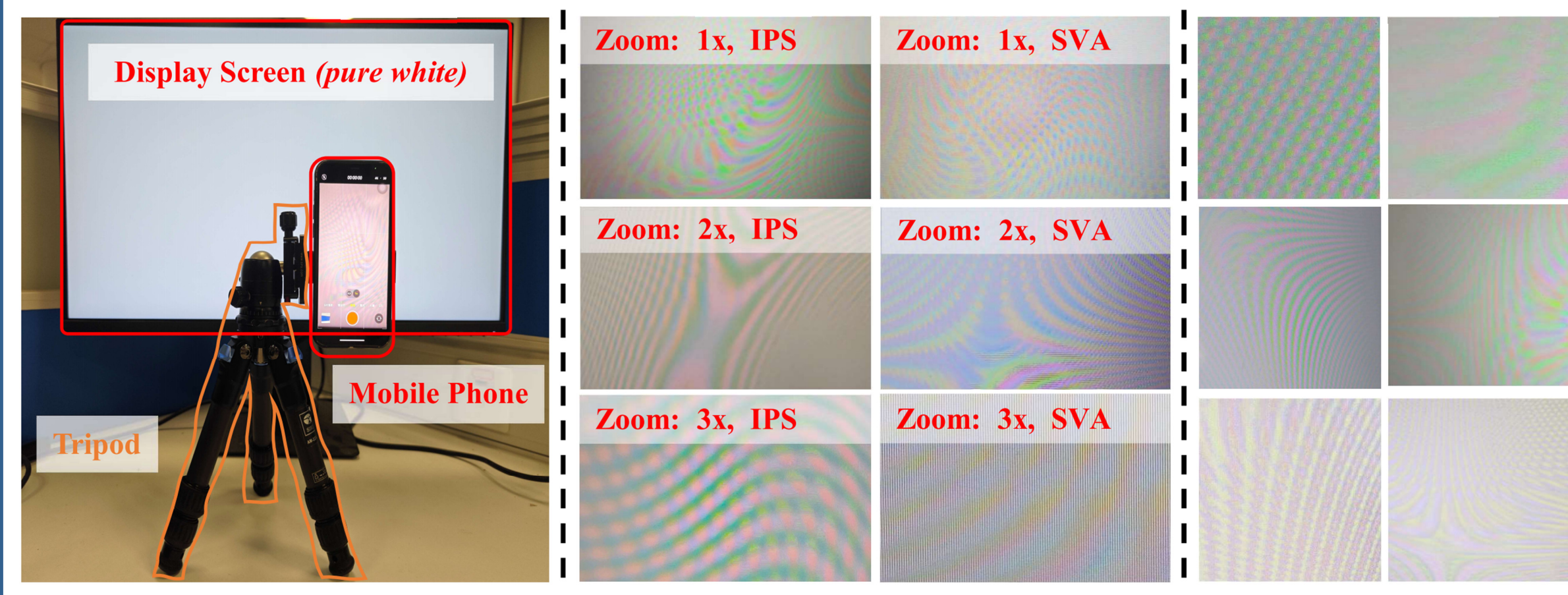
- Limited Data:** Existing methods overfit to single moiré domains
- Real-Synthetic Gap:** Handcrafted/retrieved patterns ≠ Real-world complexity
- Universal Demand:** "A single model for ALL moiré types!"

## Core Features of our UniDemoiré:

- 150K+ Pure Moiré Patterns**
- Diverse & Realistic Moiré Synthesis**
- Diffusion-Based Pattern Generation**
- Zero-Shot Generalization**

## Moiré Pattern Dataset

"White screen + Phone videos → 150K high-res moiré patterns"



Datasets		Avg. Resolution	Size	Capture settings				
Type	Name			Phone	Screen	Multi-zooming rate	Multi-camera / CMOS	Screen Panel
Moiré Image Dataset	TIP2018(R)	256 × 256	135000	3	3	X(1x-only)	X(Main-only)	IPS-only
	FHDMi(R)	1024 × 1024	12000	3	2	X(1x-only)	X(Main-only)	IPS-only
	UHDMi(R)	4328 × 3248	5000	3	3	X(1x-only)	X(Main-only)	IPS-only
Moiré Pattern Dataset	LCDMoiré(S)	1024 × 1024	10200	-	-	-	-	-
	<b>Ours(R)</b>	<b>3840 × 2160</b>	<b>150000</b>	<b>6</b>	<b>6</b>	<b>√(1x,2x,3x)</b>	<b>√(Main,Telephoto)</b>	<b>IPS, SVA</b>

## Moiré Pattern Generation

### Stage 1: Multi-scale cropping + Colorfulness-Sharpness Selection

**Algorithm 1: Data Preprocessing in MPG.**

**Input:** 4K Moiré Pattern Dataset  $\mathcal{D}_{mp}$ , patch size  $(w, h)$ .

**Output:** Selected moiré pattern patch  $I_{mp}$ .

```

while True do
  Randomly select a 4K moiré pattern  $\mathcal{I} \in \mathcal{D}_{mp}$ .
  for  $i = 1$  to  $n$  do
    1. Multi-Scale Cropping:
      Randomly select probability  $p_1, p_2$ .
      if  $p_1 \leq 50\%$  then
         $I_{mp} \leftarrow$  Random Crop( $\mathcal{I}, w, h$ ).
      else if  $p_2 \leq 33.33\%$  then
         $\mathcal{I} \leftarrow$  Resize( $\mathcal{I}, 2560, 1440$ ).
         $I_{mp} \leftarrow$  Random Crop( $\mathcal{I}, w, h$ ).
      else if  $33.33\% < p_2 \leq 66.66\%$  then
         $\mathcal{I} \leftarrow$  Resize( $\mathcal{I}, 1920, 1080$ ).
         $I_{mp} \leftarrow$  Random Crop( $\mathcal{I}, w, h$ ).
      else  $I_{mp} \leftarrow$  Resize( $\mathcal{I}, w, h$ ).
    2. Sharpness-Colorfulness selection:
       $G_{mp} \leftarrow$  RGB.to.Gray( $I_{mp}$ ).
       $L_{mp}, A_{mp}, B_{mp} \leftarrow$  RGB.to.LAB( $I_{mp}$ ).
      Sharpness  $\leftarrow \sigma(\mathcal{F} * G_{mp})$ .
      Colorfulness  $\leftarrow \sqrt{\sigma(A_{mp})^2 + \sigma(B_{mp})^2}$ .
      if Sharpness  $\geq \delta_s$  and Colorfulness  $\geq \delta_c$  then
        return  $I_{mp}$ .
  end
end
  
```

## Stage 2: Latent Diffusion Model → Diverse moiré textures

**Denosing Process**  $\times(T-1)$

$z_T$  (Noise) → Denoising U-Net (Attn) →  $z_0$  → Dec. → Moiré Pattern

**Diffusion Process**

$768 \times 768$  → Enc. →  $z_0$  → Diffusion Process →  $z_T$  (Noise)

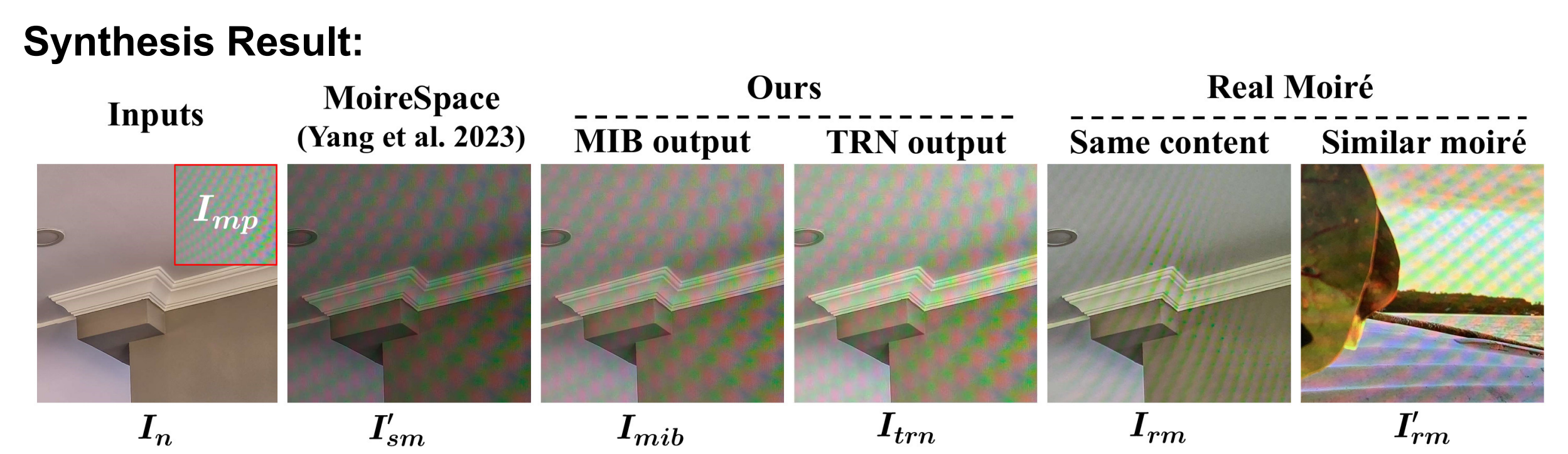
**For VAE, we have:**

$$\mathcal{L} = \min_{\mathcal{E}, \mathcal{D}} \max_{\psi} (\mathcal{L}_{rec}(I_{mp}, \mathcal{D}(\mathcal{E}(I_{mp}))) + \log D_{\psi}(I_{mp}) - \mathcal{L}_{adv}(\mathcal{D}(\mathcal{E}(I_{mp}))) + \mathcal{L}_{reg}(I_{mp}; \mathcal{E}, \mathcal{D}))$$

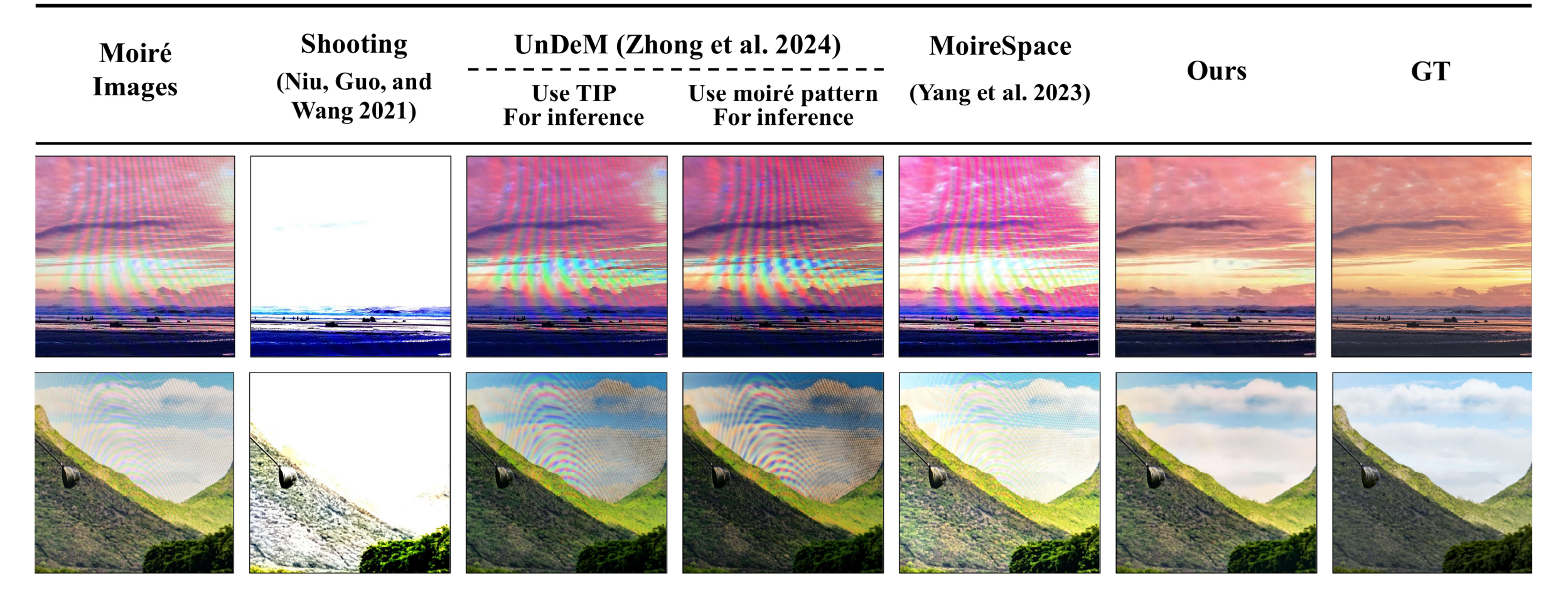
**For Diffusion, we have:**

$$\mathcal{L} = \mathbb{E}_{\mathcal{E}(I_{mp}), \epsilon, t} [\|\epsilon - \epsilon_{\theta}(\alpha_t \mathcal{E}(I_{mp}) + \sigma_t \epsilon, t)\|_1]$$

## Qualitative Results



## Demoiréing Result:



## Moiré Image Synthesis

### Stage 1: Moiré Image Blending (training free)

$I_n$  and  $I_{mp}$  → Grain Merge → Multiply Merge → Alpha Blending →  $I_{mib}$

$I_{mib} = \omega_g I_n + \omega_m I_{mp}$

$I_{mib} = (I_{mp} + I_n - 0.5) \odot I_n$

### Stage 2: Tone Refinement Network

$I_{mib}$  → Transform → RGB-uv Histogram →  $H(I_{rm})$  → Histogram Feature Loss →  $H(I_{trn})$  → Transform → RGB-uv Histogram →  $I_{trn}$

**Perception Loss** and **TV Loss**

**Mixed feature statistics:**

$$\gamma_{mix} = \lambda \sigma(f^{(k)}) + (1 - \lambda) \sigma(f_r^{(k)})$$

$$\beta_{mix} = \lambda \mu(f^{(k)}) + (1 - \lambda) \mu(f_r^{(k)})$$

**Fusion output:**

$$\gamma_{mix} \odot \frac{f^{(k)} - \mu(f^{(k)})}{\sigma(f^{(k)})} + \beta_{mix}$$

## Experiment Results

### Zero-Shot Demoiréing Comparison:

Test Dataset	Metric	Demoiréing Network: MBCNN					Demoiréing Network: ESDNet-L				
		Shooting	UnDeM <sup>†</sup>	UnDeM <sup>‡</sup>	MoireSpace	Ours	Shooting	UnDeM <sup>†</sup>	UnDeM <sup>‡</sup>	MoireSpace	Ours
UHDMi	PSNR ↑	9.2284	13.4256	14.5237	14.7826	<b>17.9162</b>	10.2568	15.2269	15.2947	14.7989	<b>17.2543</b>
	SSIM ↑	0.5180	0.3973	0.4425	0.4724	<b>0.6280</b>	0.5664	0.5873	0.5777	0.4859	<b>0.6454</b>
	LPIPS ↓	0.6664	0.6489	0.6332	0.5568	<b>0.4162</b>	0.5130	0.4190	0.4241	0.5254	<b>0.3238</b>
FHDMi	PSNR ↑	10.6750	17.8355	18.1652	18.5523	<b>19.0094</b>	11.6022	18.4335	18.5390	18.0763	<b>19.8128</b>
	SSIM ↑	0.4478	0.6802	0.6999	0.7094	<b>0.7137</b>	0.5425	0.6900	0.6812	0.7189	<b>0.7319</b>
	LPIPS ↓	0.5978	0.2606	0.2472	0.2742	<b>0.2390</b>	0.4515	0.2877	0.2986	0.2616	<b>0.2134</b>

### Cross-Dataset Demoiréing Comparison:

Cross Dataset	Source	Target	Metric	Demoiréing Network: MBCNN					Demoiréing Network: ESDNet-L				
				Baseline	Shooting	UnDeM	MoireSpace	Ours	Baseline	Shooting	UnDeM	MoireSpace	Ours
UHDMi	FHDMi	PSNR ↑	19.3848	19.2032	19.4676	19.4531	<b>19.8625</b>	20.3422	20.2407	20.4014	20.2806	<b>20.7543</b>	
			SSIM ↑	0.7436	0.7459	0.7455	0.7496	<b>0.7525</b>	0.7599	0.7579	0.7510	0.7603	<b>0.7653</b>
			LPIPS ↓	0.3002	0.2975	0.2964	0.2993	<b>0.2842</b>	0.2525	0.2632	0.2509	0.2324	<b>0.2136</b>
UHDMi	TIP	PSNR ↑	17.8107	18.3730	18.6674	18.9214	<b>19.3922</b>	18.8040	18.4543	19.3545	19.3964	<b>19.5009</b>	
			SSIM ↑	0.6627	0.6888	0.6911	0.6996	<b>0.7022</b>	0.6921	0.6930	0.6998	0.7111	<b>0.7149</b>
			LPIPS ↓	0.3580	0.3886	0.3909	0.3829	<b>0.3781</b>	0.3524	0.3849	0.3601	0.3522	<b>0.3495</b>
FHDMi	UHDMi	PSNR ↑	17.1331	17.5326	17.4870	17.6050	<b>18.7931</b>	18.0049	18.4189	17.9574	17.9751	<b>18.9240</b>	
			SSIM ↑	0.6159	0.6334	0.6331	0.6642	<b>0.7186</b>	0.5755	0.5780	0.5857	0.5548	<b>0.6658</b>
			LPIPS ↓	0.4470	0.4350	0.4285	0.4020	<b>0.3508</b>	0.4420	0.4279	0.4460	0.4579	<b>0.3405</b>
FHDMi	TIP	PSNR ↑	20.2161	20.7793	20.8261	20.1194	<b>21.0694</b>	20.6647	20.8678	20.4663	20.8107	<b>21.5786</b>	
			SSIM ↑	0.7340	0.7304	0.7381	0.7347	<b>0.7494</b>	0.7504	0.7606	0.7278	0.7582	<b>0.7668</b>
			LPIPS ↓	0.2979	0.2884	0.2891	0.2961	<b>0.2832</b>	0.2459	0.2450	0.2998	0.2468	<b>0.2310</b>
TIP	UHDMi	PSNR ↑	17.3409	17.4011	17.4407	17.4987	<b>18.2937</b>	17.4332	16.1836	16.8402	16.6296	<b>18.4978</b>	
			SSIM ↑	0.6144	0.6062	0.6066	0.6059	<b>0.6913</b>	0.5523	0.5511	0.5692	0.5748	<b>0.6866</b>
			LPIPS ↓	0.4726	0.4487	0.4473	0.4412	<b>0.3990</b>	0.4987	0.4723	0.4532	0.4387	<b>0.3231</b>
FHDMi	UHDMi	PSNR ↑	18.9458	19.2731	19.0336	19.1101	<b>20.1053</b>	19.2368	18.1936	19.2112	18.8385	<b>19.9971</b>	
			SSIM ↑	0.7369	0.7399	0.7215	0.7321	<b>0.7725</b>	0.7354	0.7399	0.7499	0.7389	<b>0.7580</b>
			LPIPS ↓	0.2494	0.2447	0.2452	0.2382	<b>0.2315</b>	0.2316	0.2320	0.2130	0.2228	<b>0.1915</b>