

A Well-aligned Dataset for Learning Image Signal Processing on Smartphones from a High-end Camera

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PROBLEM

- Not every camera is equipped with an excellent image signal processing (ISP) pipeline that converts raw sensor data into color images.
- It is labor-intensive and challenging to design an ISP pipeline with many independent modules, and thus the ISP on most smartphones is sub-optimal, even for the highly-rated ones such as iPhone.

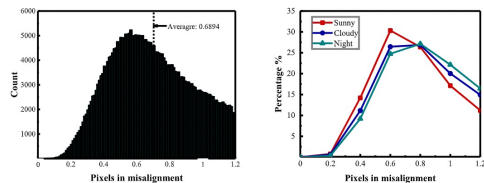
DATA COLLECTION

- Over 2000 pairs for training and testing.



iPhone 6S Nikon Z6 Camera Rig

DATA ANALYSIS



(a) Misalignment distribution (b) Dependence on illumination

We collect a large-scale sub-pixel aligned dataset with raw and RGB data pairs captured by two popular smartphones and one high-end camera. Our dataset can be used for learning ISPs to replace the sub-optimal built-in ISPs of smartphones.



SID [Chen et al. 2018] DPED [Ignatov et al. 2017] PAC [Su et al. 2019] iPhone 6S Ours



iPhone 6S ISP
Ours

APPROACH

Given an n -pixel input image $X = (X_1, X_2, \dots, X_n), X_i \in R^c$, the output $\hat{Y} = (\hat{Y}_1, \hat{Y}_2, \dots, \hat{Y}_n), \hat{Y}_i \in R^{c'}$ can be obtained by convolution operation with per-pixel kernel W :

$$\hat{Y}_i = \sum_{j \in \delta_i} W_i[i-j] X_j + b,$$

where W^i denotes the conditional convolution kernel at position i , δ_i is the neighborhood window centered at position i . In general, W^i should be a function of input content:

$$W^i = f(X, \delta_i).$$

We leverage the approximation power of neural networks to estimate the function f .

RESULTS

We provide the quantitative results with baselines. Overall, all perceptual metrics show that our proposed ISP model outperforms the baselines.

Method	Mi 3			iPhone 6S		
	LPIPS ↓	PSNR ↑	SSIM ↑	LPIPS ↓	PSNR ↑	SSIM ↑
Built-in ISP	0.261	18.82	0.632	0.262	19.65	0.626
DPED	0.474	18.71	0.665	0.453	18.46	0.667
PAC	0.182	20.18	0.699	0.204	18.50	0.650
SID	0.264	20.56	0.690	0.295	21.23	0.731
Ours	0.182	21.22	0.725	0.134	21.09	0.733

REFERENCES

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3. Andrey Ignatov, Nikolay Kobyshev, Radu Timofte, Kenneth Vanhoey, and Luc Van Gool. 2017. DSLR-Quality Photos on Mobile Devices with Deep Convolutional Networks. In International Conference on Computer Vision (ICCV).

Download the Dataset

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