Explorable Tone Mapping Operators

Chien-Chuan Su^{1, 2}



Ren Wang¹



Hung-Jin Lin¹





Yu-Lun Liu¹



Chia-Ping Chen¹



Yu-Lin Chang¹ Soo-Chang Pei²

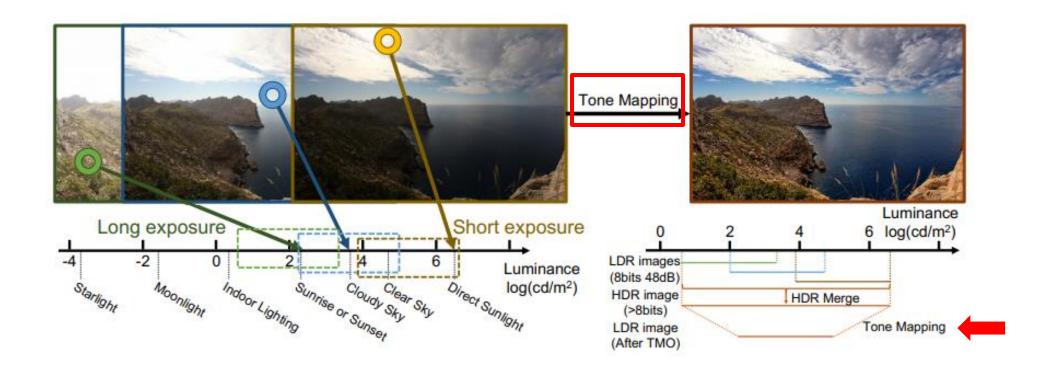


¹MediaTek Inc.

²National Taiwan University

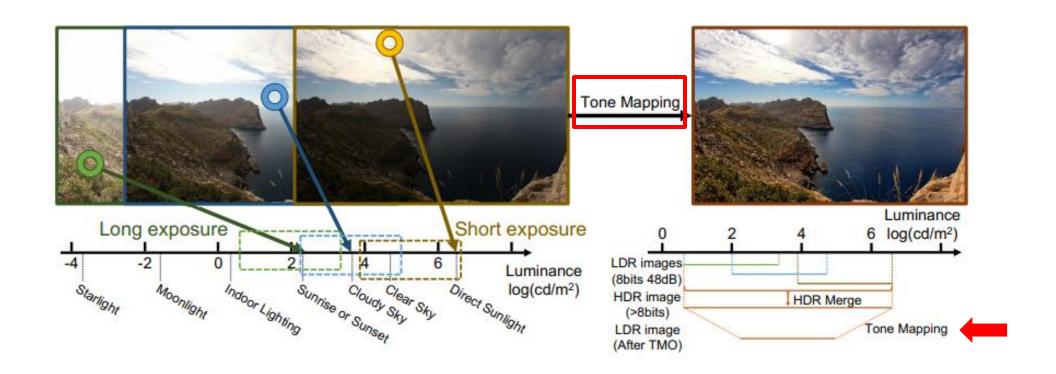
Introduction

What is Tone Mapping Operator (TMO)?



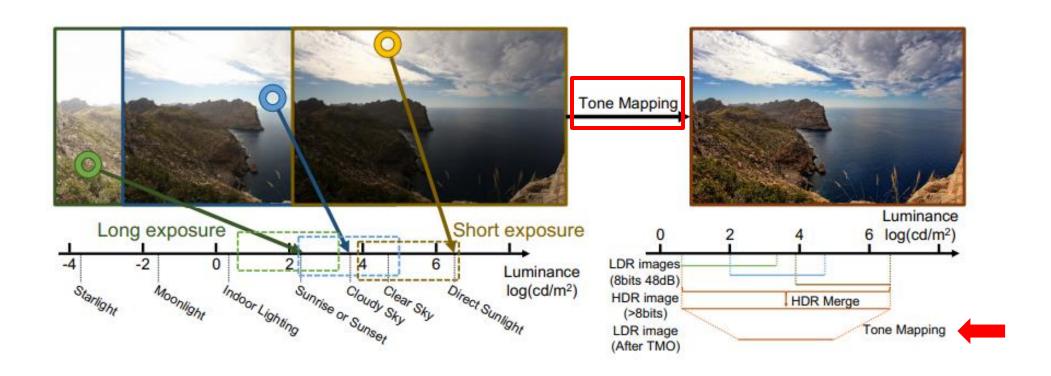
Tone Mapping Operator plays an essential role in high dynamic range (HDR) imaging.

What is Tone Mapping Operator (TMO)?



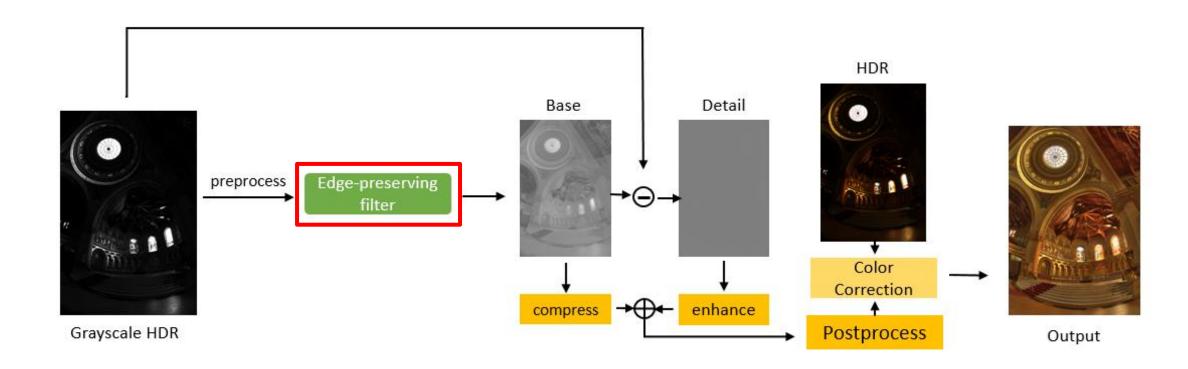
It aims to compress HDR images into low dynamic range (LDR) images so that they can be displayed on monitors.

What is Tone Mapping Operator (TMO)?



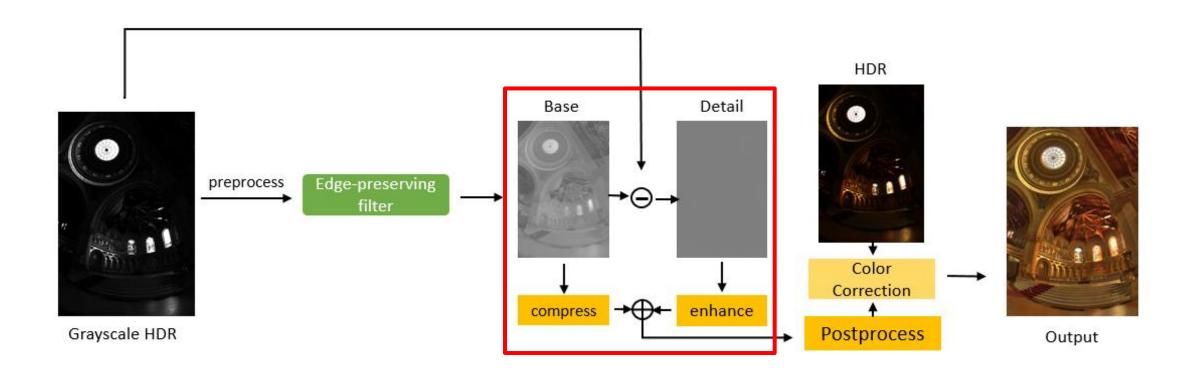
However, compressing HDR images while trying to preserve the perceptual content as much as possible is a challenging task.

Previous Filtering Method



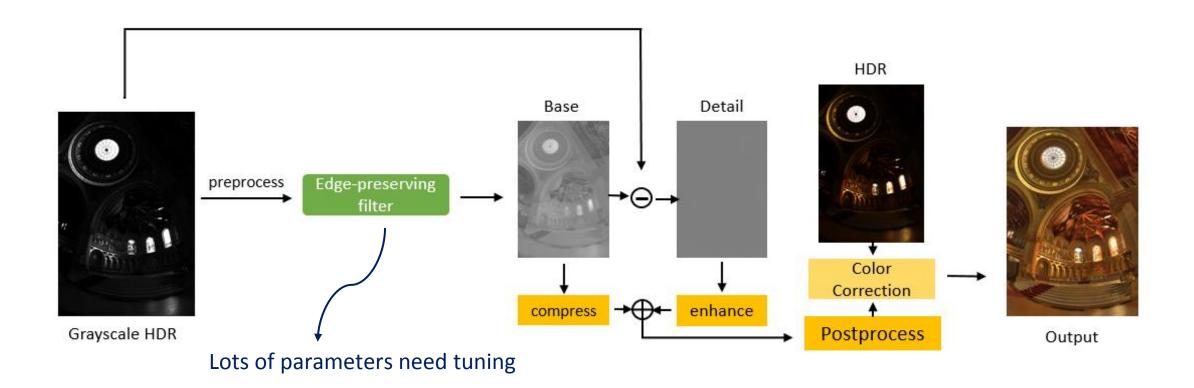
Filtering methods decompose an HDR image into a smooth base image and a detail image by utilizing an edge-preserving filter.

Previous Filtering Method



The base image is usually compressed to reduce the dynamic range, while the detail image is enhanced so that better visual content is preserved.

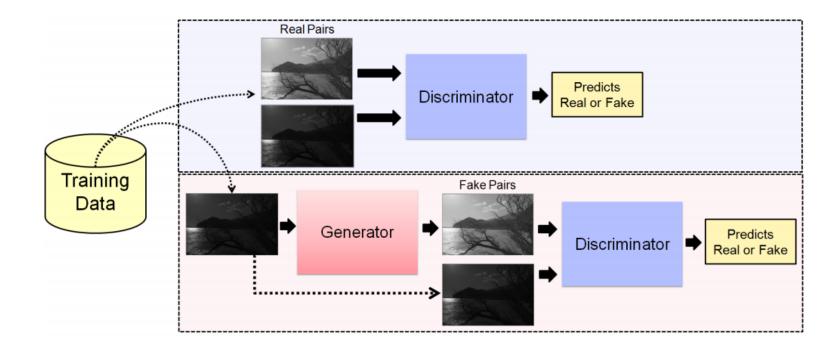
Previous Filtering Method



Nevertheless, due to the large amount of parameters, tuning these filters is usually difficult and time-consuming.

Previous Learning-based Method

Deep TMO^[2]

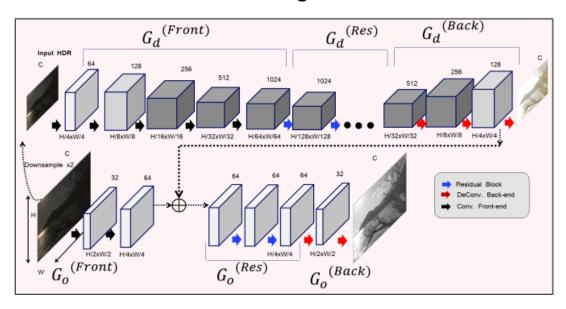


Deep TMO is a deep learning-based method that is able to tone map high-resolution images.

Previous Learning-based Method

Deep TMO^[2]

Multi-scale generator

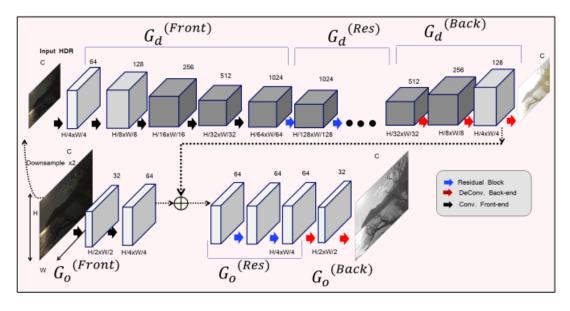


They use cGAN architecture and apply a multi-scale technique to both of the generator and the discriminator to ease artifacts.

Previous Learning-based Method

Deep TMO^[2]

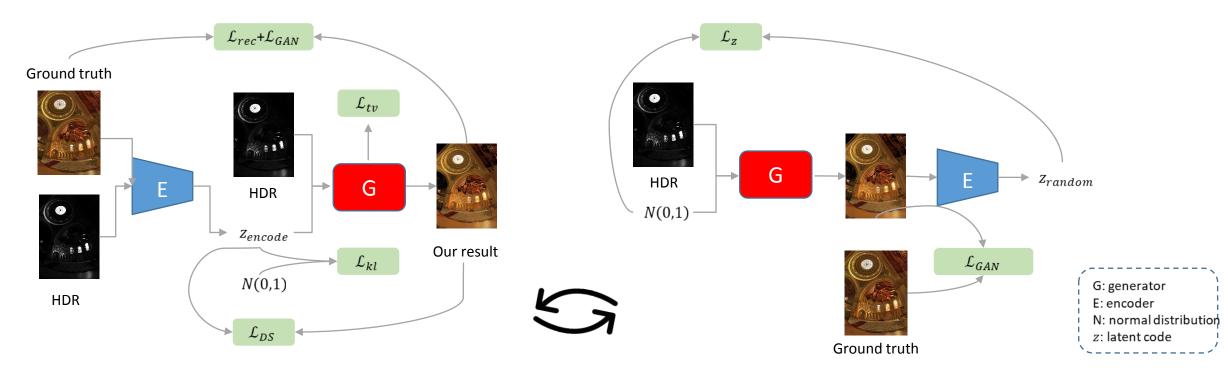
Multi-scale generator



But their method lacks the variety of subjective styles because of the one-to-one mapping characteristic.

Our Method

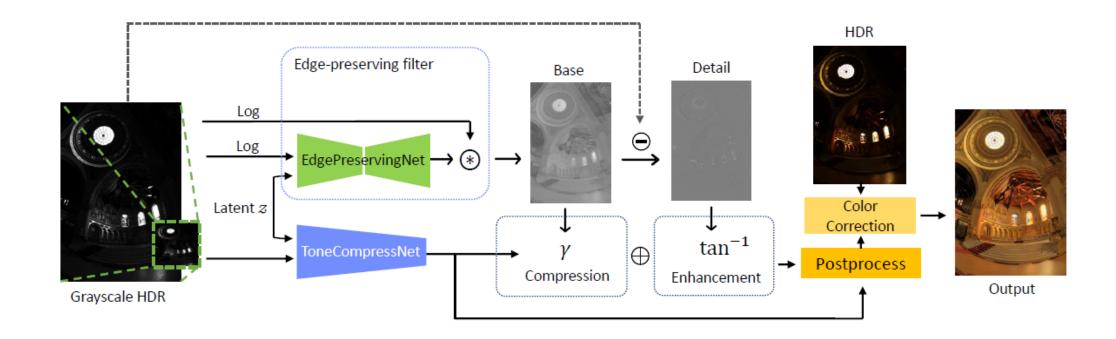
An overview of our tone-mapping method



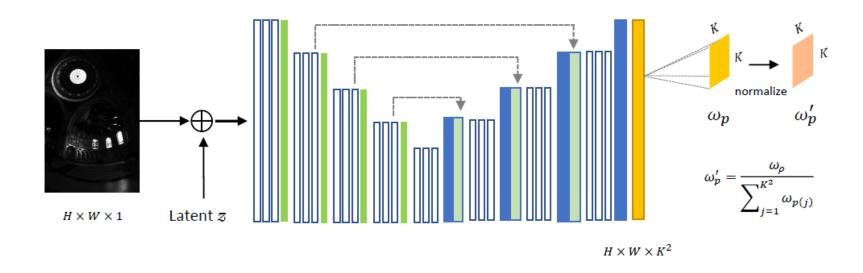
Make generated results follow the target distribution

Make encoded results close to input latent codes

Inspired by the filtering methods, our method is formulated as a BicycleGAN [3] framework.

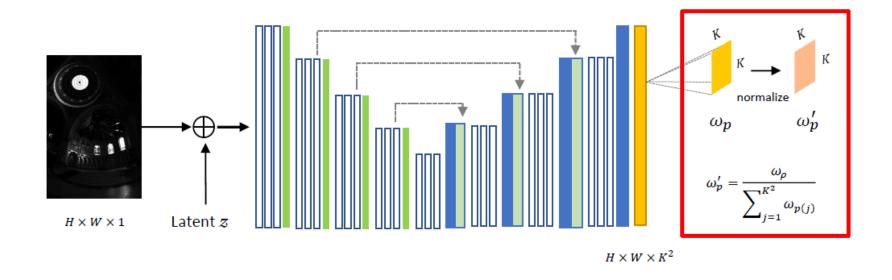


EdgePreservingNet



The EdgePreservingNet is a kernel prediction network ^[4], which is used to decompose HDR images into base images and detail images.

EdgePreservingNet



The predicted kernel is normalized to ensure the consistency of image tone.

EdgePreservingNet

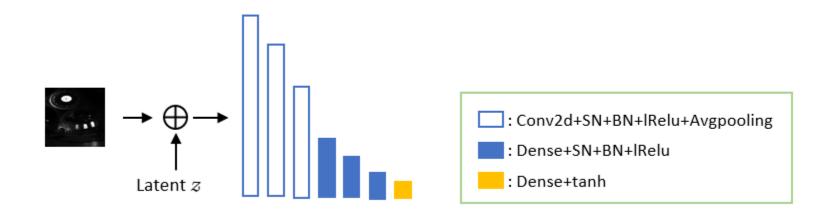


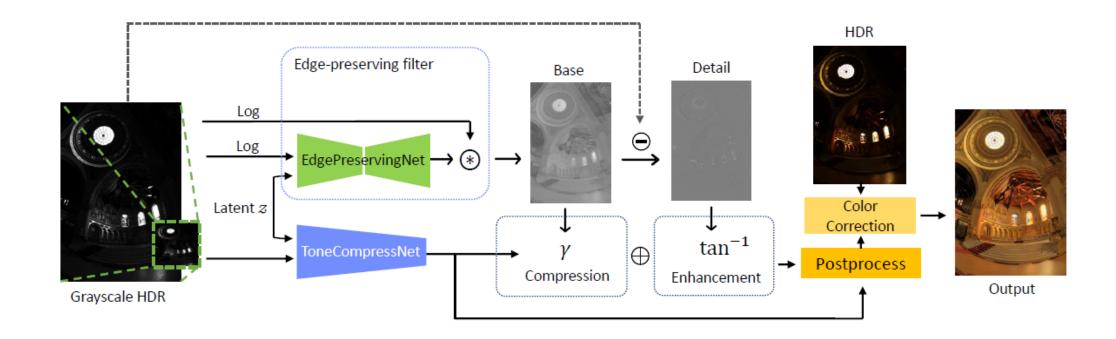
W/o normalization

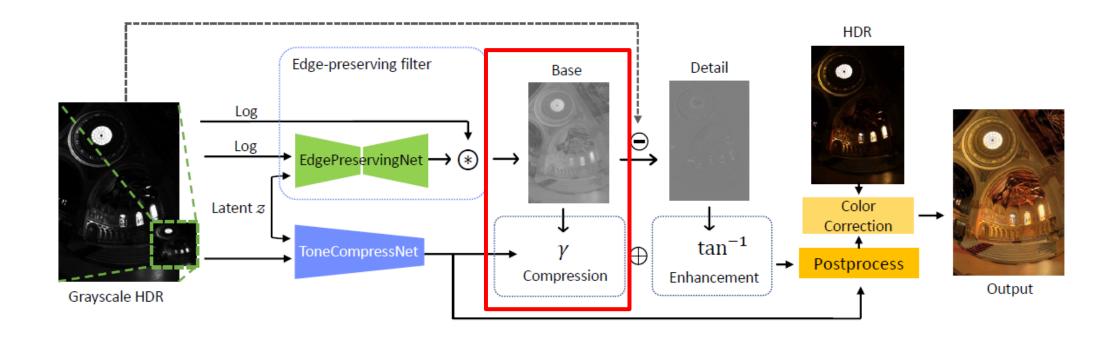


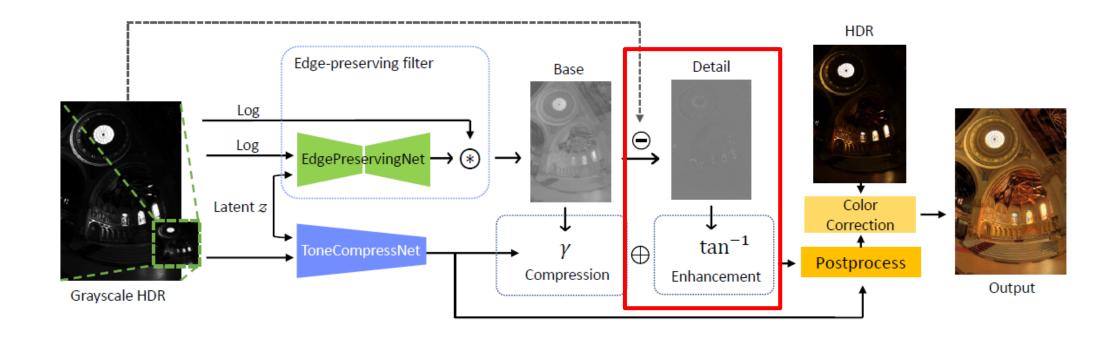
W/ normalization

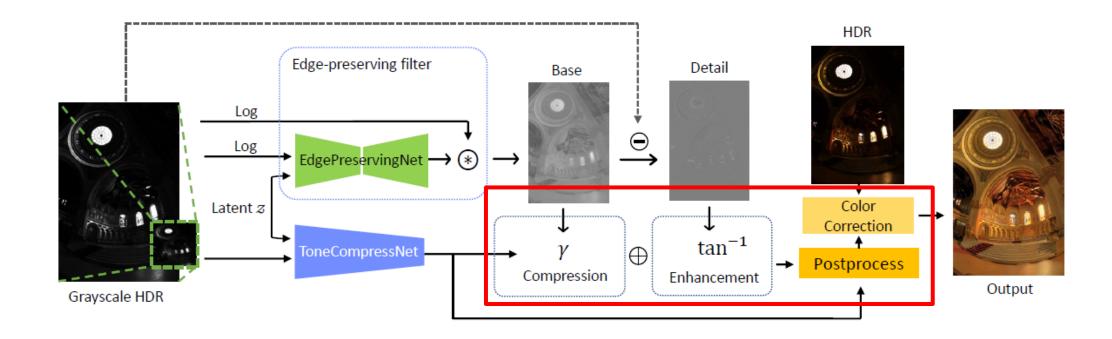
ToneCompressNet



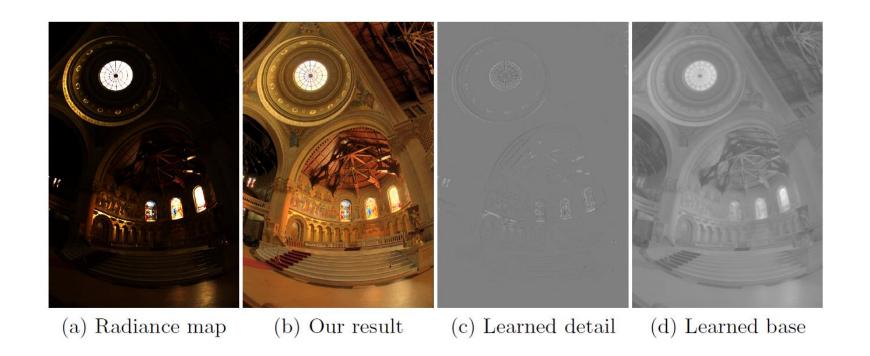








Example of learned decomposed images



Experiments

Training Dataset

- •We collect HDR images from multiple public sources
- ●1032 HDR images with a wide range of contents
 - Scenes
 - Cameras
 - Shooting settings
- Total 3096 HDR-LDR image pairs for training
 - ●Use Luminance HDR^[5] with the default parameters to generate LDR images
 - ●For each HDR image, the LDR images with top-3 highest TMQI^[6] scores are selected as the target images

We collect a dataset from multiple public sources to train our models.

Test Dataset

- Fairchild's dataset^[7]
 - ●105 HDR images
 - Diverse scenes (indoor, outdoor, daylight, night views)

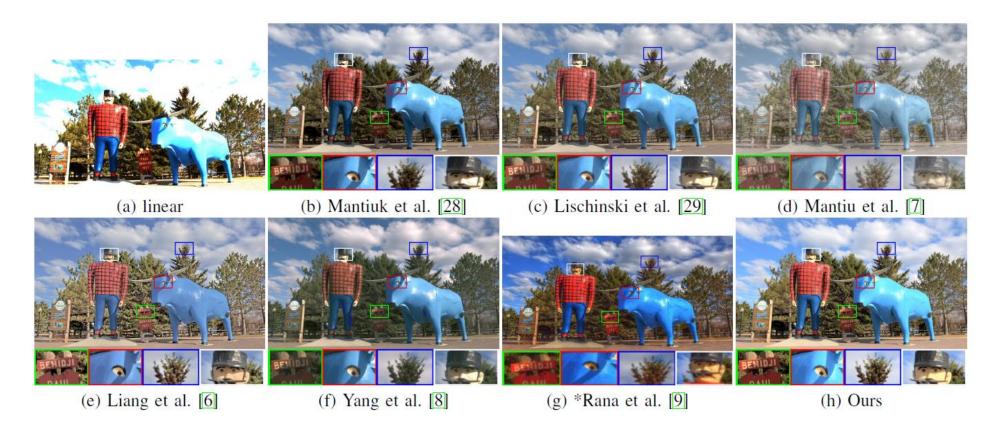






We evaluate our method on the whole Fairchild's dataset of the original resolution (about 4k).

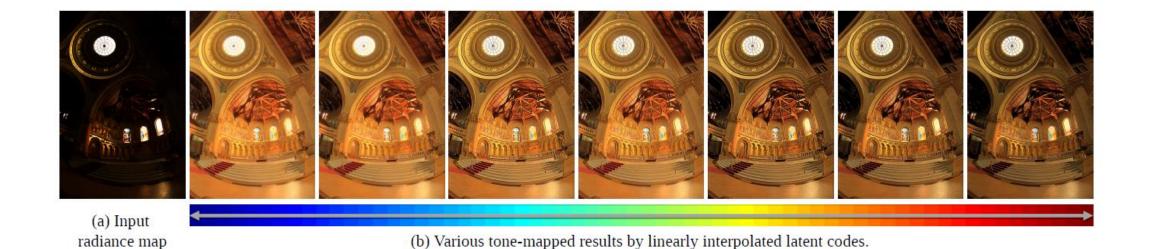
Visual Results



This figure shows that our results preserve the detail of input HDR images without halo artifacts or over enhancement.

The color appearance of our results are also visually appealing.

Diversity in Different Latent Codes



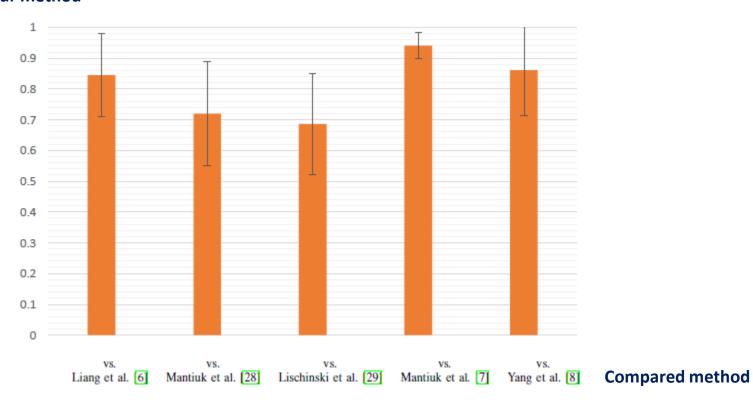
This figure shows the linearity of our results with respect to latent codes.

Quantitative Results

Methods	TMQI	Fidelity	Naturalness
Drago et al. [41]	0.8030	0.7777	0.2289
Fattal et al. [4]	0.8126	0.8217	0.2220
Durand et al. [3]	0.8211	0.7943	0.2993
Mai et al. [42]	0.8184	0.8077	0.2554
Reinhard et al. [5]	0.8282	0.7964	0.3281
Mantiuk 08 et al. [28]	0.8447	0.8336	0.3496
Lischinski et al. [29]	0.8564	0.8381	0.4110
Mantiuk 06 et al. [7]	0.8574	0.8797	0.3412
Liang et al. [6]	0.8650	0.8074	0.5024
Yang et al. [8]	0.8465	0.7564	0.4793
Ours	0.8652	0.8064	0.5096

User Study

Probability of being selected for our method

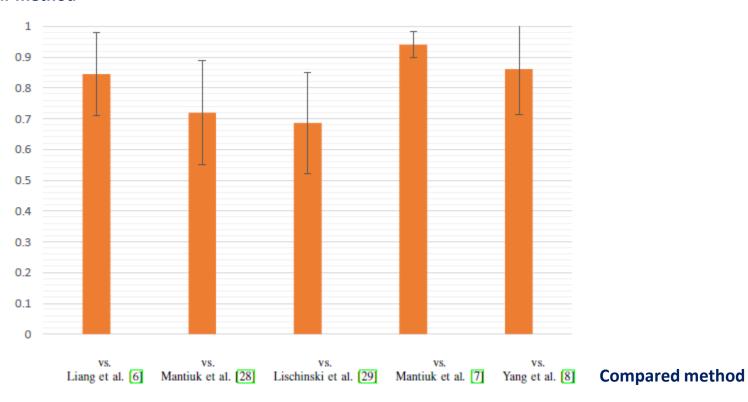


We conduct a user study to further verify the performance of our method.

Totally, 40 subjects are involved in this test.

User Study

Probability of being selected for our method



The result shows that our method achieves higher probability than the compared methods.

Conclusions

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- 3. By leveraging bilateral filters, our method compresses the most part of dynamic range while preserves the high-frequency information of HDR images
- 4. Our method performs favorably against existing methods in terms of both subjective and objective evaluations.

Thanks for your attention!