

Data-intensive Image based Relighting

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Outline

- 1 Motivation
- 2 Data-intensive IBRL
 - Introduction
 - Our Approach
 - Results
 - Conclusion



Overview

1 Motivation

2 Data-intensive IBRL

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Motivation

■ PHOTOREALISM

■ Traditional CG techniques :

- 1 Build a detailed 3D model (geometry)
 - ▶ Very time-consuming to specify the *realistic* 3D model
- 2 Specify the reflectance characteristics
 - ▶ Accurate specification of reflectance properties is difficult
- 3 Specify lighting configurations
 - ▶ Difficult to specify lighting and reflection conditions
- 4 Global Illumination (GI) techniques like Radiosity, Ray Tracing are computationally intensive

■ So... **Image Based Rendering !**



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Motivation(contd.)

- **Photorealism:** Images capture all effects
- Image Synthesis - *independent* of scene complexity
- *Collection* of samples easy and cheap

Limitations

Inherent rigidity of images being static !

- *Dynamics* in CG has been of keen interest to scientists
- *Lighting* changes is *one* way
 - ▶ Relit real/artificial scenes from novel illumination captured in natural/synthetic environments
- Traditional CG techniques apply computationally-intensive GI techniques for recomputing the scene
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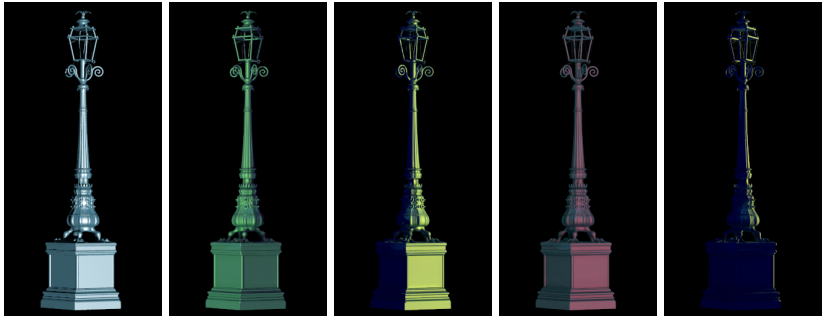
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Advantages ?

- Controlling **illumination** improves **recognition** and **satisfaction**
- Saves artist/ animator's enormous **time** and **effort** to achieve realistic environments and animations
- Applications range from **movies** and **interactive computer games** to **augmented reality**
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Problem Statement

Image-based Relighting:

Given images of a scene captured under a certain set of illumination conditions, how to render the scene under a novel illumination configuration ?



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Introduction

- Pre-rendering (synthetic scenes) or pre-acquisition (real scenes) of a collection of images in which the lighting direction is systematically varied.
- Due to linearity of scene radiance, images of the scene under novel illumination can be computed by superposition of synthesized/captured images.
- **Issue:** Collection of images is typically too large both to store in memory and to synthesize novel images fast.

Contribution

A two-stage image-based algorithm for fast relighting.

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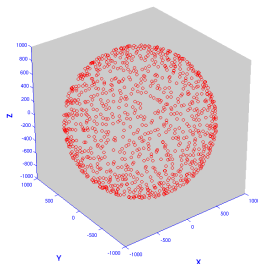
Contribution

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Input/Output

Input

A set of reference images with the same viewpoint, but under different lighting directions sampled uniformly on a sphere.



Output

Given a new lighting configuration, compute a novel relit image.



Related Work

■ Two-Stage Singular Value Decomposition [NBB04]

- ▶ Uses first level SVD to factorize the original image data into two factors. Then apply SVD again to factorize one of the factors (compute in previous SVD) into another two factors. These three factors are used for relighting.
- ▶ **Advantages:** Harness both inter and intra pixel correlations. Optimal solution, so good results.
- ▶ **Issues:** Store too much data (two sets of basis functions and a set of coefficients).

■ Illumination Adjustable Images (IAI) [WL03]

- ▶ Uses Spherical Harmonics (SH) to model each pixel along the lighting domain. Given a new lighting direction, uses the computed SH coefficients for relighting.
- ▶ **Advantages:** Only one set of SH coefficients needed to be stored. Basis functions are simple numerical functions.
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Our Approach

- 1** First, exploiting the correlation among pixels of an image, compute a set of image bases and their corresponding relighting coefficients using SVD.
- 2** Second, exploiting the coherence among the computed relighting coefficient, compute a reduced set of SH relighting coefficients.

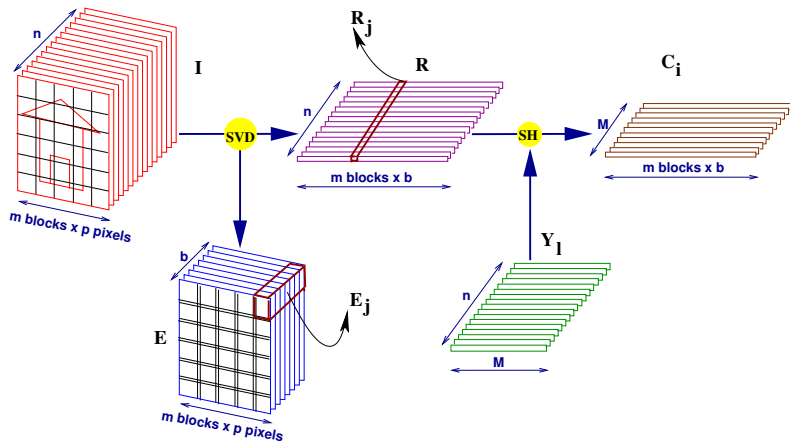


Stage 1

- Using **SVD**, factorize the set of images (**I**) i.e.,
$$\mathbf{I} = \mathbf{U} * \mathbf{W} * \mathbf{V}.$$
- Compute a rank b approximation.
- Split the singular values into two halves (taking square roots) and multiply them to \mathbf{U} and \mathbf{V} . So,
- $\mathbf{I} \approx \mathbf{R} * \mathbf{E}.$
- **R** and **E** are termed the **SVD relighting coefficients** and **eigen image bases** respectively.



Modeling



Stage 2

- Model the SVD relighting coefficients (\mathbf{R}) using **SH**.

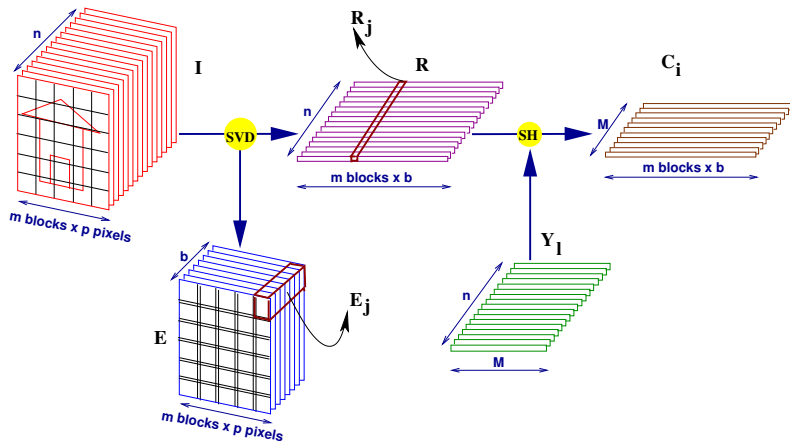
$$\mathbf{C}_{l,m} = \int_0^{2\pi} \int_0^\pi \mathbf{R} * Y_{l,m}(\theta, \phi) * \sin \theta * d\theta * d\phi$$

where $\mathbf{C}_{l,m}$ are the **SH relighting coefficients**, n is the number of lighting directions in the illumination space (θ, ϕ) and $Y_{l,m}$ is the SH function.

- Compute a degree l approximation.



Modeling



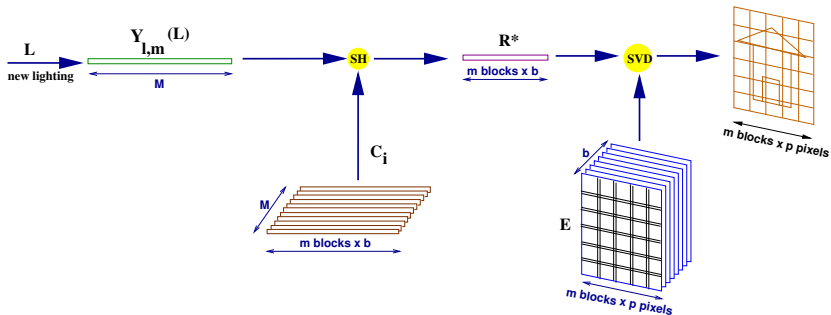
Relighting

Given a new light source $L(\theta', \phi')$, calculate

- Compute the SH functions $Y_{l,m}(\theta', \phi')$.
- Compute the product $C_{l,m}Y_{l,m}(\theta', \phi')$, the **reconstructed relighting coefficients** (\mathbf{R}^*).
- Compute the product of the **reconstructed relighting coefficients** and the **eigen image bases** ($\mathbf{R}^* * \mathbf{E}$), which is the novel relit image.



Relighting



Overview

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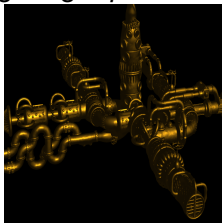
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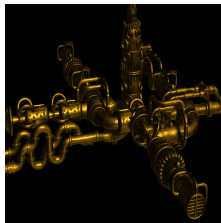


Results

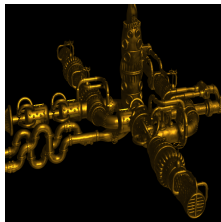
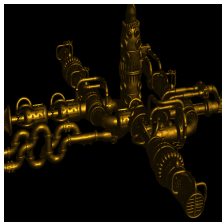
Relighting *PipeSet* under a novel illumination.



Original



Our Algorithm



Results

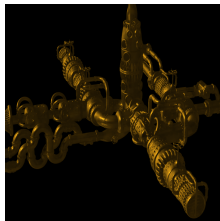
Relighting *PipeSet* under a different novel illumination.



Original

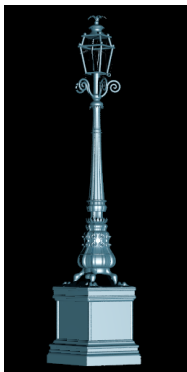


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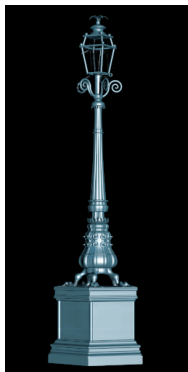


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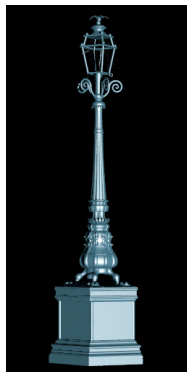
Relighting *LampPost* under novel illumination.



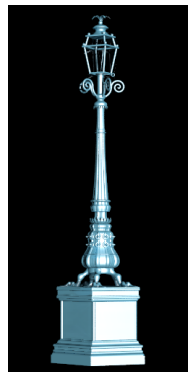
Original



Our Algorithm



Two-Stage SVD

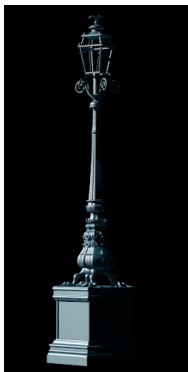


IAI

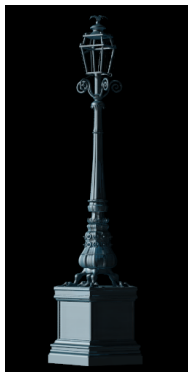


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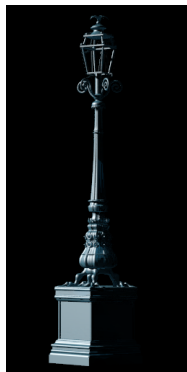
Relighting *LampPost* under a different novel illumination.



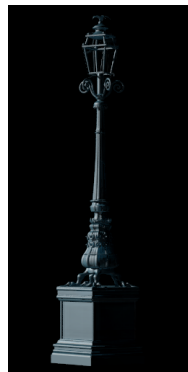
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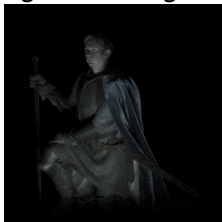


IAI



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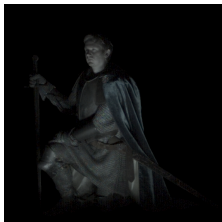
Relighting *Knight-kneeling* under a different novel illumination.



Original

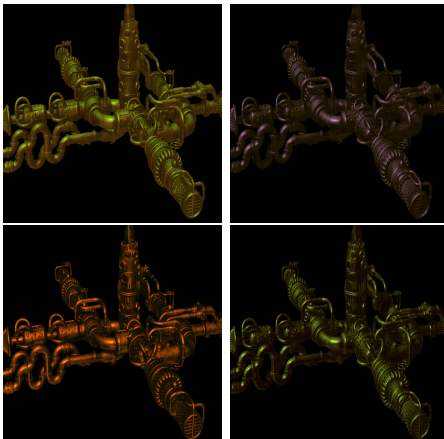


Our Algorithm



Results

Relighting using one or more novel light sources, with different intensity and color.



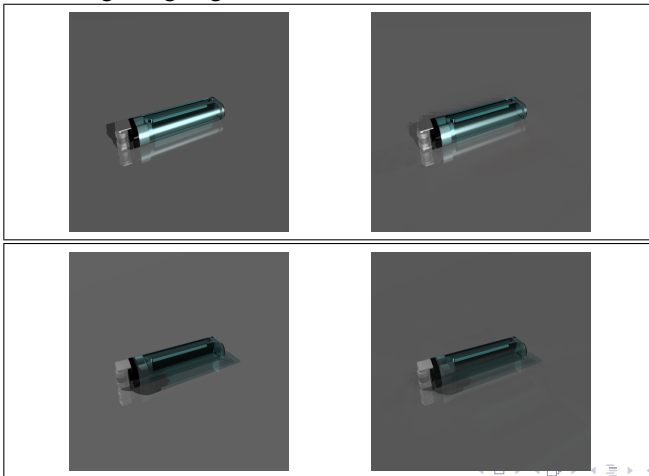
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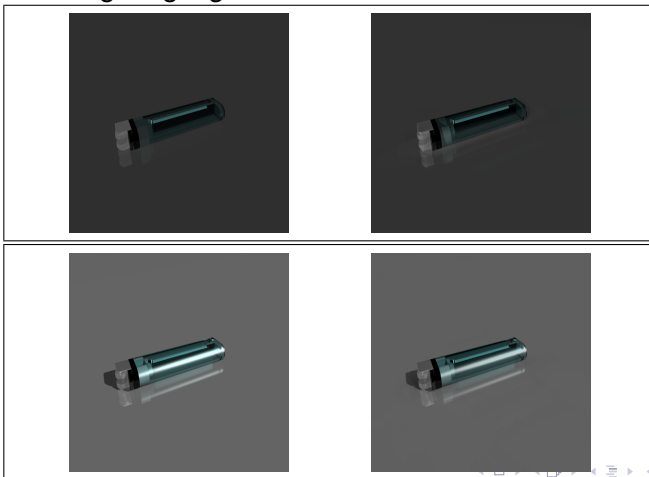
Results

Relighting *Lighter* under novel illuminations.



Results

Relighting *Lighter* under novel illuminations.



Results

	Our Algorithm			Two-Stage SVD			IAI		
	Size	Pre-P.	Relight	Size	Pre-P.	Relight	Size	Pre-P.	Relight
<i>LampPost 1</i>	130	4206.3	7.2	477	3736.3	29.0	513	34069	28.38
<i>LampPost 2</i>	419	5883.3	15.8	477	3871.4	28.8	513	55873	38.52
<i>LampPost 3</i>	144	2448.5	5.3	477	2195.4	19.4	513	56850	40.52
<i>PipeSet 1</i>	127	649.9	17.4	336	630.5	25.2	347	4566	42.31
<i>PipeSet 2</i>	142	745.5	18.5	336	645.4	25.86	661	11328	60.9
<i>PipeSet 3</i>	129	445.9	10.6	336	466.2	18.0	402	3546	31.74
<i>Knight 1</i>	617	1095.6	41.6	633	839.4	38.5	868	12077	100
<i>Knight 2</i>	588	946.6	26.5	633	604.2	28.0	790	10523	87.6
<i>Knight 3</i>	617	1044.5	32.7	633	618.5	27.7	639	7546	39.2



Overview

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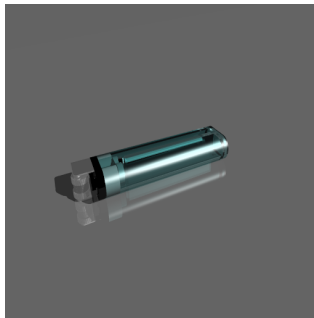
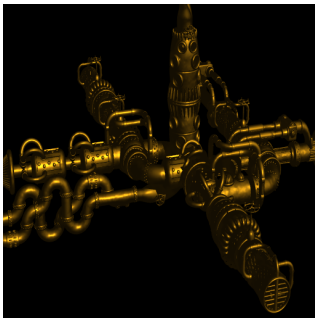
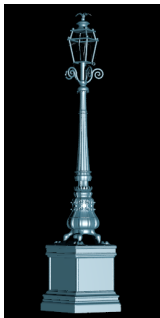
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Conclusion

- We propose a novel two-stage image based relighting algorithm
- We created three IBRL datasets (*PipeSet*, *LampPost* and *Lighter*) and have made them publicly available at <http://www.cse.iitb.ac.in/biswarup/web/data/>.



Thank you for your time !

Questions ?

<http://www.cse.iitb.ac.in/~biswarup/>



Debevec's Light Stage 6.0

