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Fast Image Based Lighting for Mobile Realistic AR

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Abstract

This paper presents a mobile implementation of realistic augmented reality using a simple image based lighting method. The front camera of the mobile device is used to interactively capture and update an environment map. Then, by making some reasonable assumptions on local geometry and object reflectance function, incident lighting is integrated in real-time. The method handles dynamic environment and soft shadows, and runs at real-time framerates on high-end devices.


Keywords: augmented reality, real-time rendering, image-based lighting, environment acquisition

1 Introduction

The term "realistic augmented reality" (realistic AR) is used to qualify an application where convincing interactions between real and virtual elements are computed. This aspect is important for user immersion: illumination computations add important visual cues to the scene (shadows, object shading, caustics...), notably enhancing global scene comprehension. The main scientific issue of realistic AR is common illumination, which is a common rendering pipeline for virtual and real (acquired) data. Achieving common illumination in a general context is hard: geometric and photometric information of the environment has to be acquired and transformed, potentially at interactive rates, to fit the rendering pipeline.
This usually needs specialized hardware (for acquisition) and complex algorithms (for transformation), and is rarely compatible with realtime constraints. Alas, mobile platforms like smartphones or tablets, which are an ideal application playground for realistic AR, have limited computational resources.

This paper presents an environment map acquisition method and an image-based rendering model that are suitable for mobile realistic AR. The front camera of a mobile device is used to interactively capture and update the environment map. Then, by making some reasonable assumptions on local geometry and virtual object material model, the rendering equation may be simplified enough to analytically integrate incident lighting in real-time by exploiting the filtering capacities of the graphics hardware. The method handles dynamic environment and soft shadows, and is simple enough to be implemented on mobile operating systems. It runs at real-time frame rates on high-end devices.

This document will be organized as follows: following this introduction is a short overview of related work on image-based lighting and realistic AR. The third section will introduce our contributions on environment capture and image-based lighting. The fourth section will present some results, along with limitations and potential extensions in the fifth section. A brief conclusion will end the paper in the sixth section.

2 Related work

Realistic augmented reality deals with coherent virtual object insertion into real environments with consistent lighting. As pointed by Jacobs and Loscos [Jacobs and Loscos 2006], major progresses were made, benefiting of improvements in acquisition systems or rendering methods. Early methods used manual light modeling (Fournier et al. [Fournier et al. 1993], Drettakis et al. [Drettakis et al. 1997], Loscos et al. [Loscos et al. 1999]). Although, manually modeling light sources is a tedious and time consuming work, which is not suited for mobile usage.

Environment maps are a convenient way to capture and represent incident lighting. Debevec [Debevec 1998] was the first to describe the use of environment map as a light source model. He acquires an environment omnidirectional image using HDR imaging and a gaze ball. Sato et al. [Sato et al. 1999] used a similar rendering technique but acquire the environment with a couple of fisheye lens cameras. Numerous similar works were proposed (Aguasanto et al. [Aguasanto et al. 2003], Supan et al. [Supan et al. 2006]) but all of them need complex calculations and/or specialized hardware, which is not suited for mobile implementation.

Potential solutions exists for handling complex lighting within an mobile AR context. Snyder [Snyder 2005] propose explicit solutions for shading an object with extended light sources, providing that the BRDF follows a specific model (Lambert or Phong power-law model). This work is interesting but only applies on uniform light sources. Lighting with an environment map has been tackled by McGuire et al. [McGuire et al. 2013] by exploiting the fact that modern graphic processors allows the filtering of cube-maps. Similarly to Snyder’s work, this work assumes that the BRDF is a specific one (normalized Blinn-Phong). Finally, Calian et al. [Calian et al. 2013] proposed an alternative by capturing shading instead of lighting with a specific probe with a custom geometry designed to fit different lighting computations. Although computation cost is low, the material of the virtual object has to be identical to the probe’s, therefore limiting possible appearances. Furthermore, a specific target has to be built for every lighting configuration.

None of these techniques suit our needs for mobile AR: they are too restrictive or too computationally expensive, and all of them use intrusive probes or gaze balls. In this paper, we propose a method which dynamically capture an environment map with cell-phone cameras and lights a virtual object with it in real-time. We extended McGuire’s to take into account every face of the cube-map in the incident luminance integration. We demonstrate that extensions to anisotropic models are also possible. This method has been implemented on a high-end tablet using OpenGL ES 3.0 and runs at real-time frame rates.

3 Contribution

This section will introduce our method for real-time mobile AR. We track the device with specifically designed frame markers using QR Codes. The surrounding lighting environment is dynamically captured and updated using the device’s front camera. The virtual object is then directly lit with this environment map using an extended version of McGuire et al.’s work [McGuire et al. 2013]. Finally, projected shadows are approximated with sphere proxies. We will now go into details of these points.

3.1 Environment capture

Incident lighting is modeled using environment maps. Of course, one could always create these maps using offline methods: a large number of applications facilitate the creation of panoramas on mobile devices. For example, Cycloramic [cit] uses the device’s vibrator to automatically rotate on a slick surface while aligning successive captures. One can easily add a wide angle lens to obtain environment maps.

Although, offline capture assumes that the lighting environment is static, which is rarely the case in interactive applications. We propose to use the front camera to update the environment map. Since the camera is calibrated, and assuming the lighting is far enough, we may infer incident lighting direction for each pixel of a frame captured by the front camera and reproject it on the environment map, as shown on figure ??. Note that with this technique, the user is potentially seen by the front camera. We do not consider this as an issue, since the user is part of the environment and his influence should be visible on the virtual object lighting.

3.2 Image-Based Lighting

The ultimate goal of every rendering algorithm is the resolution of the rendering equation [Kajiya 1986]. For any point p of our virtual object, the objective is to compute the output luminance from p in direction ω0:

\[ L(p, ω_0) = L_e(p, ω_0) + \int_{Ω_+} f(p, ω_0, ω_i)L_i(p, ω_i)\cosθ_i dω_i \]  

with \( L_e(p, ω_0) \) being the emitted luminance, \( f(p, ω_0, ω_i) \) being the reflectance function, \( L_i(p, ω_i) \) being the incident luminance and \( ω_i \) being the incident direction.

In our application, incident luminance \( L_i \) is represented with an environment map. For implementation purposes, this environment map is mapped on a cube (cube-map) around the virtual object.

If the reflectance function is simple enough, one may approximate the convolution of the incident luminance by filtering the environment map. The hardware filtering capacities of the graphics hardware may be used for this purpose. MacGuire et al. [MacGuire et al. 2013] demonstrated that cosine power law filtering may be approximated by the mipmap levels of the cube-map. For a cosine power
law with exponent $s$, the corresponding mipmap level $m$ is given by (see paper for details): 

$$m = \log(w \sqrt{3}) - 0.5 \times \log(s + 1)$$

Therefore, any reflectance function that has cosine power law components may be used with this method. MacGuire et al. propose to use the normalized Blinn-Phong reflectance model:

$$f(\omega_i, \omega_i) = \frac{1}{\pi} (k_L + \kappa_G) \frac{s + 8}{8} \max\left(\frac{\omega_i \cdot \omega_i}{\|\omega_i \| \cdot \|\omega_i\|}, 0\right)^s$$

(3)

When replacing and integrating this reflectance function in the rendering equation, the diffuse $L_d(p, \omega_0)$ and glossy $L_g(p, \omega_0)$ components may be separated:

$$L(p, \omega_0) = L_e(p, \omega_0) + \frac{1}{\pi} (k_L L_d(p, \omega_0) + \kappa_G L_g(p, \omega_0))$$

(4)

with

$$L_d(p, \omega_0) = \int_{\Omega^+} L_e(p, \omega_0) \cos \theta_i d\omega_i$$

(5)

MacGuire et al. assume that the diffuse component $L_d$ is proportional to the lowest level of the mipmap, and the glossy component is proportional to the sample in the mipmap level computed by equation $(2)$. Therefore, computing luminance at point $p$ is just the combination of two texture fetches.

However, this method may lead to visible artefacts: for the diffuse component, a single face of the cube-map is chosen and is the only one to contribute to shading. If two adjacent faces of the cube-map are very different at the lowest level, a discontinuity will appear in the diffuse component of the object. Illustrations of these artefacts may be found in the supplemental material. We propose an improvement of this method to take into account every visible faces from point $p$.

To evaluate this, let us calculate the solid angle described by a partially visible face $F$ along all directions $\omega$ on the subtending hemisphere with orientation $\vec{n}$:

$$\Omega_F(\vec{n}) = \int_{F} \frac{Hs(\vec{\omega}, \vec{n})}{\|\omega\|} d\omega$$

(6)

with $Hs$ being the Heavyside function. The contribution $W_F(\vec{n})$ of each face may then be calculated by integrating the normalized dot product $\vec{\omega} \cdot \vec{n}$ with the solid angle on the whole face:

$$W_F(\vec{n}) = \int_{F} \frac{\vec{\omega} \cdot \vec{n}}{\|\omega\|} \times \frac{Hs(\vec{\omega}, \vec{n})}{\|\omega\|^3} d\omega$$

$$= \int_{F} \frac{\vec{\omega} \cdot \vec{n} \times Hs(\vec{\omega}, \vec{n})}{\|\omega\|^4} d\omega$$

(7)

Therefore, assuming incident diffuse light from face $\vec{F} \cdot \mathcal{L}_d(\vec{F})$ and pre-computed ambient occlusion factor $\mathcal{P}_v(p)$ are known, the diffuse luminance $\mathcal{L}_d(p, \vec{n})$ of point $P$ with a normal $\vec{n}$ is given by:

$$\mathcal{L}_d(p, \vec{n}) = \frac{\mathcal{P}_v(p)}{\pi} \sum_{F \in \text{Faces}} \mathcal{L}_d(\vec{F}) W_F(\vec{n})$$

(8)

Equation $(8)$ is computed on each vertex and has to be rapidly evaluated. Since $W_F(\vec{n})$ depends roughly only on $\theta$ (and $\phi$ being euler angles), a numerical approximation may be obtained by a function parametrized by $\vec{n}, \vec{F} = \cos(\theta)$:

$$W_F(\cos(\theta)) = \int_{F} \cos(\theta) \times \frac{Hs(\cos(\theta))}{\|\omega\|^3} d\omega$$

$$\sim \left[ \max\left(0.75 + \cos(\theta), 0\right) \right]^2 \frac{1}{1.75}$$

(9)

Despite its simplicity, equation $(9)$ is very close to $W_F(\vec{n})$ and may be used in equation $(8)$ with reasonable precision.

### 3.3 Shadowing

![Sphere decomposition of 3D objects and obstruction computation](image)

There is numerous evidences that shadows enhance scene comprehension, especially on spatial relationships between objects (Madsen et al.[Madsen et al. 2003], Sugano et al.[Sugano et al. 2003]). Although, shadowing a complex geometry with complex lighting is not an easy task. In this work, we will assume that the local geometry around the virtual object may be approximated by a horizontal plane. The virtual object lies onto this plane, and shadows will be projected on it.
To compute soft shadows, one has to evaluate environment visibility for every point of the horizontal plane. Again, computing this by sampling the environment is too costly. Furthermore, since we do not use point light sources, classical shadowing techniques like shadow maps are not usable here. We chose instead to replace the actual geometry with sphere proxies. Then, the obstructed part of the environment map may be calculated and incident lighting may be dimmed accordingly (see figure 3). Additional details are given in the supplemental material. Figure 2 shows some results of our image-based rendering method with shadows.

4 Results

We will now demonstrate the capacity of our system to cope with dynamic lighting environments. Figure 4 shows a time lapse sequence of this ability. On the first frame, the virtual object is fully lit by the environment. On the second frame, the operator places his finger on the front camera of the device. One may see that the lighting modification is reported on the shading of the object. This application also runs on an iPad Air at approximately 25 frames per second, the frame resolution is 640 × 480.

5 Discussion and Possible Extensions

As said in section 3.2, the BRDF of the virtual object has to follow the Blinn-Phong model. This is the price to pay for real-time shading with complex light sources. A extension to micro-facet model BRDF seems doable by evaluating the anisotropic integration of the environment map, using hardware texture gradients for example. Although, at present time, texture gradients only apply on a single face of the cube-map and does not handle face continuity. Mobile implementation also presents some limitations due to hardware restrictions. First, shadow evaluation with sphere proxies is a little too costly for mobile devices: the decomposition of the geometry has to be limited to a reasonable number (around 50) of spheres to maintain an interactive framerate. Second, at present time, it is impossible to stream video from both cameras (front and back) at the same time on iOS, probably for efficiency and bandwidth reasons. To cope with this, we chose to mainly use the back-facing camera and to rapidly switch every five seconds on the front-facing camera to take a picture. Finally, in current implementation, captured images are low dynamic range (LDR). Strong light sources may not be identified in the environment map and no hard shadows will be rendered. High dynamic range (HDR) acquisition is a tough problem in real-time applications: the traditional method involves successive captures, possibly with quite long exposure times and naturally limiting the update rate of the environment map. Some works have been done on HDR acquisition with a single frame (Rempelet et al. [Rempelet et al. 2007]) but most of them are too slow or too approximative to fit mobile AR requirements. One potential solution for this problem may be to incorporate knowledge of the environment in the HDR reconstruction: by segmenting the LDR image in high luminance zones, one may identify potential light sources by knowing the context.

6 Conclusion

A new solution for common illumination between real and virtual objects was presented in this paper. Assuming distant lighting, Blinn-Phong BRDF and planar local geometry, it handles both dynamic environments and soft shadows while being simple enough to run on a modern mobile device. It is achieved at real-time frame rates, leaving some computational resources for other purposes. Finally, as shown in previous section, numerous extensions are possible.

References


Figure 4: Results with dynamically updating envmaps.