Polynomial Wavelet Trees for Bidirectional Texture Functions

Jerome Baril*,1,3  Tamy Boubekeur†,2  Patrick Gioia1,3  Christophe Schlick§,1
LaBRI-INRIA, Bordeaux, France1  TU Berlin, Germany2  Orange Labs Rennes, France3

1 Introduction

Bidirectional Texture Functions (BTF) currently offer the highest quality representation for the appearance of complex real-world materials (see Figure 1). However, BTF acquisition leads to a huge amount of data and thus cannot be used as is. In a pixelwise approximation, working with light varying material for each eye direction is more convenient since BTF include strong parallax effects arising from high depth varying materials. The light transition is smooth and can be fitted with approximation functions but the multiplicity of physical effects captured during acquisition leads to errors in approximated data [Malzbender et al. 2001; Meseth et al. 2004; Filip and Haindl 2004]. Usually, a finer fitting is found with a multi-scale approximation since data are decomposed into varying importance sets to drive locally the complexity of approximation functions. In this paper, we present a new kind of factorization for compressing BTF, called the Polynomial Wavelet Tree (PWT). The key idea is to separate directional and spatial variations by projecting the spatial BTF domain onto a wavelet domain and to approximate the resulting light-dependent wavelet coefficients with varying degree polynomial functions that spans the lighting directions. Our method improves upon previous polynomial approximations and is designed for high quality materials rendering using GPU.

2 Polynomial Wavelet Tree

The BTF is a six dimensional function of \((x, y) \in \mathbb{R}^2\) the spatial coordinates, \((\theta, \phi) \in \mathbb{S}^1\) the viewpoint directions and \((\theta, \phi) \in \mathbb{S}^2\) the light directions. Our method consists in decomposing the BTF into an adaptive structure generated with three steps (see Figure 1). The first one is a projection of the BTF onto a wavelet basis. Then, wavelet coefficients are factorized using a polynomial approximation. Finally, the PWT is rendered with an efficient pre-computed GPU algorithm.

Weaves for BTF

Multi-resolution wavelet analysis decompose a function into a low resolution part and a set of coefficients that capture details from higher resolutions. We found enough to apply the wavelet expansion in spatial dimensions only to use the good properties of a wavelet decomposition: energy compaction and decorrelation. We use Wavelet Packets to decompose images into small subbands and then, to compute statistical informations about the contribution of a subband in the original data. Note that while a geometric wavelet decomposition can be applied in direction space subject to data resampling, this can lead to inaccurate approximation.

Adaptive Data fitting

Fitting wavelet coefficients instead of data themselves is more efficient because low frequency transitions are smooth and higher frequency coefficients can be quantized with less importance. We choose to use varying degree bivariate polynomials along the light directions, \((\theta_1, \phi_1)\), to approximate the data and, in this way, make the light direction space continuous. First, light space is projected onto a better suited space for data fitting \((\theta, \phi) \mapsto \Pi(\theta, \phi)\), where \(\Pi(\theta, \phi) \in \mathbb{R}^2\) is a paraboloid parameterization of the light hemisphere [Heidrich and Seidel 1998]. The data fitting process is computed on each node of the decomposition tree (i.e. per pixel) for a fixed point of view along the parameterized light directions using Singular Value Decomposition. The polynomial degree is chosen per subband according to its level of importance found during statistical analysis. The polynomial coefficients in the wavelet decomposition tree describe the PWT. A simplification of the PWT is done in a post-process by excluding minor contribution branches of the tree (i.e. smallest importance subbands) according to a quality criterion, a fixed amount of energy, or a size criterion.

Rendering

The inverse Wavelet Transform is partially pre-computed on polynomial coefficients and resulting textures are stored in a mipmap texture. Note that, in this way, our GPU data structure is independent of the wavelet basis and thus makes it possible to use high degree wavelet basis. Moreover, hardware texture filtering can be used for minification (embedded in our representation) and magnification. The synthesis of a pixel is done in one pass: \(n\) texture access (\(n < 3\), degree of bivariate polynomial) per pixel and few arithmetic operations for fixed view and light directions. Each viewpoint is stored using this method. Interpolation between viewpoints is done using a paraboloid map to store barycentric weights of intermediate view directions.

3 Results

The quality of our approximation outperforms existing solutions (see Figure 1 (b)) and, thanks to adaptive PWT simplification, the size of the data is less than fifty percent that of PTM. The synthesis remains efficient with a partial reconstruction of the first levels of the PWT and avoids aliasing with a better low resolution approximation than classical mipmap generation. Last, PWT is not limited to BTF and can be applied on any spatially varying material representation. Note that polynomial approximation is not robust on specular materials. However, our method is independent of any approximation scheme and, therefore, better approximation functions could be used.

References