# High-quality Image-based Interactive Exploration of Real-World Environments<sup>1</sup>

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**Figure 1:** The system described in this paper enables users to virtually explore large environments from the comforts of their sofa (a). A standard wireless gamepad (b) allows the user to easily navigate through the modeled environment which is displayed on a large wall screen (c). This results in a compelling experience that marries the realism of high quality imagery acquired from real world locations together with the 3D interactivity typical of computer games.

### **1** Introduction

Interactive scene walkthroughs have long been an important application area of computer graphics. Starting with the pioneering work of Fred Brooks [1986], efficient rendering algorithms have been developed for visualizing large architectural databases [Teller and Séquin, 1991, Aliaga *et al.*, 1999]. More recently, researchers have developed techniques to construct photorealistic 3D architectural models from real world images [Debevec *et al.*, 1996, Coorg and Teller, 1999, Koch *et al.*, 2000]. A number of real-world tours based on panoramic images have also been developed (see the Previous Work section). What all of these systems have in common is a desire to create a real sense of *being there*, i.e., a sense of virtual presence that lets users experience a space or environment in an exploratory, interactive manner.

In this paper, we present an image-based rendering system that brings us a step closer to a compelling sense of being there. While many previous systems have been based on still photography and/or 3D scene modeling, we avoid explicit 3D reconstruction in our work because this process tends to be brittle. Instead, our approach is based directly on filming (with a set of high-resolution cameras) an environment we wish to explore, and then using image-based rendering techniques to replay the tour in an interactive manner. We call such experiences *Interactive Visual Tours* (Figure 1).

In designing this system, our goal is to allow the user to move freely along a set of predefined tracks (*cf.* the "river analogy" in [Galyean, 1995]), choose between different directions of motion at decision points, and look around in any direction. Another requirement is that the displayed views should have a high resolution and dynamic range. The system is also designed to easily incorporate



Figure 2: Simplified block diagram of our video processing and rendering pipelines

multimedia objects such as navigation maps, video textures, audio, and pop-up stills.

Our system is not the first one to propose interactive video-based tours (see the Previous Work section). We believe, however, that our system is the first to deliver fully interactive photorealistic image-based tours on a personal computer at or above broadcast video resolutions and frame rates. Previous image-based tours have generally been of much lower visual quality since they are either based on a single camera taking spherical images or high-resolution panoramas taken at discrete locations (Section 3 has a more detailed comparison). None of the tours, to our knowledge, provides the same rich set of interactions, e.g., the ability to navigate continuously from room to room with a natural, game-like controller, or to jump from room to room using map navigation. Compared to approaches based on 3D reconstruction [Debevec *et al.*, 1996, Coorg and Teller, 1999, Koch *et al.*, 2000], our system can deal with environments of much greater visual complexity, e.g., it is capable of navigating the rich foliage of a botanical garden without any loss in realism. We postpone a full comparison of our system with previous work to the Discussion section, once we have described our own system in more detail and shown some of our results.

In order to achieve our goals of providing high-resolution interactive image-based tours, we designed an integrated acquisition, authoring, and playback system from the ground up (Figure 2). The novel components we developed as part of this system include:

- a multi-camera high-resolution video head connected to a small portable RAID array and power supply (Section 3);
- an acquisition platform to conveniently transport the sensor and capture the space we wish to explore (Section 4);
- a suite of image processing algorithms to de-noise, de-mosaic, un-warp, and de-vignette the images (Section 5);
- a spatial stitching/de-ghosting algorithm designed to minimize visible parallax in overlap regions (Section 6);
- a high dynamic range video stitching algorithm that merges successive frames taken at different exposures while the camera is moving (Section 7);

- a feature tracking and pose estimation algorithm designed to stabilize the acquired video and aid navigation (Section 8);
- a video compression scheme designed for selective (partial) decompression and random access (Section 9);
- authoring tools for inserting branch points in the tour, adding sound sources, and augmenting the video with synthetic and hybrid elements; and (Section 10).
- an interactive viewer that allows users to easily navigate and explore the space that has been captured (Section 11).

Each of these components is described below in its own section. The overall complexity of the system, however, is hidden from the end-user, who gets to intuitively navigate compelling interactive tours of remote locations using a simple game control pad (Figure 1).

# **2 Previous work (panoramic photography)**

Panoramic photography using mechano-optical means has a long history stretching back to the 19th century [Malde, 1983, Benosman and Kang, 2001]. The *stitching* of photographs using computer software has been around for a few decades [Milgram, 1975], but first gained widespread commercial use with the release of Apple's QuickTime VR system in 1995 [Chen, 1995]. Since then, a large number of improvements have been made to the basic system, including constructing full-view panoramas [Szeliski and Shum, 1997, Sawhney and Kumar, 1999], dealing with moving objects [Davis, 1998], and using panoramas for image-based rendering [McMillan and Bishop, 1995] and architectural reconstruction [Coorg and Teller, 1999]. (See [Benosman and Kang, 2001] for a collection of some more recent papers in this area.) Most mid- to high-end digital cameras now ship with image stitching software in the box, and there are over a dozen commercial stitching products on the market today (*http://www.panoguide.com/software* lists a number of these).

Panoramic (360°) movies were first introduced by Kodak at the World's Fair in 1964 and are a popular attraction in theme parks such as Disneyland. Interactive video tours were first demonstrated by Andrew Lippman in his Movie Maps project, which stored video clips of the streets of Aspen on an optical videodisc and allowed viewers to interactively navigate through these clips. In mobile robotics, 360° video cameras based on reflecting curved mirrors (*catadioptric* systems [Baker and Nayar, 1999]) have long been used for robot navigation [Ishiguro *et al.*, 1992]. More recently, such systems have also been used to present virtual tours [Boult, 1998, Aliaga and Carlbom, 2001,

Aliaga *et al.*, 2002] and to build 3D image-based environment models [Coorg and Teller, 1999, Taylor, 2002].

For example, Boult [Boult, 1998] developed a campus tour based on re-projecting catadioptric spherical images streaming from a camcorder to give the user the ability to look around while driving through campus. Teller *et al.* [Coorg and Teller, 1999] have built detailed 3D architectural models of the MIT campus based on high-resolution still panoramic images taken from a specially instrumented cart with a pan-tilt head. Aliaga *et al.* [Aliaga and Carlbom, 2001, Aliaga *et al.*, 2002] have built image-based interactive walkthrough of indoor rooms based on a dense sampling of omnidirectional images taken from a mobile robot platform. Taylor's VideoPlus system [Taylor, 2002] uses a similar setup but relies more on a sparse traversal of a set of connected rooms to explore a larger region of interest. His paper also contains a nice review of previous work in the area of environment modeling and visualization using catadioptric images.

Daniilidis' Omnidirectional Vision web page (//www.cis.upenn.edu/~kostas/omni.html) lists almost 50 different research projects and commercial systems based on omnidirectional cameras. Of the systems listed on his page, the majority employ curved mirrors and a single camera, while a few employ multiple cameras. One clever design (*http://www.fullview.com*) uses mirrors to make the optical centers coincident, which removes the parallax problems described in this paper (but results in a slightly bulkier system). The system built by iMove (*http://www.imoveinc.com*) is the most similar to ours, in that it uses a collection of outwardly pointing video cameras, but has a lower spatial resolution than the system built for us by Point Grey Research (*http://www.ptgrey.com/products/ladybug/*).

# 3 Omnidirectional camera design

The first step in achieving our goal of rendering full-screen, interactive walkthroughs is to design a capture system. The system needs to efficiently capture full spherical environments at high resolution. One possible approach is to capture many high-resolution stills and to then stitch these together. This step then has to be repeated at many locations throughout the environment [Coorg and Teller, 1999]. While this yields very high resolution panoramas, it can require a long amount of time, e.g., days instead of the few minutes that it takes us to digitize a home or garden. We decided instead on omnidirectional capture at video rates.

A common approach to omnidirectional capture is to use catadioptric systems consisting of mirrors and lenses. There are two major mirror based designs: those using curved mirrors with a single lens and sensor, and those using multiple planar mirrors each with an associated camera. These systems have the desirable feature that they capture a large cross section of a spherical



**Figure 3:** Our six-sensor omnidirectional camera (a). The six images taken by our camera are arranged horizontally around a pentagon, with a sixth sensor looking straight up. During processing, the overlapping fields of view are merged and then re-sampled. During rendering, these images become the texture maps (b) used in the 3D environment map.

environment and require no complex stitching, since they are designed to have a single optical center. In order to enable full-screen experiences, our system needs to capture enough pixels to provide at least a  $640 \times 480$  resolution in any given  $60^{\circ}$  field of view. A curved mirror design would need at least a  $2K \times 2K$  video rate sensor in order to approach these requirements, but the desired resolution would only be within a fraction of its field of view. This is because these designs have a non-uniform distribution of image resolution, which drops significantly towards the center of the image. In addition, these designs have blind spots at the center of the image, where the field of view is blocked by the camera itself.

Another class of omnidirectional sensors uses multiple individual cameras arranged in such a way that they fully capture a scene. A disadvantage of this approach is that it is difficult to arrange the cameras so that they share a common center of projection. The captured data must then be post-processed with a more complex stitching and parallax compensation step to produce a seamless panoramas (see our section on Image Stitching). However, parallax can be minimized by packing the cameras very tightly together. On the plus side, such a system can capture most of a full viewing sphere, can be made high-resolution using multiple sensors, and can be built in a rugged and compact design. For these reasons, this was ultimately the design that we selected.

Since there was no commercially available system that met all of our requirements, we designed our own multi-camera system, and contracted Point Grey Research to build it. The result is the Ladybug camera (*http://www.ptgrey.com/products/ladybug/*) shown in Figure 3. The acquisition system is divided into two parts: a head unit containing the cameras and control logic, and a storage unit tethered to the head via a fiber optic link. The head unit contains six  $768 \times 1024$  video-rate

sensors and associated lenses, with five pointing out along the equator, and one pointing up, giving complete coverage from the North Pole to 50° below the equator. The lenses were chosen so that there is a small overlap between adjacent cameras, allowing for stitching of the captured data. The sensors run at a video rate of 15 frames/second and are synchronized. The storage unit is a portable RAID array consisting of four 40GB hard drives, which allows us to store up to 20 minutes of uncompressed raw video. All components are powered by a small portable battery, and the complete system can easily fit in a backpack or on a small robot (Figure 4), giving us a lot of flexibility in terms of acquisition.

# 4 Data acquisition

Before we can capture large real-world scenes with our novel camera, we need to design a capture platform to carry the sensor around, as well as a capture strategy to ensure that all the desirable data gets collected. In this section we describe these two components.

#### 4.1 Capture platform

To capture omnidirectional videos of scenes, we designed three different mobile platforms. Because of the extremely wide field of view of our omnidirectional camera, each of our mobile platforms was designed to avoid imaging itself and the camera person. (We considered using a steady-cam rig, since it is capable of producing very smooth camera trajectories. Unfortunately, the person carrying the steady-cam rig would always be visible to the omnidirectional camera, and so we rejected this option.)

**Tripod platform.** Our first design was a tripod mounted on a dolly (Figure 4a). This platform is moved along the planned path by a human crouched next to it. The platform is stable (assuming flat ground), compact, cheap, and easy to assemble. Unfortunately, in order to avoid being seen, the operator has to crouch in an uncomfortable position, making it awkward to move. It is also incapable of handling relatively rough terrain.

**Radio-controlled robot cart.** Our second design was a radio-controlled robot platform (Figure 4b). A number of customized changes were made, including adding a monopod stabilized with tension wires, extending one of the wheels to increase the footprint, and adding another layer of support for holding the controlling laptop. While this solution is significantly more expensive than the tripod, it removes the operator from the camera view and it is capable of repeatable paths



**Figure 4:** Acquisition platforms: (a) Wheeled tripod. (b) Tele-operated robotic cart. (c) Helmet-mounted acquisition system.

with little jitter. It also has the potential for autonomous exploration and acquisition [Taylor, 2002, Aliaga and Carlbom, 2001]. However, it cannot handle stairs and rough terrain.

**Head-mounted helmet.** In order to handle arbitrary terrain, we mounted the camera on a skydiving helmet (Figure 4c). This platform can be used practically anywhere (we have even used it while sky-diving), and the pace taken along the path during acquisition can be reasonably rapid. Unfortunately, camera jitter becomes a significant issue.

For the results presented in this paper, we used the tripod platform to capture our first tour and the head-mounted helmet for the later two.

#### 4.2 Trajectory planning

As with any regular video production, we first develop a "script". In our case, this consists of deciding which portions of the scene will be filmed and devising a path (traversal) that will take the camera through those areas. When available, we use maps or blueprints both to plan the video acquisition and to help the user navigate the final tour. During the filming, the camera-person walks along the pre-planned trajectory, and someone else records sounds for incorporation into the authored experience.



**Figure 5:** *Radial distortion and vignetting correction: (a) image of calibration grid, (b) original photograph of a building showing lens distortion, (c) corrected image, with straight lines converging on a vanishing point; (d) original image showing vignetting effect, (e) image after vignetting correction.* 

# 5 Image pre-processing

The first step of our off-line processing pipeline is to perform some image processing tasks to convert the raw video values coming from our sensor chip to color images and to correct for radial distortion and vignetting.

#### 5.1 De-mosaicing

Images are copied directly from the sensors onto the RAID array as raw Bayer color filter array values (i.e., individual red, green, or blue samples arranged in a checkerboard array consisting of 50% G pixels and 25% R and B pixels) [Chang *et al.*, 1999]. We use the algorithm described in [Chang *et al.*, 1999] to interpolate to full RGB images while preserving edge detail and minimizing spurious color artifacts. We also undo the camera gamma so that the rest of the processing can be done in a linear color space.

#### 5.2 Radial distortion

Since the field of view of our cameras is large ( $\sim 100^{\circ}$ ), it introduces significant lens distortion. To generate seamless mosaic images, the accuracy of the lens calibration is of paramount importance. In fact, even a small degree of inaccuracy in the correction of the images will affect the quality of the stitching in the regions where two images overlap.

We found that for our lenses, a parametric radial distortion model does not achieve sufficient accuracy. We therefore developed a slightly more elaborate two-step correction algorithm. We start

from the standard plumb-line method where a parametric radial model is fitted to the image data with the purpose of straightening up edges that appear to be curved [Brown, 1971, Devernay and Faugeras, 1995]. Then, a second step is added to remove the small residual error by first finding all corners of the image of a grid and mapping them back to their original locations. The output displacements are then interpolated with a bilinear spline. The final result is a pixel-accurate lens correction model (Figure 5). This technique is able to produce more accurate results than just fitting a standard parametric radial distortion model.

#### 5.3 Vignetting

Cameras with wide fields of view also suffer from a vignetting effect, i.e., an intensity drop-off away from the center. This drop-off is due to both optical (off-axis illumination) and geometric (aperture) factors. We compensate for these effects by first capturing an image within an integrating sphere whose internal surface is almost perfectly Lambertian. We then fit a global function that accounts for this effect. The function we use is of the form  $I(r) = I_0(r)[1 - \alpha r]/[1 + (r/f)^2]^2$ , where r is the radial distance from the principal point (assumed to be the center of the image), I(r) is the sampled image,  $I_0(r)$  is the hypothesized un-attenuated intensity image, f is the camera focal length, and  $\alpha$ is the geometric attenuation factor. Details of extracting these parameters can be found in [Kang and Weiss, 2000].

#### 5.4 Geometric calibration

Once each camera has been unwarped to fit a linear perspective model, we need to compute the linear transforms that relate camera pixels to world rays relative to the camera head. One way to do this would be to track a large number of points while rotating the head and to then perform a full bundle adjustment (structure from motion reconstruction [Hartley and Zisserman, 2000, Triggs *et al.*, 1999]. However, we have found that a simpler approach works well enough for our purposes. We simply take an image of an outdoor scene where there is negligible parallax and use an image stitching algorithm to compute the camera intrinsic parameters and rotations [Szeliski and Shum, 1997]. We then estimate the inter-camera translations from the design specifications. While these are not the exact displacements between the optical centers, they are close enough that we can compensate for inter-camera parallax using the technique described in the next section.



**Figure 6:** Our approach of using multiple intermediate virtual viewpoints for stitching images  $I_1$  and  $I_2$  (corresponding to cameras  $C_1$  and  $C_2$  respectively). Each black rectangle represents a pixel column in the overlap region. For clarity, we show only three intermediate viewpoints covering three different pixel columns.



**Figure 7:** *Parallax compensation in image overlap: (a) direct feathering introduces blurring, (b) magnified version of the region marked in (a), (c) stitching results after MPPS. Notice the much cleaner edges in (c).* 

### 6 Image stitching and parallax compensation

Since the camera centers of our omnidirectional capture system are not coincident, the amount of observed parallax can be significant in the image overlap regions, especially when portions of the scene are close by. Under such conditions, using simple feathering (blending) to produce a stitched omnidirectional image is not acceptable, as can be seen in Figure 7a,b.

To handle this problem, we adapted a stereo matching process to recover a *multi-perspective image* [Rademacher and Bishop, 1998] in the overlap region, where each column in the overlap region is matched using a slightly different virtual viewpoint. This idea is shown in Figure 6, where each black rectangle represents a pixel column within the overlap region formed between images

 $I_1$  and  $I_2$ . Each intermediate viewpoint (in black) covers a pixel column.

For a given intermediate viewpoint, stereo is applied to find the most photoconsistent appearance given the two input images  $I_1$  and  $I_2$ . This is accomplished by searching along a set of predetermined depths (i.e., *plane sweeping*) and finding the minimum difference between projected local colors from the input images at those depths.

We used the shiftable window technique [Kang *et al.*, 2001] in our stereo algorithm to handle occlusions in the overlap region. The size of the search window is  $3 \times 3$ . For a given hypothesized depth at a pixel in the overlap region, the corresponding windows in the two input images are first located. The match error is computed as sum-of-squared-differences (SSD) between these windows. For a given depth, shifted versions of the corresponding windows that contain the central pixel of interest are used in the computation. The most photoconsistent depth is the depth whose shifted window pair produces the smallest matching error. Note that this operation is performed on a per-pixel basis along each column. As a result, the final computed depths can vary along each column.

Because these viewpoints vary smoothly between the two original camera viewpoints ( $C_1$  and  $C_2$  in Figure 6), abrupt visual discontinuities are eliminated in the stitched image. We call this technique Multi-Perspective Plane Sweep, or MPPS for short. The results of applying this technique are shown in Figure 7c. Details of MPPS can be found in [Kang *et al.*, 2003].

# 7 High dynamic range capture and viewing

The real world contains a lot more brightness variation than can be captured by digital sensors found in most cameras today. The radiance range of an indoor photograph can contain as many as six orders of magnitude between the dark areas under a desk to the sunlit views seen through a window. Typical CCD or CMOS sensors capture only two to three orders of magnitude. The effect of this limited range is that details cannot be captured in both the dark areas and bright areas at the same time. This problem has inspired many solutions in recent years.

One method of capturing greater dynamic range of still scenes is to shoot multiple exposures, which appropriately capture tonal detail in dark and bright regions in turn. These images can then be combined to create a high dynamic range (HDR) image. Automatically combining such images requires a knowledge of the camera's response curve as well as the camera settings used for each exposure. See, for example, [Mann and Picard, 1995, Debevec and Malik, 1997, Mitsunaga and Nayar, 1999] for details of techniques to create HDR images from multiple exposures. Each of these techniques produces an HDR image with at least 16 bits per color component. An alternate



**Figure 8:** Generating an HDR image from stills. (*a*,*b*,*c*) Multiple exposures of a scene from short to long, and (d) the result of combining these exposures and tonemapping. Notice the result contains details found in the short through long exposures.

method is to combine these images directly in an image editing tool such as Adobe® Photoshop®. The following two web sites have tutorials on how to do this:

- http://www.luminous-landscape.com/tutorials/digital-blending.shtml
- http://www.digitalsecrets.net/secrets/DynamicRanger.html.

Viewing or displaying high dynamic range images is also a problem. A typical computer monitor has a dynamic range of about two orders of magnitude, and printers have even less. In order to display an HDR image, its range must be compressed to match that of the display. This operation, known as tonemapping or tone reproduction, has been explored in [Larson *et al.*, 1997, Durand and Dorsey, 2002, Fattal *et al.*, 2002, Reinhard *et al.*, 2002]. An example of generating an HDR image from multiple exposures is shown in Figure 8.

#### 7.1 High dynamic range video

In our work, we generate *HDR video*, i.e., we generate one high dynamic range image at each time interval. This is a challenging problem, since we cannot capture all the desired exposures at the same time. Furthermore, different regions in the image may be in motion. Our solution to this problem is described in [Kang *et al.*, 2003]; here, we provide a very brief overview.

Our camera allows the exposure settings to change at every frame time. We use this feature to capture a video consisting of alternating long and short exposures. We then must solve the problem of registering these frames to each other so that we can synthesize the missing long and short exposure frames at each given instant. The matching task is difficult because frames captured at



**Figure 9:** *Example frames from an HDR video sequence. The top row is the alternating exposure input video. The bottom row is the HDR video generated from this input.* 

different exposures are quite dissimilar. We facilitate the matching process by boosting (increasing) the brightness of the lower exposed images to match the higher exposed ones. Once the registration is complete, we compute a full radiance map by appropriately combining warped versions of the original video frames, using only the pixels with the most reliable correspondence and radiance values.

The radiance map is subsequently tonemapped for viewing. We adapted the technique of Reinhard *et al.* [Reinhard *et al.*, 2002] to operate on HDR video. Their technique is modified customized to use statistics from neighboring frames in order to produce a tonemapped output without any temporal artifacts.

An example of HDR video generated from a sequence of alternating exposures is shown in Figure 9.

### 8 Camera pose estimation and bifurcation handling

Despite the research efforts of recent years, none of the existing vision algorithms for camera pose recovery (ego-motion) [Hartley and Zisserman, 2000] can robustly handle large-scale environments such as the ones captured by our system. The main problem is that the error accumulation over long sequences degrades the accuracy of the camera translation.

Fortunately, in our case, full motion estimation is not necessary. In fact, accurate rotations are sufficient to obtain compelling 3D walkthroughs. Moreover, while in traditional camera systems small translations can be mistaken for small rotations, this ambiguity disappears in an omnidirectional acquisition system such as the one used in our work.

As discussed later, the accurate estimation of the viewer's orientation is useful for: (i) removing high-frequency camera jitter from the input data, and (ii) achieving smooth transitions when moving across path intersections (i.e., bifurcation points).

#### 8.1 Point tracking

Our approach to orientation estimation starts with robust, omni-directional feature tracking. Our algorithm follows a standard vision approach in terms of corner detection and robust matching [Hartley and Zisserman, 2000, Beardsley *et al.*, 1996] (see also [Aliaga *et al.*, 2003] for a system that shares some characteristics with ours). However, unlike previous approaches, our tracker is capable of following feature points that move from one camera viewpoint to an adjacent viewpoint. The algorithm that we have designed proceeds as follow:

- 1. Feature points are detected to sub-pixel precision in all six views using a Harris corner detector [Schmid *et al.*, 1998].
- 2. Since the calibration parameters are known, 2D corner positions are converted into 3D ray vectors with respect to a unique coordinate frame at the center of the camera head (Figure 10a).
- 3. For each 3D ray, points of intersections with all the cameras are computed. The image image in which the point of intersection is closest to the image center is chosen, along with a  $5 \times 5$  window of pixels centered on the point. This is done to minimize the distortion of sampled local appearance.
- 4. Pairs of 3D rays are then matched based on cross-correlation. The matched pairs are used to initialize a RANSAC [Fischler and Bolles, 1981] process for the robust estimation of the essential matrix and the removal of outliers [Hartley and Zisserman, 2000]. Once the essential matrix between two camera positions (i.e., two frames) is known, we use it to guide the matching of unmatched corners, which increases the number of good point matches without decreasing the quality of the estimated essential matrix.
- 5. Finally, the set of robust point matches obtained are concatenated along the time axis to produce sets of long point tracks.

Using 3D rays instead of 2D corner features directly (step 2 above) allows matching of features across different cameras. This, in turn, facilitates the construction of long feature tracks across time



**Figure 10:** (a) A plan view of the photo-bubble (omnidirectional image) model. A point **a** in camera 1 corresponds to a 3D ray  $\mathbf{r}_{\mathbf{a}}$  through the camera center. The two points **p** and **q** on the overlap between two cameras correspond to the same feature in space. (b,c) Feature points are detected and tracked across time and across adjacent cameras. The differently colored curves show the position of tracked points at different time instances, while the larger circles show the position of the tracked point in the current frame.

and different cameras. As can be seen in Figure 10b,c, the red track corresponding to a point on the table is an example of tracking across different cameras (from camera 1 to camera 2).

Notice that the rays  $\mathbf{r}_{\mathbf{p}}$  and  $\mathbf{r}_{\mathbf{q}}$  (defined by the points  $\mathbf{p}$  and  $\mathbf{q}$  respectively, in Figure 10a), would coincide only if the optical centers of the six cameras coincide and the camera calibration parameters were known exactly. In practice, the six optical centers do not coincide. As a result, slight discrepancies between corresponding 3D rays need to be taken into account when matching rays that project onto regions of camera visual overlap.

The large number of frames in a complete sequence (almost 10,000 frames in the smallest of our demos) and the enormous number of matched points (over 2 million tracked points) make the simultaneous recovery of structure and motion using standard bundle adjustment intractable [Hartley and Zisserman, 2000, Triggs *et al.*, 1999]. Moreover, applying bundle-adjustment directly on the tracks alone would not solve the drift problem completely. However, the estimation of the rotation component only suffices for our application.

#### 8.2 Orientation estimation and data stabilization

The robust essential matrices produced in step 4 of the point tracking algorithm described earlier are used to estimate pair-wise inter-frame rotations [Hartley and Zisserman, 2000]. In this step, rotation estimates for all pairs of images that share an adequate number of point matches are computed. After initializing the orientation estimates by chaining together the inter-frame rotation estimates,

we optimize the objective function  $\sum_{ij} w_{ij} \mathcal{N}(\mathbf{q}_i \hat{\mathbf{q}}_{ij} \mathbf{q}_j^{-1})^2$ , where  $\mathbf{q}_i$  are the absolute orientation quaternions,  $\hat{\mathbf{q}}_{ij}$  are the inter-frame rotation estimates,  $w_{ij}$  are weights based on the confidence in each rotation estimate, and  $\mathcal{N}$  is the norm of the vector component of the product quaternion (which should be close to 0 when all frames are aligned). The resulting non-linear least squares problem is solved using the standard Levenberg-Marquardt algorithm [Hartley and Zisserman, 2000].

Because of accumulated errors, rotation estimates will inevitably drift from their correct values. We reduce this effect by: (i) correcting the drift at frames where the path self-intersects, and (ii) using vanishing points, when possible, as constraints on the absolute orientation of each frame [Schaffalitzky and Zisserman, 2000, Antone and Teller, 2002].

The estimated absolute orientations are used to reduce the jitter introduced at acquisition time. This is achieved by applying a low-pass filter to the camera rotations to obtain a smoother trajectory (*damped stabilization*) and subtracting the original orientations (using quaternion division) to obtain per-frame correction. The stabilization can be either performed at authoring time by resampling the omni-camera mosaics onto a new set of viewing planes (which can potentially increase compression efficiency), or at run-time, by simply modifying the 3D viewing transformation in the viewer. We use the latter approach in our current system, since it results in less resampling of the imagery.

We have found that, with our current setup, compensating for only the rotational component is enough to achieve satisfactory video stabilization, although a small amount of lower-frequency up-and-down nodding (from the walking motion) remains. (Note that translational motion cannot be compensated for without estimating 3D scene structure.) Furthermore, we decided not to use additional hardware components for the detection and removal of jitter (*e.g.* accelerometers) in order to keep the system simpler. A related technique for stabilizing regular, monocular video sequences may be found in [Buehler *et al.*, 2001].

### 9 Compression and selective decompression

During the playback of our video tour content, it is essential for the experience to be interactive. Image data must therefore be fetched from the storage medium and rendered rapidly. For this reason, we need a compression scheme that minimizes disk bandwidth while allowing for fast decompression. There are two important requirements for such a scheme. First, it must allow for temporal random access, i.e., it must be possible to play the content forward, backward, and to jump to arbitrary frames with minimal delay. Second, it must support spatial random access so that sub-regions of each frame can be decompressed independently. This is important because we have six high resolution images and transferring this whole set at high frame rates over AGP to a



Figure 11: Block diagram of playback engine showing decoder and caching configuration.

graphics card is impossible. Our solution was to develop our own compression scheme that allows for selective decompression of only the currently viewed region and uses caching to support optimal performance.

When rendering the interactive tour, we set up a six-plane Direct3D environment map that matches the camera configuration (Figure 3b) and treat the processed video streams as texture maps. To allow for spatial random access, we divide each view into  $256 \times 256$  pixel tiles and compress these independently. This choice of tile size is based on a compromise between compression efficiency (a large tile allows for good spatiotemporal prediction) and granularity of random access. Our codec uses both intra (I) and inter (B) coded tiles. The inter tiles are bidirectionally predicted and are similar to MPEG's B-frames. These tiles get their predictions from nearby past and future intra-coded frames, where predictions can come from anywhere in a camera view, not just from the co-located tile. Although this breaks the independence of the tile bitstreams, it increases compression efficiency, and, with appropriate caching of intra coded tiles, it does not significantly affect performance. Our use of B-tiles has some similarity to [Zhang and Li, 2003].

Tiles are divided into  $16 \times 16$  macroblocks. For I-tiles, we use DCT coding with spatial prediction of AC and DC coefficients in a manner similar to MPEG-4. B-macroblocks can be forward, backward or bidirectionally predicted. There are also two "skipped" modes which indicate that either the forward or backward co-located macroblock is to be copied without requiring the transmission of any motion vector or DCT information. Additionally, forward and backward motion vectors are independently spatially predicted.

Our caching strategy (Figure 11) minimizes the amount of repeated decompression necessary during typical viewing. Tiles are only decoded for the current viewpoint and are cached in case the user is panning around. In addition YUV tiles from reference (intra) frames are cached because they are repeatedly used during the decoding of consecutive predicted frames. When a B-tile is decoded,



**Figure 12:** *Rate-distortion curves comparing our codec with JPEG compression. We used a 500 frame hometour sequence and varied the number of B-frames between each I-frame. Compression ratio is the ratio of the total bitstream size to the total number of bits in the sequence when using the uncompressed RGB format.* 

the decompressor determines which I-tiles are needed for motion compensation and requests these from the cache. If they are not available, the cache may call upon the decoder to decompress additional I-tiles.

In addition to the compression scheme described above, we have also implemented a tile-based method that uses JPEG compression with no temporal prediction. This has a greater decoding speed but a much lower compression efficiency than our new scheme. The rate-distortion curves for our codec are plotted in Figure 12, showing the signal to noise ratio against the compression factor. The JPEG compressed sequence has the worst performance as indicated by the fact that its curve is lowest on the graph. Our intra codec performs quite well, but its performance is improved by adding predicted B-frames, with two Bs for every intra frame being about optimal for this sequence. Using bidirectional coding, we are able to obtain compression ratios of 60:1 with high visual quality. With regard to performance, we are working to optimize our decoder so that the system can reach a rate of thirty  $768 \times 1024$  frames per second to support the required smoothness of interactivity. Currently we maintain about ten frames per second with our new codec but can achieve the 30 fps target with the less space efficient JPEG compression. Since our codec enables playback from media with lower data bandwidths, we are investigating the possibility of playing back our datasets from standard DVD media.



**Figure 13:** The use of maps to enhance the navigation experience: (a) Map of the botanical garden in Bellevue (one of the demos shown in the results section). (b) The approximate camera path has been drawn in red. (c) Some key frame numbers mapped to positions on the path. (d) Some audio sources captured in situ have been placed on the map. The current viewer position and orientation is marked in green.

# **10** Multimedia experience authoring

In this section, we describe how the final walkthroughs are enhanced by adding multi-media elements using simple semi-automatic authoring tools.

#### **10.1** Bifurcation selection and map control

Whenever maps of the visited site are available, we incorporate them into our viewer to enhance the user experience. For instance, to help the user navigate the space being visited, we display the current location and orientation on a map (Figure 13).



**Figure 14:** *Bifurcation handling: Given a self-intersecting path we want to find the best bifurcation, i.e., the best pair of frames*  $f_i$ ,  $f_{i'}$  corresponding to the same map location.

To enable the location display feature, we first plot an approximate acquisition path on the map (red lines in Figure 13a) during the authoring stage. Frame numbers are then associated with different key locations along this path (Figure 13a). At viewing time, given the current view (with known frame number), the 2D position of the observer on the map can then be estimated and updated using simple linear interpolation.

In places where two branches of the acquisition path intersect (e.g., the red and the green paths in Figure 14), we first manually select two ranges of corresponding frames for the transition (e.g.,  $[f_a, f_b]$  and  $[f_{a'}, f_{b'}]$  in Figure 14). Our system then automatically estimates the best intersection, i.e., the best pair of corresponding frames ( $f_i, f_{i'}$  in Figure 14). This is achieved without explicit position information. Instead, we seek the pair of frames with the minimum "visual distance" between them. Our algorithm constructs a matrix of such distances between each omnidirectional image (which we also call "photo-bubble") in the range [ $f_a, f_b$ ] and every other photo-bubble in the range [ $f_{a'}, f_{b'}$ ]. The best pair  $f_i, f_{i'}$  has the smallest distance  $d(f_i, f_{i'})$ .

The visual distance between two photo-bubbles is defined as follows:

- Given two frames  $f_1$  and  $f_2$ , we first run a point matching algorithm, estimate the rotation between the two corresponding photo-bubbles, and register them by rotating one of the two images into the other;
- We compute the residual optical flow (angles between corresponding 3D rays after rotation).
- The distance  $d(f_1, f_2)$  is defined as the median length of the residual flows.

The above distance measure is related to the parallax between the two photo-bubbles and is zero if the two photo-bubbles were taken from exactly the same location in the map (assuming a static scene). The median filter is used because of its robustness with respect to outliers (bad matches).

In our current implementation, the user must traverse a bifurcation node  $((f_i, f_{i'})$  in Figure 14) to move to a new path. An alternative would be to use new-view synthesis to generate novel intermediate views, which would allow the user to move from one path to another using a synthesized path (the



**Figure 15:** Augmented reality in our surround models: (*a*,*b*) Different snapshots of our surround walkthrough showing the living room being augmented with a video-textured fireplace and the TV set showing the program of our choice. (*c*) A photograph of the TV set with the blue grid attached. (*d*) Another snapshot of our viewer showing a different view of the living room. Notice how the occlusion of the TV set by the left wall is correctly handled by our technique.

dotted blue path in Figure 14) and could result in a more seamless transition. (View interpolation could also be used to compensate for the fact that paths may not exactly intersect in 3D.) However, because it is difficult to avoid artifacts due to occlusions and non-rigid motions such as specularities and transparency in new view synthesis, we have not yet investigated this alternative.

#### **10.2** Object tracking and replacement

We have found that the richness and realism of real-world environment visualizations can be greatly enhanced using interactive and dynamic elements. To support the insertion of such elements, during the data capture process, a planar blue grid can be placed in strategic places, e.g., in front of a TV screen or inside a fireplace (Figure 15c). During post-processing, our tracking algorithm automatically estimates the position of the target together with an occlusion mask, and this data is stored along with the video (in a separate alpha channel). During visualization, this allows us to replace the selected object with other still images, streaming video, or video-textures (for repetitive elements such as fire or waterfalls [Schödl *et al.*, 2000]) while dealing correctly with possible occlusion events. For example, we can replace the paintings in a room, or change the channel and display live video on the television set (Figures 15a-b). Note that the blue grid is only necessary if the surface may be partially occluded in some views (Figure 15d). In easier cases, we can track and replace regular geometric objects such as picture frames, without the need for any special targets in the scene.

#### **10.3** Audio placement and mixing

The tour can also be made richer by adding spatialized sound.

Audio data is acquired *in situ* with a directional microphone, cleaned up using a standard audio editing application, and then placed on the map to achieve a desired effect (Figure 13d). At rendering time, the audio files are attenuated based on the inverse square distance rule and mixed together by the audio card.

This simple technique increases the realism of the whole experience by conveying the feeling of moving closer or farther away from objects such as waterfalls, swaying trees, or pianos. In future work, we would like to investigate the use of true acoustic modeling for more realistic *auralization* [Funkhouser *et al.*, 1999].

#### **11** Scene specification and rendering

The final component of our system is the interactive viewing software (viewer). This component handles user input, the graphical user interface (visual overlays) and the rendering of video data. For each walkthrough, the authoring stage outputs a file in our XML-based scene description language (SDL). This file contains sound positions, stabilization information, bifurcation points, object locations, map information, and object tracking data. The SDL was developed as a flexible way of overlaying a user interface onto the raw video data. It allows us to rapidly prototype new user interfaces as the project evolves.

The viewer takes user input from a mouse or gamepad as shown in Figure 1. The gamepad controls are mapped in a similar way to driving games. The left joystick allows the user to pan left/right/up/down, while the left and right trigger button move forwards and backwards respectively along the path. The forward direction gets swapped as the user rotates 180° from the forward direction of capture. This causes the user to follow the original path backwards, but since we film only static environments, it appears to be a valid forward direction.

The rendering process is completely done on the PC's 3D graphics hardware. As shown in Figure 3b, the environment model is a set of six texture-mapped planes corresponding to the six camera focal planes. The positions of these six planes are specified through the SDL. This model works well for our particular camera geometry; for other camera designs, a sphere or parabaloid could be specified via the SDL. The virtual camera is placed at the center of this geometry and its rotation is controlled via the joystick.

As the user navigates the environment, the viewer computes the appropriate data to request from

the selective decompressor. The input video is often traversed non-sequentially. In the case of fast motion, the stride through the input frames can be greater than one. In the case of a bifurcation, the next frame requested can be in a totally different section of the original video. To accommodate this, we use a two stage pipeline. During a given display time, the next frame to display is computed and requested from disk while the viewer decompresses and renders the frame already in the compressed cache. The system loads all the compressed bits for a given surround frame into the compressed cache. However, only a subset of these bits actually need to be decompressed and rendered for any given view. Based on the user's viewing direction, the viewer intersects the view frustum with the environment geometry and computes the appropriate subset. Once the appropriate decompressed tiles are in the RGB cache, the visible planes are texture mapped with those tiles.

At every frame, the viewer also applies the stabilization information in the SDL file. This is accomplished through a simple transformation of the current view matrix. This could have been done as a pre-process and doing so might improve compression efficiency. However, deferring this to render time, where it is a very inexpensive operation, allowed us to fine tune the stabilizer without re-processing the entire sequence.

The SDL also specifies additional user interface elements to display. The viewer monitors the bifurcation points and displays the appropriate overlay as a user approaches one of these. Figure 16 is a screen shot of the UI for a home tour. Notice the navigation bar at the top indicating a branch point: navigating to the left leads to the dining room and to the right leads to the living room. Figure 17d shows an alternative UI in which arrows are overlaid on the environment. Figure 16 also shows an information bar at the bottom of the screen. Certain segments of the environment ("hotspots") get tagged as containing additional information. Here, the art case is in view, so icons are displayed on the bottom bar indicating additional information about these art pieces. Selecting one of the icons via the gamepad pops up a high resolution still image and an audio annotation. This user interface also contains a map, indicating the current position within the tour. The gamepad can be used to select a different room and the user is quickly transported to this new position. The position and style of all these elements are easily specified via the SDL.

### **12 Results**

This section describes the interactive visual tours that we created with our system. We captured three different environments and built interactive experiences around each of these.



**Figure 16:** A screen shot of a home tour. The navigation bar at the top indicates available directions to maneuver. The information bar at bottom indicates additional available multimedia items. The map view shows the current viewer position.







**Figure 17:** Some snapshots of our demos: (*a*,*b*) The first home tour. (*a*) A snapshot of a section of the dining area. Notice that the TV set has been augmented with a program of our choice and the occlusion of the stairs hand rail has been dealt with correctly. (*b*) A view of the living room, showing why high dynamic range imaging is desirable. (*c*,*d*) A tour of the Bellevue botanical gardens. (*c*) A snapshot from our garden model with the superimposed map. The map shows the current position and orientation of the observer together with the acquired navigation path. (*d*) When at a bifurcation, multiple arrows (in blue) appear on the ground to show the allowed directions. (*e*,*f*) An enhanced home tour. (*e*) In the final demo (also characterized by a complex, self-intersecting acquisition path) the available choices at a given bifurcation are shown in the top bar. The map still shows the current location and orientation of the living room with the indoor and outdoor details simultaneously visible thanks to our HDR video technique.

A simple indoor environment. Our first demo was a basic home tour (Figure 17a,b). The data was acquired along a single loop through the ground floor of the house. The final experience contains some dynamic elements such as video playing on the TV and flames in the fireplace. To do this, we placed blue grids in the environment at those locations. These were later tracked by our authoring software. Once tracked, the renderer projects video onto these surfaces. Figure 15 shows the resulting insertion. The setup process took approximately 45 minutes for path planning and grid placement. The acquisition was done using our omnidirectional camera mounted on a rolling tripod, as seen in Figure 4a. The entire capture process was done in one pass that took 80 seconds. The processing time from raw data to the final compressed format took several hours.

For all demos described in this section, our interactive viewer was used to allow users to experience captured environments. The viewer was run on a 2GHz Pentium PC with an NVIDIA GeForce2 graphics card in full-screen  $1280 \times 1024$  mode with a horizontal field of view of about  $80^{\circ}$ , and the compressed video data was streamed from the hard drive. Under these conditions, we are able to maintain a rendering speed of 20 frames/sec during translation and 60 frames/sec if rotating only. The difference in frame rates occurs because during translation, the system is fetching frames from disk, while during rotation, the compressed frame is already in memory.

In this first demo, we observed that users did not need to arbitrarily navigate the environment (away from the acquired path) to get a sense of the three-dimensional space. We did observe, however, that users wanted to branch at obvious locations in the house, for example up the stairs or down a hallway. Also, in this demo, many image areas are saturated. To enhance the realism of the experience, we decided that high dynamic range imaging was essential.

A more complex outdoor environment. The next environment we captured was an outdoor botanical garden (Figure 17c,d). Because of the more rugged nature of this location, we mounted the camera on a helmet. This allowed us to navigate staircases as well as uneven paths. In this case, we did more advanced planning and captured a large portion of the gardens. The planned path crossed itself in several locations. At these positions, we placed small markers on the ground so that the camera person could walk through these desired intersection points. The capture process was again done in a single pass that took about 15 minutes. Audio data was also collected from various locations in the garden.

Unfortunately, the helmet mount introduced much more jitter in the video. Therefore, in order to remove much of this unwanted motion, we developed the stabilization technique described in this paper. This increased the video processing time somewhat, which in this case took about 24 hours unattended. There was also some manual authoring involved: video frames needed to be coarsely

registered to locations on the map and the position of audio sources needed to be indicated relative to the same map (Figure 13). Another step of the authoring phase was to roughly indicate the branch points. Our automated bifurcation selection system then computed the best branch near the user selected point.

In order to indicate bifurcations to the user, arrows were rendered into the scene indicating the available directions of travel (Figure 17d). Unfortunately, we found that this method of indicating branches was somewhat confusing as the arrows were sometimes hard to find. The viewer application also rendered a blend of distance-modulated sound sources. The addition of sound considerably enhanced the feeling of presence in this virtual environment.

An enhanced home tour. Our final demo was a tour of a high-end home (Figure 17e,f). To capture staircases and outside areas, we again used a head-mounted camera. The preparation time took about 45 minutes, which consisted mainly of planning the capture route and again placing small markers. The capture was done in a single 15 minute pass.

For this tour, we also addressed the dynamic range issue by shooting with the omnidirectional camera in HDR video mode as described in Section 7. Throughout the tour, the user can simultaneously see details inside and outside the home. The resulting imagery is much more pleasing than the first demo where all the window regions were saturated.

In this tour, in order to alleviate the confusion around available branch points, we adopted a heads up display user interface. As shown in Figure 17e, branch points are now indicated in a separate area at the top of the display. We observed that users had an easier time navigating the environment using this UI.

#### 13 Discussion

Now that we have described our system in detail, let us discuss some of its advantages and disadvantages compared to other alternatives, as exemplified by previous work in this area.

Our system can be viewed as an extension of the idea first proposed in Andrew Lippman's seminal MovieMaps project [Lippman, 1980]. Because of the technology available at the time, viewers of MovieMaps had to branch between a fixed number of video clips filmed in the streets of Aspen. In our system, users have continuous control over both position and viewpoint, and the system is lightweight and portable enough so that new content can be filmed in almost any location. Interactive tours based on catadioptric and multi-camera sensors similar to ours have previously been demonstrated [Boult, 1998, Taylor, 2002]. However, the unique design of our capture and processing steps makes our system the first to provide interactive video tours of sufficiently high resolution and fidelity on a personal computer to make the experience truly compelling, i.e., to be comparable or better than broadcast video. The capture system designed by iMove (*http://www.imoveinc.com*) has a similar multi-camera design, but at a slightly lower resolution. Systems using multiple cameras and planar reflectors, such as the one built by FullView (*http://www.fullview.com*) have no parallax between the cameras, which is an advantage over our system, but result in bulkier (and hence less portable) systems. We believe that our complete acquisition, post-processing, authoring, compression, and viewing system goes well beyond what has been demonstrated with such commercial systems to date, especially in the areas of high dynamic range, ease of navigation, and interactivity.

Another line of research, exemplified by the work of Aliaga *et al.* [2001, 2002, 2003] uses a denser sampling of viewpoints combined with new view interpolation to give the user greater freedom of motion. While this is an advantage of their system, their approach requires fairly accurate geometric impostors as well as beacons to locate the camera, and still sometimes leads to visible artifacts. Moreover, the time required to scan a single room can be quite long. While our system constrains the user to follow the original camera's path, we believe this is a reasonable tradeoff, since a much larger environment can be filmed (and therefore explored) in the same amount of time.

Another competing approach is to build full 3D models of the environment (see, for example [Debevec *et al.*, 1996, Koch *et al.*, 2000, Coorg and Teller, 1999] and the articles in the IEEE Computer Graphics and Applications Special Issue on *3D Reconstruction and Visualization of Large Scale Environments*, November/December 2003). The most ambitious of these efforts is Teller *et al.*'s City Scanning Project at MIT (//*city.lcs.mit.edu/city.html*) [Teller *et al.*, 2003], in which dozens of buildings have been reconstructed from high-resolution, high dynamic range omnidirectional panoramas. While these results allow the viewer to have unrestricted motion throughout the digitized scene, they fail to capture beautiful visual effects such as reflections in windows, specularities, and finely detailed geometry such as foliage. Our approach in this project has been to retain the best possible visual quality, rather than trying to build more compact geometric models. In the next stage of our project, we plan to reconstruct 3D surfaces and layered models wherever it is possible to do so reliably and without the loss of visual fidelity, and to use these models to enable more general movement, more sophisticated kinds of interactions, and to increase compression efficiency.

# 14 Conclusion

Compared to previous work in image-based visual tours, we believe that our system occupies a unique design point, which makes it particularly suitable to delivering interactive high-quality visu-

alization of large real-world environments. Because our sensors combine high spatial and dynamic range resolution (through the use of multiple exposures and temporal high dynamic range stitching), we can achieve visualization of remarkably high fidelity, while at the same time maintaining a continuity of motion. Because we do not build 3D models or attempt to interpolate novel views by merging pixels from different viewpoints, our rendered views are free from the artifacts that plague geometric modeling and image-based rendering techniques whenever the geometry (or camera position) is inaccurately estimated. Our lightweight camera design makes it easy to film in locations (such as gardens and complex architecture) that would be awkward for cart or mobile-robot based systems. We compensate for the possible lack of smooth motion using stabilization techniques, which also enable us to automatically jump among different video tracks without visual discontinuities. To deal with the large amounts of visual data collected, we have developed novel compression and selective decompression techniques that enable us to only decompress the information that the user wants to see. Our flexible authoring system makes it easy to plan a tour, integrate the captured video with map-based navigation, and add localized sounds to give a compelling sense of presence to the interactive experience. Finally, our 3D graphics based rendering system, combined with a rich and intuitive user interface for navigation, makes the exploration of large-scale visual environments a practical reality on today's personal computers.

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