GPU assisted light field capture and processing

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What is this all about?

Ninjas! Motivation: ninjas are cool, and can do lots of tricks that we also want to.
Ninja illusions

Motivation: visualize 3D scenes in thin air.
Light field

• Amount of light in every direction through every point in space
• Can be described by the 5D plenoptic function.
• Light field displays aim to reconstruct the plenoptic function with various restrictions (e.g. discretization, RGB instead of radiance, etc.)
What are light field displays?

• Generic sense:
  – Any collection of (visible) light sources that can be used to emit light rays with known properties within a time slice, can be considered a light field display.
  – Light ray properties can include light ray origin, direction, light color (wavelength), intensity, polarity and phase
  – This includes everything from a single LED light source, 2D displays, stereoscopic and autostereoscopic displays to full parallax holographic systems.
What are light field displays?

• Restrictive sense
  – The closer the display’s capability is to emit controlled light in every direction through every point in space, the closer we are to a real light field display

• Special cases we are interested in:
  – Horizontal parallax only autostereoscopic display
  – Full parallax autostereoscopic display
The HoloVizio System

• Optical modules
  – Project light beams to hit the points of a screen with multiple beams under various angles of incidence
  – The exit angle depends only on the relative position of the given screen point and the relevant modules

• Holographic screen
  – Direction selective property with angularly dependent diffusion characteristics
  – The screen diffusion angle $\delta$ is equal to the angle $g$ between the neighboring modules

• Emission angle geometry determined
  – No optical road-blocks like at Fresnel, or lenticular lenses
The HoloVizio System

- **FOV**
  - Controlled FOV determined by the module optics and arrangement
  - Wider FOV, more complexity; the freedom given to the viewer is the challenge
  - No favorable zone in the FOV, viewers can freely move in the whole FOV area

- **Angular resolution / FOD**
  - Large number of light beams can be emitted from the same small screen pixel area
  - The angular resolution determine the FOD
  - The smallest feature (voxel) the display can reconstruct is the function of the angular resolution and the distance from the screen \( (p = p_0 + s \times \tan \Phi) \)
  - The achievable resolution is decreasing with the distance from the screen

![Diagram showing FOV and Angular resolution](image)
The HoloVizio System

- Light field reconstruction instead of views
- Specific distributed image organization
  - A module is not associated to a direction
    - The projected module image not a 2D view of the final 3D image
  - Each view of the 3D image comes from more modules
    - The number of contributing modules does not change over the FOV (no point in the FOV where only one single module image would be seen, like at multiview)
  - Distributed changes within the 3D image on different image areas
    - Smooth and continuous transition, no single border occurs between views
    - Continuous motion parallax
HoloVizio Displays

- Large-scale HoloVizio Systems
  - HoloVizio 722RC
    - 72”, 34.5 Mpixel, 16:9
    - 50-70° FOV, 0.9° Φ
    - Input: Gigabit Ethernet
    - PC-based render cluster
    - LED based projection engines
      - Better colours
      - More uniform image
      - Less maintenance
HoloVizio Displays

- The full angle HoloVizio monitor

- **HoloVizio 80WLT**
  - 78 Mpixel, 30” (16:10)
  - 180 degrees FOV
  - Total freedom 3D experience, no invalid zones, no repeated views

- 2D equivalent image resolution 1280 x 768 (WXGA)
- LED colors
- Multiple DVI inputs
HoloVizio Displays

• The world first glasses-free 3D cinema system:
• HoloVizio C80
  – 3,5 m dia Holoscreen (140”)
  – No glasses, no optical contradiction
  – LED based 3D projection unit
  – Exceptional 1500 Cd/m2 brightness
  – 40 degrees FOV
  – 2D compatible
  – Fitting cinema rooms, 3D simulators
“Screenless” HoloVizio display

Floating 3D visual reality workstation

Towards mixed reality
- Light and shadow effects
- Simulated touch
Benefits

• View with naked eye
  – No glasses
  – No tethers
  – No tracking
  – No positioning
  – No image jumps
• Wide Field Of View
• Virtual object position
  – Behind screen
  – In front of screen
• Look behind objects
• Multi-user collaboration
• 3D movie is more like a theatrical performance

In short, none of these!

She’s got the right idea!
Light field capture

- Capture equipment is a collection of light sensors that sample radiance in a section of 3D space.
- Typically an array of pinhole or plenoptic cameras are used to capture the scene.
- There are camera systems with both sparse and dense sampling along the baseline.
- Depending on the characteristics of the display, we need to adjust the design of the camera system that provides a capture as close as possible.
Ninja hiding

Motivation: hide or remove image errors (focus, occluders, etc.)
Dense light field capture

• Dense light fields are typically captured by plenoptic cameras.
• Small baseline
• Microlens array optics
• Integral imaging
• Results in a high number of 2D images.
• Both horizontal and vertical parallax are captured.
Dense light field capture

LF image by Todor Georgiev
Light field refocusing

- With dense light field capture depth of field effects can be achieved as most required ray for the effect are captured.
- Both focus and aperture of the final image can be changed by averaging the captured rays based on the ray table of the result camera.
- Use the mathematical model of the lens, or ray-trace the camera optics on the GPU to get the correct camera rays.
- Ideally suited for GPU, can be precomputed into simple lookup table.
Light field completion

- By Yatziv, Sapiro and Levoy
- Register light field image
- Generate median image from the light field to find object behind occluder.
- Removes occluders, such as fences, leaves, thin branches, etc from the final image.
- Light Field Completion Using Focal Stack Propagation by Terence Broad is also a fast and simple approach for the same task.
Sparse light field capture

- Sparse light fields are captured with pinhole camera arrays.
- Density depends on camera size.
- Can be horizontal parallax only.

Stanford Multi-Camera Array  Fraunhofer IIS LF camera
Light field conversion

- Resamples captured light field into the display’s light field.
- Display’s light rays are ray traced through the display’s optics on the GPU to provide the final rays on the observer line.
- For all display rays we can calculate the cameras that have the closest eye positions to the ray.
- By projecting the ray’s end into the camera we can define a sampling position from the camera system.
- We can generate a color for the display ray by interpolating between the selected camera pixels.
Light field conversion, rendering

- Render from converter tables, interpolate two closest cameras (GLSL).

```glsl
#version 430
uniform layout(binding = 0) sampler2DArray cameraTexture;
uniform layout(binding = 3) sampler2D lutTexture0;
uniform layout(binding = 4) sampler2D lutTexture1;
uniform int cameraCount;
in vec2 texcoord;
layout(location = 0) out vec4 color;
void main()
{
    vec4 camCoordinates0 = texture(lutTexture0, texcoord);
    vec3 camCoordinates1 = texture(lutTexture1, texcoord).rgb;

    if(camCoordinates0.w < -0.5)
    {
        color.rgb = vec3(0.0, 0.0, 0.0);
    }
    else
    {
        vec3 color0;
        vec3 color1;
        if(camCoordinates1.x >= cameraCount)
        {
            color1.rgb = vec3(0.0, 0.0, 0.0);
            camCoordinates0.w = 1.0;
        }
        else
        {
            color1 = texture(cameraTexture, camCoordinates1.yzx).rgb;
        }
        if(camCoordinates0.x >= cameraCount)
        {
            color0.rgb = vec3(0.0, 0.0, 0.0);
            camCoordinates0.w = 0.0;
        }
        else
        {
            color0 = texture(cameraTexture, camCoordinates0.yzx).rgb;
        }
        color.rgb = mix(color1, color0, camCoordinates0.w);
    }
    color.a = 1.0;
}
```
Color spaces

- Bayer RGGB, 8 bpp this is what the camera sees. Final pixel needs multiple pixels for RGB conversion.
- YCbCr planar, 12-16 bpp, some GigE cameras provide this. Can be converted by a 4x4 matrix into RGB. Usually stored in three separate texture arrays. Only need to sample a single value from the 3 color planes.
- RGB, 24 bpp, if this mode is selected, the conversion is done on the CPU on the receiver size by the IDS driver.
GPU Bayer demosaic

Based on Morgan McGuire’s work. Utilizing Malvar-He-Cutler algorithm.

```cpp
#pragma version 450
layout (local_size_x = 32, local_size_y = 32) in;
layout(binding = 0) uniform sampler2DArray cameraTexture;
layout(binding = 0,rgba8) uniform writeonly image2DArray cameraTextureResult;
uniform ivec2 redPosition;
void main()
{
    uvec2 cameraTextureSize = textureSize(cameraTexture, 0).xy;
    uint cameraCount = textureSize(cameraTexture, 0).z;
    vec2 inverseTextureSize = 1.0 / vec2(cameraTextureSize);
    vec2 currentTexCoord = (vec2(gl_GlobalInvocationID.xy) + vec2(0.5, 0.5)) * inverseTextureSize;
    vec2 alternate = vec2(mod(gl_GlobalInvocationID.xy + redPosition, 2));    // if 0,0 do red, if 1,0 green 0,1 second green 1,1 blue
    const vec4 KC = vec4(4.0, 6.0, 5.0, 5.0) / 8.0;
    for (uint i = 0; i < cameraCount; ++i)
    {
        float C = texture(cameraTexture, vec3(currentTexCoord, float(i))).r;
        vec4 Dvec = vec4(
            texture(cameraTexture, vec3(currentTexCoord + vec2(-1.0, -1.0) * inverseTextureSize, float(i))).r,
            texture(cameraTexture, vec3(currentTexCoord + vec2(-1.0, 1.0) * inverseTextureSize, float(i))).r,
            texture(cameraTexture, vec3(currentTexCoord + vec2(1.0, -1.0) * inverseTextureSize, float(i))).r,
            texture(cameraTexture, vec3(currentTexCoord + vec2(1.0, 1.0) * inverseTextureSize, float(i))).r,
        );
        vec4 PATTERN = (KC.xyz + C).xyz;
        float D = dot(Dvec, vec4(1.0, 1.0, 1.0, 1.0));
        vec4 value = vec4(
            texture(cameraTexture, vec3(currentTexCoord + vec2(0.0, -2.0) * inverseTextureSize, float(i))).r,
            texture(cameraTexture, vec3(currentTexCoord + vec2(0.0, 1.0) * inverseTextureSize, float(i))).r,
            texture(cameraTexture, vec3(currentTexCoord + vec2(-2.0, 0.0) * inverseTextureSize, float(i))).r,
            texture(cameraTexture, vec3(currentTexCoord + vec2(-1.0, 0.0) * inverseTextureSize, float(i))).r,
        );
        vec4 temp = vec4(
            texture(cameraTexture, vec3(currentTexCoord + vec2(0.0, 2.0) * inverseTextureSize, float(i))).r,
            texture(cameraTexture, vec3(currentTexCoord + vec2(0.0, 1.0) * inverseTextureSize, float(i))).r,
            texture(cameraTexture, vec3(currentTexCoord + vec2(2.0, 0.0) * inverseTextureSize, float(i))).r,
            texture(cameraTexture, vec3(currentTexCoord + vec2(1.0, 0.0) * inverseTextureSize, float(i))).r,
        );
        const vec4 KA = vec4(-1.0, -1.5, 0.5, -1.0) / 8.0;
        const vec4 KB = vec4(2.0, 0.6, 0.0, -4.0) / 8.0;
        const vec4 KD = vec4(0.0, 2.0, -1.0, -1.0) / 8.0;
        vec4 result = vec4(C * PATTERN, PATTERN * KC, KC * KC, KC * KC * KC);
    }
}
```
// Conserve constant registers and take advantage of free swizzle on load
#define kE (kA.x y w z)
#define kF (kB.x y w z)

value += temp;

// There are five filter patterns (identity, cross, checker, theta, phi). Precompute the terms from all of them and then
// use swizzles to assign to color channels.
//
// Channel Matches
// x cross (e.g., EE G)
// y checker (e.g., EE B)
// z theta (e.g., EO R)
// w phi (e.g., EO R)

#define A (value[0])
#define B (value[1])
//#define D (Dvec.x)
#define E (value[2])
#define F (value[3])

// Avoid zero elements. On a scalar processor this saves two MADDs and it has no
// effect on a vector processor.
PATTERN.yzw += (kD.y z * D).xyz;

PATTERN += (kA.xyz * A).xyzx + (kE.xyw * E).xyx2;
PATTERN.xw += kB.xw * B;
PATTERN.xz += kF.xz * F;

vec3 finalColor = (alternate.y == 0.0) ?
  ((alternate.x == 0.0) ?
    vec3(C, PATTERN.xy) :
    vec3(PATTERN.z, C, PATTERN.w)) :
  ((alternate.x == 0.0) ?
    vec3(PATTERN.w, C, PATTERN.z) :
    vec3(PATTERN.yx, C));
imageStore(cameraTextureResult, ivec3(gl_GlobalInvocationID.xy, 1), vec4(finalColor, 1.0));
Ninja equipment

Motivation: sometimes you still need lots of hardware…
The HV 48 G LFC camera array

- 48 modified IDS cameras.
- 1280*1024 resolution each.
- 4x10 Gb Ethernet connection to array.
- Arc camera setup.
Camera broadcaster

- Massively parallel, high throughput broadcaster software.
- Handles 48 IDS Gbit ETH cameras. Each camera is handled on a different thread. Per camera thread for camera driver and camera receiver.
- Per cluster node thread for cutting and transmitting camera images.
- Additional threads for synchronization.
- 2*(20+20 HT) core CPU to provide sufficient I/O.
- Overall > 100 CPU threads.
- Had to assign threads to CPU cores to prevent image loss (frame skip).
Camera broadcasting

- Camera images need to be broadcasted to rendering nodes.
- Only broadcast what the renderer needs. Calculate based on the converter lookup tables.
Generating clones

Motivation: replace real cameras with virtual ones
Sparse to dense light field generation

• For high angular resolution displays 48 cameras are not enough.
• Disparity and depth estimation methods can be used to enhance the captured light field image.
• Simple disparity estimation can be done in real-time.
• Consistent depth estimation is more costly and is used for offline processing.
Disparity estimation

- Disparity is the pixel distance between two corresponding points on a rectified stereo image pair. Camera parameters and disparity determine estimated depth.
- Several algorithms exist for disparity estimation. Most of them are GPU accelerated.
- We have implemented stereo block matching. We are experimenting with various estimation algorithms. A comprehensive list is available at: http://vision.middlebury.edu/stereo/eval3/
Disparity on GPU

- Estimate on lower resolution image first, but careful, resolution reduction in the horizontal also applies to disparity values.
- Propagate to higher resolution and multiply with the horizontal change.
- This way in every step you only need to search in small window around the current disparity, not the whole disparity range.
- Left to right and right to left disparity can differ, if both are calculated the precision can be increased.
Rectification

• Based on the calibrated intrinsic and distortion camera parameters, a pairwise rectification of camera images is needed.

• Focal length, principal point and distortion parameters are used to create an inverse mapping.
Rectification

```
#version 440
layout (local_size_x = 32, local_size_y = 32, local_size_z = 1) in;
layout (binding = 0) uniform sampler2DArray sourceImage;
layout (binding = 1,rgba8) uniform restrict readonly image2DArray destinationImage;

struct DistortionParameters {
    float principalPointXSource;
    float principalPointYSource;
    float focalLengthXSource;
    float focalLengthYSource;  // 4 float
    float principalPointXResult;
    float principalPointYResult;
    float focalLengthXResult;
    float focalLengthYResult;  // 8 float
    mat3 rectificationMatrixInverse; // 17 float
    int sourceIndex;
    int resultIndex;  // 17 float+2uint
    float distortion[4];  // 22 float+2uint
};
layout (std430,binding = 0) readonly buffer DistortionParametersSSBO {
    DistortionParameters distortionParameters[];
};

void main() {
    ivec3 storePos = ivec3(gl_GlobalInvocationID.xyz);
    DistortionParameters distortionParametersCurrent = distortionParameters[storePos.z];
    float focalLengthX = distortionParametersCurrent.focalLengthXSource;
    float focalLengthY = distortionParametersCurrent.focalLengthYSource;
    float focalLengthXResult = distortionParametersCurrent.focalLengthXResult;
    float focalLengthYResult = distortionParametersCurrent.focalLengthYResult;
    // Destination projection to destination camera coord. system.
    vec2 centralPos = vec2(storePos.x*distortionParametersCurrent.principalPointXResult,
                          storePos.y*distortionParametersCurrent.principalPointYResult)/vec2(focalLengthXResult, focalLengthYResult);
    // Destination camera coord. system to world space world space to source camera space.
    vec3 centralPosRectified = distortionParametersCurrent.rectificationMatrixInverse*vec3(centralPos.x, 1.0);
    centralPos = centralPosRectified.xy/centralPosRectified.z;
    float r2 = centralPos.x*centralPos.x+centralPos.y*centralPos.y;
    float r4 = r2*r2;
    float k1 = distortionParametersCurrent.distortion[0];
    float k2 = distortionParametersCurrent.distortion[1];
    float p1 = distortionParametersCurrent.distortion[2];
    float p2 = distortionParametersCurrent.distortion[3];
    // Undistort in source camera space.
    float sourceXCentral = centralPos.x*(1.0+k1*r2+k2*r4)+2.0*p1*centralPos.y*r2+2.0*p2*centralPos.x*centralPos.y;
    float sourceYCentral = centralPos.x*(1.0+k1*r2+k2*r4)+2.0*p1*centralPos.x*r2+2.0*p2*centralPos.y*centralPos.x;
    float sourceX = (focalLengthX*sourceXCentral+distortionParametersCurrent.principalPointXSource)/focalLengthX;
    float sourceY = (focalLengthY*sourceYCentral+distortionParametersCurrent.principalPointYSource)/focalLengthY;
    sourcePos.x = sourceXSource.xy*(sourcePosSource.x,sourcePosSource.y)/vec2(textureSize(sourceImage,0).xy);
    sourcePos.y = sourceYSource.xy*(sourcePosSource.x,sourcePosSource.y)/vec2(textureSize(sourceImage,0).xy);
    vec4 colorToWrite = texture(sourceImage, sourcePos).rgba;
    storePos.z = distortionParametersCurrent.resultIndex;
    imageStore(destinationImage,storePos,colorToWrite);
```
Image interpolation

- If a single disparity value is calculated, simply calculate the interpolated image by translating the original pixel with the interpolated disparity value.
- Resulting image will contain holes that can be filled with various image processing algorithms, e.g. inpainting.
QoE evaluation

• QoE research conducted by Holografika and using subjective evaluation has shown that simple disparity generation methods yield disturbing visual artifacts when the camera’s captured LF ray density falls below a display dependent threshold.

• Depth reconstruction algorithms however generate significantly better results from the same source LF at higher computation costs.
Multi-view depth estimation

- Estimate depth for segments only. Reduces depth search complexity. Refine depth estimated for segments. Fuse depth maps from multiple views to obtain a final result.
- (Full algorithm description has been submitted to Siggraph ASIA, we will only describe the basics.)
Superpixel segmentation

- Segmentation using simple linear iterative clustering considers color, brightness and spatial features, but no texture features.
- Multipass, iterative algorithm that searches for similar patches on the image. Very fast implementations available in CUDA, we have rolled our own using GLSL compute.
Initial depth estimation

• Uses depth sweep
• Photoconsistency is evaluated as truncated sum of squared differences.
• Photo-consistency cost values between the reference and mapped pixels are weighted based on similarity of the reference pixels colors to the superpixel color and are accumulated over the superpixel area and across the views.
Depth map refinement

• Energy minimization for smoothness and consistency.
• The smoothness term enforces spatial smoothness by penalizing inconsistency between neighbor superpixels of similar color.
• The consistency term ensures that the depth maps agree with each other.
Unstructured lumigraph rendering

- The light rays from the input views are appropriately weighted based on angular similarity and blended to form the output light rays. Additionally to the angular similarity, we detect the edges on the reconstructed depth maps and incorporate this information into the weighting function in order to attenuate smoothly the weights of the input rays that lay close to the depth discontinuities. This is done due to the reasons that depth estimation errors are mainly located at occlusions close to depth discontinuities and that pixels along the objects boundaries usually contain mixed foreground and background colors.
Conclusions

• Light field capture is hard
• Needs lots of cameras/bandwidth/CPU just to capture the original images.
• Bandwidth can be reduced by selecting “what we need”
• Bandwidth can be reduced by transferring “raw” camera images
• Bandwidth can be reduced by replacing real cameras with virtual cameras
• This can vastly increase quality for offline processing.
• If we have the correct depth estimation, we can render LF images directly, we don’t need to do costly image interpolation
Contributors

• Aleksandra Chuchvara, Tampere University of Technology, on multi-view depth estimation.
• Aron Cserkaszky, Holografika, on disparity estimation and QoE related algorithms
• Gábor Puhr, Holografika, on disparity estimation related algorithms
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Ninja images taken from:
Question?