

# 16-682 - MRSD Project II | ILR #07 Individual Lab Report #07 |February 16, 2017

SAMBUDDHA SARKAR

Team G eXcalibR

Huan-Yang Chang Man-ning Chen Sambuddha Sarkar Siddharth Raina Yiqing Cai

# **1. INDIVIDUAL PROGRESS**

#### 1.1 Overview

In this ILR I would describe about trajectory generation, multiple image capture and exporting data from the virtual scene. The virtual environment is being modeled in an open source platform: Blender 3D 7.68a. It is a Maya based platform and is programmable by Python 3. Topics covered have been listed below for a quick overview.

1.2 Test Trajectory

1.3 Image Capture (100 samples)

1.4 Exporting Camera Extrinsics & Intrinsics

## 1.2 Calibration Target Mesh Model: Icosahedron

Using the model editor, a trajectory was created in the virtual scene with something called a NurbsPath (it is basically an editable vector with user desired subdivisions or 'control points'). This trajectory is shown in Fig. 1.2.1. The Icosahedron was then constrained to travel along the path specified by the NurbsPath. The increments in its motion were not synchronized with the camera shutter as to simulate real world motion blur. This helped us to pin down the window of rate of change of position of the calibration target so as to avoid motion blur, the results were desirable. The multi camera view of the path is shown in Fig. 1.2.2.



Fig. 1.2.1, Trajectory of calibration target



Fig. 1.2.2, Multi-camera positioning for Trajectory

# 1.3 Image Capture (100 samples)

The multi-camera locations were locked and the calibration target was animated to move along the NurbsPath and cameras were scripted to capture 100 images. The path was chosen such that the calibration target covered the whole FOV(field of view) of the respective camera. The FOV of camera 1 is shown in Fig 1.3.1. The rendered FOV of camera 1 is shown in Fig. 1.3.2. The orientation of the faces on the calibration target is very specific and this has to mapped exactly to the designated vertices of the loosahedron. This mapping has to be exploited so that we know the geometric calibration is working. So the path was designed in such a manner that the target changed its orientation in 3D space. Fig. 1.3.3 shows 100 rendered images from the FOV of camera 1 and you can see how the face angles change throughout.



Fig. 1.3.1, FOV of Camera 1 (Not Rendered)





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Fig. 1.3.3, 100 Images (Post Rendering)

#### 1.4 Exporting Camera Instrinsics & Extrinsics

The camera Intrinsics and extrinsics were exported from blender using python. The script is shown on the left and the output on the left in Fig. 1.4. The mesh data of the calibration target is exported in form of surface normals and centre of the faces.

<pre>y.context.space_data.viewport_shade = 'WIREFRAME'</pre>	CAMERA INTRINSICS
<pre># # EXPORT CAMERA DATA\start} # f = open('6:/Blender/camera data.txt', 'w', encoding= 'utf-8')</pre>	Camera.000 focal length = 30.000000mm sensor width = 14.131200mm
#	sensor height = 10.350000mm Resolution (Width x Height) = 4096.000000X3000.000000 px
f unite//CAMEDA INTDINCTCCVe/	Camera.001
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for cam in bpy.data.cameras: f.write('\n%s\n focal length = %fmm\n sensor width = %fmm\n sensor height = %fmm\n Resolution (Width x	sensor height - 18.00000mm Receiving (Midth > Height) - 4005 00000000000000000000000000000000
Height) = %fX%f px\n' %(cam.name, cam.lens, cam.sensor_width, cam.sensor_height,bpy.data.scenes["Scene"].render.resolution_x,	Company (Martin X Height) - 4050.0000000000000000000000000000000000
<pre>bpy.data.scenes["Scene"].render.resolution_y)) 1</pre>	focal length = 30.000000mm
2 #{CAMERA INTRINSICS\end}	sensor height = 18.00000mm Recolution (Hidth x Height) = 4006 00000000000000000000000000000000
	Resolution (which x height) - 4050.0000000000000 px
#{CAMERA EXTRINSICS\start}	CAMERA EXTRINSICS
f.write('\n\nCAMERA_EXTRINSICS\n')	Camera.000 is located at:
num_camera = len(bpy.data.cameras)	x = 57.837948cm
for idx in range(0, num_camera):	z = 30.793760cm
<pre>cam = bpy.data.objects[idx] #Location</pre>	w = 0.839586
<pre>f.write('\n%s is located at:\nCartesian\n x = %fcm\n y = %fcm\n z = %fcm\n' %(cam.name, cam.location.x, cam.location.y, cam.location.z,))</pre>	x = 0.534598 y = 0.051800
#Orientation f write 'Quaternion\n w = %f\n x = %f\n x = %f\n z = %f\n' %(cam rotation quaternion w	z = 0.081353
<pre>cam.rotation_quaternion.x, cam.rotation_quaternion.y, cam.rotation_quaternion.z))</pre>	Camera.001 is located at: Cartesian
#{CAMERA EXTRINSICS\end}	x = 73.238808cm y = 10.035370cm
	z = 5.332790cm Quaternion
2 f.close() 3 #(EXPORT CAMERA DATA\end)	w = 0.324192 x = 0.206426
	y = 0.495849 z = 0.778731
ан	Camera.002 is located at:
	<
#{EXPORT MESH DATA\start}	target_deta.txt - Notepad
	File Edit Format View Help
#Scene Information scene info = bpy.context.scene	Image Number 0
e Mach	Normal of Face 0 is: <vector (0.1876,="" -0.5774,="" -0.7947)=""> Normal of Face 1 is: <vector (0.6071,="" -0.7947)="" 0.0000,=""></vector></vector>
<pre>ico_obj = bpy.data.meshes['Icosphere']</pre>	Normal of Face 2 is: <vector (-0.4911,="" -0.3568,="" -0.7947)=""></vector>
	Normal of Face 4 is: <vector (0.1876,="" -0.7977)="" 0.5774,=""></vector>
0 1 f = open('G:/Blender/target_data.txt', 'w', encoding= 'utf-8')	Normal of Face 5 is: <vector (0.9822,="" -0.1876)="" 0.0000,=""> Normal of Face 6 is: <vector (-0.7946,="" -0.1876)="" -0.5774,=""></vector></vector>
f.write('MESH DATA OF TARGET PER IMAGE')	Normal of Face 7 is: <vector (-0.7946,="" -0.1876)="" 0.5774,=""> Normal of Face 8 is: <vector (0.3035,="" -0.1876)="" 0.9342,=""></vector></vector>
for frame in range(scene info.frame start, scene info.frame end+1):	Normal of Face 9 is: <vector (0.7946,="" -0.5774,="" 0.1876)=""></vector>

Fig. 1.4, Exporting blender data

# 2. CHALLENGES

There were no challenges yet. But as the number of Cameras in the scene will increase, challenges in computational power can be expected in the future. The solution for tackling this is building our own render farms by slaving multiple GPU's to one master computer and distribute the rendering over these GPUs. Another solution is renting render farms (a.k.a. online servers) online and using them to compute the images. The second option is less work, but the former option is just so cool. Though I think Oculus will lean towards the second option i.e. using online server farms.

## **3. TEAM WORK**

The project work was divided among the team members and the task was assigned according to the strengths of the team members. The task division has been listed below in Table 3. The divided tasks can be completed in parallel; hence others can pitch in when some team members fall behind in their work.

Team Member	Task
Huan-Yang Chang	Robot Studio Un-optimized Path generation
Siddarth Raina	Formulating metrics for effective noise
	suppression or removal
Man-nig Chen	Color Calibration: Macduff technique
Yiqing Cai	Test trajectory for sampling using 4 cameras
	and creating heat map to demonstrate
	coverage of FOV
Sambuddha Sarkar	Trajectory generation, multiple image capture

Table 3, Task Division.

#### **4. FUTURE PLANS**

My future plans until the next progress report is to come up with a pipeline for image generation using blender. Also exporting the face pattern geometric data to validate the geometric calibration algorithm. The development and deployment of this pipeline will take about 90 days.

## 5. BONUS: IMAGE RENDER SAMPLES

Some image samples from the 100 sample range. (showing the drastic changes in the orientation of the target to test the robustness of the geometric calibration algorithm: this is from camera 2)





