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Development of a Virtual Reality Cycling Simulator

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Abstract

This paper presents a cycling simulator implemented using consumer virtual reality hardware and additional off-the-shelf sensors. Challenges like real time motion tracking within the performance requirements of state of the art virtual reality are successfully mastered. Retrieved data from digital motion processors is sent over Bluetooth to a render machine running Unity3D. By processing this data a bicycle is mapped into virtual space. Physically correct behaviour is simulated and high quality assets are added to create a full immersive virtual reality experience. Our whole source code is published under an open source licence on github. Ultimately, the cycling simulator will be used for behavioral studies to measure how different street design configurations and traffic levels impact perceived cycling stress level, cognitive reactions and mobility behaviour.

Index Terms

Computer graphics, simulation, game engines, human mobility, virtual reality, movement tracking

I. INTRODUCTION

Urban transport policies are shifting their focus to active modes of transportation: walking and cycling. To make these modes of transport attractive for a wide audience, transport infrastructure must cater for the needs of active mobility for a safe and comfortable cycling experience. Incorporating such infrastructure in the existing urban fabric and adapting existing street designs proves to be a considerable challenge, especially with regards to the limited available road space and budget constraints. Urban designers and transport planners who want to improve the infrastructure conditions in particular for cycling often face difficult trade-offs, since there is a lack of understanding with regards to basic questions how different road users perceive and value possible infrastructure scenarios. To address this matter the right tools need to be chosen. According to Simpson[17] creation and interaction within a three dimensional visual simulation is the key to modern urban planning. Recent research also shows that it becomes feasible to develop immersive virtual reality for spatial planning support. The advantage of these systems is the sensation of being in the environment [18]. Also can such an environment be used to measure actual perception and discuss design decisions within the scenario. By taking these factors into account we developed a functional virtual reality cycling simulator prototype within the environment of a state of the art game engine. New experimental street designs can easily be imported into our existing framework. Those can be enriched with additional assets and other objects provided by Unity to create a more realistic environment. With built-in measurement and evaluation tools proper scientific data can be extracted.

To achieve all this, a robust hard- and software framework is developed which is able to track all essential parts of the bike. Technologies like Bluetooth Low Energy, Gyroscopes and Accelerometers were combined into small tracking sensors. Furthermore all components achieve the performance required by state of the art virtual reality applications. The results of our development brought us a good tool for further research. In its current state, the cycling simulator is capable of supporting behavioral studies to measure how different street design configurations and traffic levels impact perceived cycling stress level, cognitive reactions and mobility behaviour. The source code is published on github under MIT license [4].

II. RELATED WORK

In this section we introduce existing work in the area of bicycle simulation, followed by problems of the latest generation of virtual reality technologies and how different approaches handle them. Furthermore we introduce a theory that explains why motion sickness occurs during VR simulation.
A. Existing projects within the area of VR bicycle simulation

The idea of a biking simulation within a virtual environment is not new. Multiple researchers have successfully implemented prototypes over the years. We will give an overview of those implementations.

Schulzyk et al. built a high-end bicycle simulator with a motion platform which simulates physical forces and the street. This allowed them to create a higher depth of immersion in combination with a three walled cave [15]. More similarities to our simulator has Mestre et al. with their research about the effect of virtual reality in sportive applications. They also used a stand and not a motion platform, but they have no further sensors to achieve more freedom in the virtual scene [9]. Furthermore Ranky et al. developed a framework which allows to transform every bike into a cycling simulator by adding various sensors all over the bike. They used a monitor as output, which does not create a very immersive environment [14]. Another bicycling simulator was developed by Hurwitz et al. at the Oregon State University. The bike was mounted on a movable platform and a large projection screen was used for the virtual environment [6]. There are also projects which propose similar sensors for tracking as we do, but not in the cycling context. Kim et al. developed a data glove which was able to motion track the users hand. They also used accelerometers and Bluetooth to achieve the movement tracking [7].

A great inspiration and knowledge source for this project was the cycling simulator CyclingSPACES [2] developed by Atlantis Games in collaboration with the Breda University of Applied Sciences - NHTV [10].

B. Technical limitations with state of the art virtual reality

After many years of development and an existence as niche product VR found its way into the consumer market. Small form factor displays with high pixel densities are widely available on the market and are used to build modern Head Mounted Displays (HMDs). Nevertheless there are still technical issues like the lack of performance. The graphics hardware needs to render a resolution equivalent to 3024x1680 pixels for VR games and applications, and need to do so at a sustained 90FPS [13]. The amount of pixels is therefore ~ 2.5 times higher than what is required by common display hardware with a resolution of 1920x1080. The 3D graphics acceleration manufacturer NVIDIA [12] proclaims that any performance dip, even the slightest stutter, could cause discomfort and ruin the VR experience. Technologies like Asynchronous Shaders [1], Simultaneous Multi-Projection [12], Lens Matched Shading [11] and many more try to improve this situation. The goal is to decrease bottlenecks by improved load distribution, gain performance by executing complex operations within a single render pass, and decrease the amount of thrown away, but still rendered pixels.

Not just graphics hardware needs to reach new performance regions, but also input devices need to improve. Input lag needs to decrease until it can not be perceived by the user.

C. Challenges for movement in virtual space

A well known problem in virtual environments is the fact that users can get sick. This kind of sickness is called motion- or cyber-sickness. Vertigo, headache and feeling of nausea can occur after a few moments in VR space.

The most accepted theory is the so called Sensory Conflict Theory [8]. It is based on the premise, that discrepancies between the senses which provide information about the body’s orientation and motion cause a perceptual conflict which the body does not know how to handle. Two senses are included in this process, the vestibular and visual sense. By looking at a virtual environment and perceive movement in the peripheral vision a sense of vection\(^1\) is created. Hence the visual system tells the body, that he or she is moving into a certain direction. However, since this is actually not true, the vestibular sense does not provide any information about acceleration or deceleration. Based on experience, this phenomenon does not exist during actual physical driving, which leads to the consequence that a conflict occurs and cyber sickness may ensue.

III. Overview

In this section we introduce our bicycle simulator framework. We also talk about the sensors, which are needed for tracking and where to mount them.

A. Simulator overview

Our framework for the biking simulator is split into the following parts:

- **Software**
  - 3D models and assets
  - Traffic simulation
  - 3D compatible game engine

- **Hardware**
  - Bicycle

\(^1\)medical expression for impression of self motion
Fig. 1: Biking simulator (top to bottom): HMD tracking system, HMD as user output, bike with sensors for tracking as software input and resistance unit with bike stand.

- Bike stand with resistance unit
- Sensors and tracking system
- Head Mounted Display

This paper will focus on the hardware and all the software which is needed to transfer and transform the sensor and tracking data into VR space. We call this part the Controller (Fig. 1), which then can be used as an independent unit in any 3D scenario. Traffic simulation, 3D modeling decisions and scenario designs are depending on the research question of the study which uses the Controller. Hence they are different from case to case and independent to this work and therefore not discussed.

B. Design decisions and requirements

To enable the best possible VR experience multiple aspects need to be considered. Bicycle parts that are important for our simulator need to be identified and suited sensor technologies need to be found.

By considering the characteristics of a bicycle all movable parts need to be tracked. The reason behind this is clearly that the participant can fully interact with the bicycle. Another important point is the tilting angle of the bike, because a cyclist always tilts the bicycle during a change of direction. We conclude that for a full coverage of a bicycles possibilities in VR space, these parts need to be tracked:

- Handlebars
- Rear wheel
- Pedals
- Seat axis

All these components are parts of the actual bike, which means that their relative position to the bicycle stays always the same. Hence only the rotations of these parts need to be detected. Well suited for this job is a Digital Motion Processor (DMP), which is a Gyroscope and an Accelerometer on the same chip. Although it is required, that the handlebar tracking has a lower latency than the reciprocal of the refresh rate of the used HMD. The reason behind this, is that delayed direction changes may lead to motion sickness. Fig. 2 shows all the sensors attached to the bicycle.

We used the HTC Vive as HMD because it provides a superb tracking system. Also are the Vive controllers (device (a)) capable of deliver the fast response required for the steering mechanism. Devices b) to d) are DMPs which send movement data to the render computer. Commercially available hardware parts within a reasonable price range, were combined to build sensors, which perfectly fit into the specific use case. (b) measures the tilting of the physical bike, while (c) and (d) measure
Fig. 2: Sensors attached to the bike. a) Vive controller for handlebars tracking. b) Sensor for bike leaning. c) Wheel rotation sensor. d) Pedal turn sensor e) Resistance unit

the wheel and pedal rotation. e) is the resistance unit which reacts to slopes and other physical forces. Positive slopes in the 3D scene increase the cycling resistance.

On the software side we used Unity as 3D game engine. This makes it possible to use the C# scripting system for our own software, which allows a communication with external hardware and evaluation of its data stream. On the other side can urban planners or architects import their designs without further knowledge about computer graphics. Additional information about the implementation is described in the next section.

IV. IMPLEMENTATION

In this section we describe details about the implementation of hard- and software and discus design decisions and their advantages.

A. Hardware

To overcome performance problems mentioned in section II-B, suitable components for real time purposes had to be chosen. This means that reaction times of those components combined have to be lower than the time needed to render a single frame. Hence all those components need to do their job, if possible, in hardware. The Arduino platform was chosen due to its flexible possibilities such as simple attachment of sensors and multiple communication channels. The microprocessor board needed to be small, so we went for the Adafruit Feather. Also built-in control electronics for Lithium-ion batteries, such as charging and low-voltage protection were key requirements.

A suitable DMP needed to be found within the required price range. We chose the MPU6050 because is has a good accuracy and is supported by multiple compatible C++ libraries. It communicates over Inter-Integrated Circuit (I²C), a well known and by the Feather supported bus protocol. The MPU6050 also provides an output where interrupt signals are sent to the microprocessor. With this feature we can decrease the input lag even more, by immediately processing the data when it is available and doing the rest when no data is ready. The last component was to find a fitting communication channel. Of course a cable would be the fastest but also most complicated way, because of movable parts. The decision towards to Bluetooth LE was made because of the low energy consumption, but also because of the sufficient data transmission rates of 1 MBit/s. The chosen device was the Adafruit EZ-Link which is capable of opening a Communication port (COM). By combining these components on a single printed circuit board we finalized the design for a DMP based sensor. Fig. 3 shows the block diagram.
B. Bluefruit Firmware

To fulfill the performance requirements mentioned in section IV-A a fast firmware needs to do its job and therefore is the usage of hardware interrupts an important factor. These interrupts are rare, because they need many additional circuits and all of the pins are occupied in our application by communication buses already. The only way to still use interrupt routines are so called Pin Change Interrupts (PCINT). These interrupts get triggered by any logic level change on the whole port, which means that up to 8 pins can trigger this event. By defining a port mask, only interrupts on the chosen port will be executed and the rest will be discarded.

The flow diagram in Fig. 4 shows the whole firmware. Two interrupt routines can be seen as independent blocks. These can be called at any time after the initialisation phase. The firmware is designed as an interrupt based state machine. As soon as an interrupt gets triggered, global volatile variables change the behaviour within the main loop and a notification will be sent over the available communication channel. DMP data is sent as a binary coded quaternion to prevent performance loss through string parsing operations.

C. Unity Software

As mentioned in subsection III-A the Biking Simulator uses Unity as its 3D engine. Hence provided tools by the Unity ecosystem are used in our software. The Serial communication part is written in plain C# and does not need any Unity related components. Furthermore this part needs the full .NET 2.0 framework and not just the (by Unity default) .NET 2.0 subset. This limits the compatibility to other platforms than Microsoft Windows, which is no issue due to HTC Vive minimum system requirements. Serial communication gets started by Unity if external hardware needs it. It runs in its own thread and communicates over two queues, one for incoming and the other for outgoing communication. The Serial framework receives data as a byte stream which needs to be interpreted. The data receives a header and gets enqueued. Unity, on the other side, is dequeuing data and uses it to update animations such as pedal movement and wheel spinning. The whole procedure is an implementation of the Producer-Consumer Pattern.

As soon as the movement data is complete a Unity Monobehaviour class updates the virtual scene. These classes have a predefined Update method which is called once every frame from the game engine. We use this fact to update our virtual bike with the latest movement data from the DMPs.

As mentioned in III-B we are tracking rotations. To get useful angles the data needs to be prepared. As a first step we need to translate the rotation of a tracked object into its local space. The reason behind this is the direction of the physical bike itself. The simulator should still work after turning the bike. Hence we define a reference quaternion, which is the first incoming data from a sensor. All following incoming quaternions can be translated with the following equations:
Fig. 4: Bluefruit firmware flow diagram

\[ Q = q_0 + i q_1 + j q_2 + k q_3 \]  \hspace{1cm} (1)

\[ Q^{-1} = \frac{q_0 - i q_1 - j q_2 - k q_3}{q_0^2 + q_1^2 + q_2^2 + q_3^2} \]  \hspace{1cm} (2)

\[ T = Q \ast R = t_0 + i t_1 + j t_2 + k t_3 \]  \hspace{1cm} (3)

where

\[ t_0 = (r_0 q_0 - r_1 q_1 - r_2 q_2 - r_3 q_3) \]
\[ t_1 = (r_0 q_1 + r_1 q_0 - r_2 q_3 + r_3 q_2) \]
\[ t_2 = (r_0 q_2 + r_1 q_3 + r_2 q_0 - r_3 q_1) \]
\[ t_3 = (r_0 q_3 - r_1 q_2 + r_2 q_1 + r_3 q_0) \]

and

\[ t \cdot i = q \cdot j \cdot r \cdot k = -(r \cdot j \cdot q \cdot k) \]
\[ t \cdot j = q \cdot k \cdot r \cdot i = -(r \cdot k \cdot q \cdot i) \]
\[ t \cdot k = q \cdot i \cdot r \cdot j = -(r \cdot i \cdot q \cdot j) \]
\[ i \cdot i = j \cdot j = k \cdot k = -1 \]

\[ Q_{local} = (Q_{reference})^{-1} \ast Q_{global} \]  \hspace{1cm} (4)

Note that this operation is non-commutative and therefore the factors need to stay in the right order. \( i, j \) and \( k \) are imaginary components, hence the multiplication of those results to \(-1\).

Every sensor rotates around a specific axis in its local space. To find this axis we visualised \( Q_{local} \) by setting it to the Transform component of a random mesh. By rotating the actual sensor a clear rotation around one axis gets visible. Depending on the chosen axis different computation need to be done. Listing 1 shows this behaviour.
if (sensorAxis == RotationAxis.X)
{
    Vector3 v = q * Vector3.forward;
    angle = Mathf.Atan2(v.y, v.z);
}
else if (sensorAxis == RotationAxis.Y)
{
    Vector3 v = q * Vector3.forward;
    angle = Mathf.Atan2(v.x, v.z);
}
else if (sensorAxis == RotationAxis.Z)
{
    Vector3 v = q * Vector3.up;
    angle = Mathf.Atan2(v.x, v.y);
}

Listing 1: C# code snippet which computes the absolute angle between the reference and the actual quaternion given an axis.

As seen on line 4, 9 and 14 in Listing 1, we always use the vector components which are axis independent to calculate the rotation. Another point is that at least one component of the multiplied vector needs to be axis independent, hence on line 13, the up-Vector instead of the forward-Vector is chosen. The reason behind these two constraints is a simple one, imagine the rotation around the specified axis. A plane which has the specified axis as its normal, is defined by two other axes. By rotating around the specified axis the vector components of the two other axes change. Hence we need to extract the rotation out of the two components which are dependent to this plane. The whole code is available on Github under MIT license[4].

D. Signal data smoothing methods

After retrieving all absolute angles the animations and other movements can be updated. Due to the fact that the sensors send data asynchronous and suffer from typical problems like drifting, some smoothing algorithms need to be used. This is an important factor against motion sickness, twitchy movements are not natural and enhance indisposition.

A typical input signal is visible on the top in Fig. 5. The amplitude often falls to zero and raises again to a doubled or even tripled mean value. This can be explained by the frame rate of the Unity application. When no sensor data is available within two or three frames, then the angle between those frames will be zero. In the next available data frame, the value of the angle adds all the missing angles. The amount of high frequencies in the raw signal is simply too high to use it directly for movement calculation.

To overcome this problem we tried two methods with different settings.

1) Median filtering: By looking at the raw signal at the top region of Fig. 5, we can see that in between of those spikes we often have the actual needed amplitude. Hence we take the median, if the number of samples is even we define the median as mean between the two middle values. Given the premise that positive and negative spikes have the same probability of appearance, the actual amplitude has to be in between. As we can see in the middle curve of Fig. 5, the results were not very satisfying. Taken ten samples a lot peaks were unfiltered because more positive than negative peaks occurred in the raw signal. This problem minimizes with more samples but never totally disappears.

2) Average filtering: The idea was that the signal will have the same pattern over time, which means that the average of the last n samples should not alter drastically. This assumption turned out working quite well which is visible in the bottom curve of Fig. 5. The principle works like a typical low pass filter, more samples reduce high frequencies.

V. PERFORMANCE

In this section we deal with response times, delays and performance trade-offs. In the first subsection we talk about the response times of our Bluefruit sensors, followed by a discussion about performance versus quality.

A. Sensor response times

The last biking simulator prototype built by Atlantis Games [2] had an input lag of up to two seconds because of the internal hardware design of the Elite Axiom. This was a great issue and needed to be improved. Especially during the breaking process, where reaction time is essential, two seconds are not sufficient.

The reaction times of our Bluefruit sensors are shown in the table below:

Considered this data we achieve up to 30 times lower delay than previous implementations. The number of delivered valid quaternions per second is 68, which is high enough. The sensor operates with a bit rate of 19200bit/s, even better performance can be achieved by increasing this up to 57600bit/s (maximum numbers in table 1). Note that the error rate also increases with the transmission speed in the 2.4GHz frequency spectrum. Hence the rate of 19,200bit/s is a good trade-off between speed and transmission errors.
Fig. 5: Sensor signals with ten samples on the left, with 20 samples on the right hand side. Top to Bottom, Raw input signal without smoothing, median filter which still has spikes, average filter which works fine.

Fig. 6: In game view of the bicycle simulator. Redesigned scenario with biking lanes in Tiong Bahru - Singapore. In the left half is the physical bicycle with a user wearing a Head Mounted Display.

B. Smoothing performance comparison

We introduced two smoothing algorithms in section IV-D. Results were shown with 10 and 20 samples, which taken the video frequency of 90Hz (HTC Vive), correspond to \( \sim 111 \) and \( \sim 222 \) milliseconds. Both smoothing algorithms can be implemented very efficiently and therefore just the number of samples are important. The best example is the breaking mechanism. If you immediately decrease the speed from 30 km\( \text{h}^{-1} \) to 0 km\( \text{h}^{-1} \), the bike will take the chosen \( n \) samples in frames to stop. It is anyway not realistic that a bike stops that drastically, hence \( \sim 222\text{ms} \) are not much and need to be extended by additional braking time. On the other side if the bike would brake immediately because of extremely low speeds the delay could be felt by participants.
Table I: Cycling simulator performance data: *Bluefruit* vs. *Elite Axiom*

<table>
<thead>
<tr>
<th></th>
<th>Typical</th>
<th>Min</th>
<th>Max</th>
<th>Elite Axiom</th>
</tr>
</thead>
<tbody>
<tr>
<td>Response Delay [ms]</td>
<td>80</td>
<td>65</td>
<td>100</td>
<td>up to 2000</td>
</tr>
<tr>
<td>Data frequency [Hz]</td>
<td>68</td>
<td>63</td>
<td>74</td>
<td>~60</td>
</tr>
</tbody>
</table>

VI. LIMITATIONS

As mentioned in chapter II-C, the lack of physical forces increases the probability of getting motion sick. Braking and accelerating feels strange or can make the user sick because of a conflict of vestibular and visual senses. A hydraulic platform which simulates physical forces would help to decrease this issues. How much effort is needed to implement such a platform is despite of the tricky bicycle physics not known. Maybe it would be infeasible to overcome the problems of motion sickness. Physical resistance is the next problem of the current implementation. The *Elite Axiom* bike trainer reacts way too slow, in worst case up to $\sim 7s$, to adjust the pedalling resistance. Furthermore it can not accelerate the wheel, which is a problem when the bike is driving downhill in the virtual environment. While going downhill the physical wheel stops because the participant does not have to pedal to moving forward. When the participant then tries to brake the breaking is not recognized as the wheels are standing still. Hence to create a more realistic simulation the wheel needs to be accelerated by an engine and limited by a real time resistance unit. Another approach is to install a sensor at the brake to solve the downhill drive problem. The disadvantage of this would be that the wheel “ticking” noise would be missing or needed to be simulated by software.

VII. CONCLUSION

We show a possible way to build a state of the art bicycle simulator with reasonably priced and commercially available hardware. We talk about the problems of modern *Virtual Reality* systems and describe how motion sickness occurs. With this information in mind, we present fast *Digital Motion Processor* based sensors for tracking purposes. Program code is provided which makes all sensor data usable in the *Unity* game engine environment. In game scenes and the actual physical simulator is shown in fig. 6.

On the other hand many functions can be added to our present system. A hydraulic platform which provides force feedback would address the issues of motion sickness, and a system which is capable of accelerating and decelerating the wheel in real time would improve functionality of the *Bicycle controller*. Our future plans with the simulator include the integration of the controller and traffic simulation in a game play that is designed to study cognitive processes and behavioural reactions while cycling in different street design and traffic level scenarios. To this end, we also will enhance the current set-up with audio and human body sensor measurements.

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Stefan Müller Arisona is a professor of computer science at the Institute of 4D Technologies (i4Ds) of FHNW Switzerland, and co-director of the CIVAL. He received his PhD from the University of Zurich in 2004, and was a post-doctoral fellow at ETH Zurich (2005 2007) and UC Santa Barbara (2007 2008). From 2008 2010, he was software architect at Procedural Inc., and is co-author of the CityEngine generative modelling software, which was nominated for the Sci-Tech awards of the Academy of Motion Picture Arts (Tech Oscars) in 2015. His research focuses on combining principles from computer graphics, digital media, and human-computer interaction and to apply them to a variety of fields such as architectural and urban modelling, digital art and entertainment, or digital media authoring.

Michael Joos is a software engineer specialising in real-time 3D applications. During his university years, he took an internship at SIEMENS’ Corporate Technology R&D department in Munich, where he first started to develop 3D interactive applications. After finishing his studies, he took on the role of software engineer for the video games industry for nearly 10 years. He has also taught computer science courses at the DigiPen Institute of Technology, and worked for a social media company as a full stack engineer. Given his experience with interactive virtual 3D environments, he helps the different teams at the Future Cities Laboratory with their project requirements, such as generating 3D visualisations of heat emissions for the Cooler Calmer Singapore project, develop a cycling virtual-reality simulator with the Engaging Mobility team, and guide and mentor computer science interns.
Alexander Erath leads the research project Engaging Mobility at the Future Cities Laboratory. Before, he coordinated FCL’s research group Mobility and Transportation Planning and led in this role the implementation and further development of the large-scale, agent-based transport demand model MATSim Singapore. He obtained his PhD in 2011 from ETH Zrich (Swiss Federal Institute of Technology) where he studied the vulnerability of transport infrastructure. His main research interests are active mobility, the interaction between transport infrastructure and the built environment, travel behaviour, accessibility-based transport planning and multi-agent, activity-based transport demand modelling.