



SCIENCE AND TECHNOLOGY

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ABSTRACT

In recent microprocessors or ASIC chips, the operating frequency is set by the target market. This leads to very tight timing and power constraints for the proposed circuit design. The industrial shift for adopting lower technology nodes also presents a new challenging frontier as transistors get less efficient as they undergo scaling. Analog designers are expected to optimize these conventional designs and yet meet the reduced power constraints and performance metrics imposed by various applications.

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Chapter 1

Introduction

The environmental impact of global warming has accelerated the interest to adopt non-conventional power generation [1]. There are several promising sustainable energy alternatives, among which the adoption of Thermoelectric (TE) materials to scavenge the by-product heat generation is widely accepted [1]. It is crucial for applications associated with energy harvesting to possess a high figure of merit $ZT (\geq 1)$ [1].

The ZT can be related to other thermoelectric parameters by $ZT = (\frac{\sigma_e \times S_B^2 \times T}{\kappa_{ph} + \kappa_e})$, where σ , S_B , $\kappa_{ph} + \kappa_e$, and T are the electrical conductivity, Seebeck coefficient, total thermal conductivity, and temperature value respectively [1].

Experimental and theoretical identification of two-dimensional (2D) [1] or three-dimensional (3D) efficient TE materials is laborious and time inefficient [1]. It is also a colossal task to compile databases of thermoelectric parameters for various synthesized TE materials and their variations with doping (n-type or p-type) [1]. Computational methods using density functional theory (DFT) are also time-consuming and demand high computational complexity for exploring TE materials [1].

Efficient TE materials require a large ZT which in turn requires to maximize the Seebeck coefficient absolute value, minimize the thermal conductivity and possess a high electrical conductivity. Optimizing these parameters is a complicated task as they are inherently dependent and conflicting in nature [1]. Thus, optimizing AT requires a thorough understanding of these various transport properties and their interrelated material characteristics.

The Seebeck coefficient depends on this energy-dependent conductivity around a fermi window centered about the fermi energy level, which is given by the Mott expression (Eq. 1.1) [1].

$$S_B = \frac{\pi^2}{3} \left(\frac{K_B^2 \times T}{q} \right) \left[\frac{d[\ln(\sigma(E))]}{dE} \right]_{E=E_F} = \left(\frac{8\pi^2 K_B^2 T}{3qh^2} \right) m_d^* \left(\frac{\pi}{3n} \right)^{2/3} \quad (1.1)$$

where n is the carrier concentration and effective mass m_d^* of the carrier when present in the conduction band or valence band. This effective mass (m_d^*) is obtained from the function of the density of states (DOS) and is thus also known as m_{DOS}^* [1]. The underlying assumption for the final closed form expression is the presence of a parabolic band and an energy-independent scattering approximation [1]. The electrical conductivity (σ_e) can be approximated by the Drude model in terms of its carrier concentration (n) and mobility (μ) as shown in EQ. 1.3. Thus, the influence of carrier concentration impacts both the parameters contradictory as shown in Fig. 1.1.



Figure 1.1: Figure 1.1

$$\sigma_e = nq\mu = \frac{nq^2\tau}{m} \quad (1.2)$$

$$\kappa = \kappa_{ph} + \kappa_e = \left(\frac{\pi^2}{3}\right)\left(\frac{nK_B^2 T \tau}{m}\right) + L_n \times \sigma_e T \quad (1.3)$$

$$L_n \approx \left(\frac{\pi^2}{3}\right)\left(\frac{K_B}{q}\right)^2 \quad (1.4)$$

1.1 Section 1.1

1.1.1 Sub-Section 1.1.1

Content.

1.1.2 Sub-Section 1.1.1

Content.

Chapter 2

Raycasting Adventures

This is going to be more of a summary and some implementation details of my first ever experience with computer graphics and making games.

Two weeks ago, not knowing anything about computer graphics and game development, I embarked on a journey down the rabbit hole of creating one of the pioneering games of video game history, such as the OG *Doom* and *Wolfenstein 3D*.

I decided to create a raycasting engine in *C++* using the *Simple DirectMedia Layer* graphics library. And for a beginner like me, it was quite the challenge.

The whole idea of making this in a very bare-bones library and language like *SDL2* and *C++* was because I wanted to create something that would be truly cross-platform from the start. And by cross-platform I mean **cross-platform**. The game would run on all desktop operating systems (Linux, Windows, Apple) and all mobile operating systems (Android, iOS, Raspberry Pi, Chrome)! And of course the PSP. I could talk a lot about why the PSP, but I'll save that for another time.

Introduction

Now you must be asking - "Bruh, what even is a Raycaster".

The simplest way to explain it is that it's a rendering technique to create a 3D perspective from a 2D map, what we like to call as 2.5D.

Although this is about the making of a raycasting game engine, I had to

give a lot of thought on how game engines are structured in the first place.

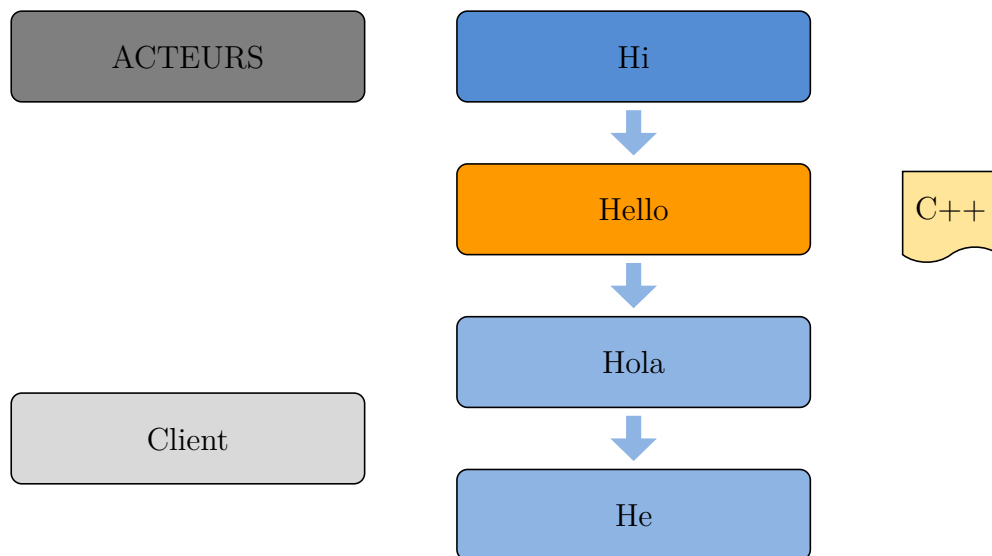
I wanted my project to follow the *Object-Oriented* design pattern as well so that it would be easy to extend and plug in components without touching a lot of code, enabling a deeper level of abstraction. I taught me a lot about inheritance and composition in C++.

The Game Engine

I'll talk about the game architecture in this section.

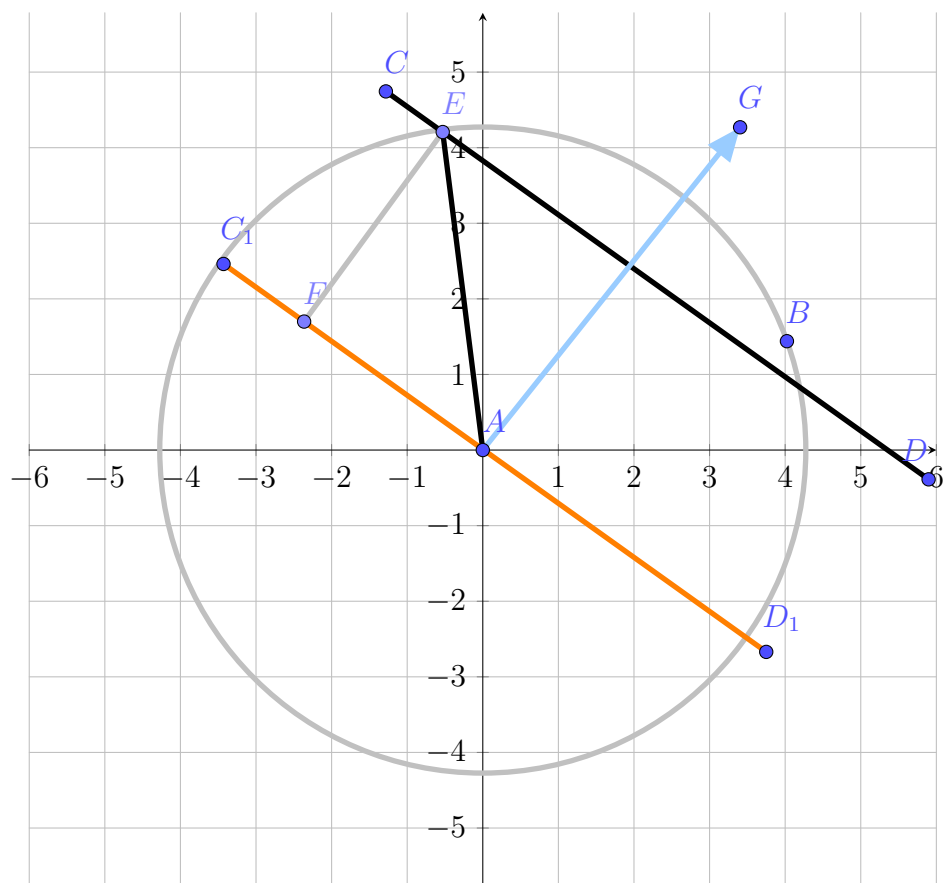
The engine was written in C++ and then compiled using the *CMake* build system, which enables me to go completely cross-platform.

The high level architecture of the engine looks something like this:



Raycasting

The biggest leap of faith I took for this engine was to create the raycasting logic. So, here's how raycasting actually works.



Chapter 3

Acknowledgement

I would like to acknowledge me for writing this and every resource I used to put this together, thank you.

References

- [1] J. T. Shreeja Das, Santanu Mahapatra and D. Saha, “Machine learning assisted search of thermoelectric materials with enhanced power factor, figure of merit, and air stability.” *Workshop on Spintronics and Magnetism on 2D Materials, EPFL*, 23rd to 27th August 2021.

This is appendix A