Michel Karam

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EDUCATION

Portland State University

Portland, OR

Bachelor of Science in Computer Science

Sept. 2023 - April 2026

- GPA: 3.89
- Key Coursework: Operating Systems, Data Structures, Algorithms & Complexity, DBase Management Systems, APL Linear Algebra

Portland Community College

Portland, OR

Associate of Science

Jan. 2022 - Mar. 2023

• Completed the Early College High School (ECHS) Program, earning college and high school credits concurrently.

SKILLS

Programming Languages: C, C++, Python, Java, JavaScript, Assembly x86/x86-64, Lua, HTML, CSS

Frameworks & DevTools: React, React-Native, Node.js, Expo, VS Code, Vim, RESTful API, Git, NASM, Modern C++, GDB, Valgrind, PostgreSQL, MongoDB

Core Competencies: Functional Programming, Data Structures, Algorithms, Space & Time Complexity, Object-Oriented Programming (OOP), AI/ML Integration, Functional & Structural Testing, Low-Level Debugging

Power Skills: Brain Storming, Attention to Detail, Organization & Time Management, Dependable & Responsible, Active listening

Projects

Prep&Count: 3-in-1 Fitness AI Application | React Native, Node.js, Express, MongoDB, JavaScript, RESTful APIs, Git

- Implemented a comprehensive fitness platform combining workout tracking, nutrition logging, and AI-driven meal planning through seamless integration with the OpenAI API.
- Architected scalable REST APIs and robust backend infrastructure to support real-time data sync, secure
 user sessions, and advanced analytics.
- Integrated intelligent meal plan generation by leveraging **AI outputs** and interfacing with a third-party **recipe API** for dynamic, personalized recommendations.
- Crafted an intuitive and responsive **UI/UX** experience with modular React Native components to drive user engagement and mobile performance.

SlimNet: Light CNN Comparison Framework | Python, TensorFlow, Keras, OpenCV, Saliency Maps

- Trained a custom CNN on the MNIST dataset to create a baseline for evaluating lightweight deep learning models in constrained environments
- Applied post training quantization (PTQ), and quantization-aware training (QAT), to reduce model size and increase bandwidth while preserving accuracy.
- Saliency Maps with Back-Propagation (BP) to interpret attention maps across both models, revealing key differences in focus, and abstraction.
- Analyzed results to draw **theoretical insights** into the relationship between layer depth, feature representation, and classification accuracy in CNN's.