# Kiet C. Lam

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## **EDUCATION**

**Hamilton College,** Clinton, NY *Bachelor of Arts in Data Science* 

August 2022 - May 2026

• GPA: 3.70

• Relevant coursework: Design Principles, Linear Algebra, Multivariable Calculus, Probability and Statistical Inference, Algorithms and Data Structures

# **SKILLS**

**Programming language:** Python, R, SQL, C++

Machine learning frameworks: Tensorflow, Keras, PyTorch

Data science tools: Pandas, Numpy, tidyverse, Scikit-learn, Matplotlib, Tableau, Seaborn, ggplot2, Microsoft

Excel, Google Sheets

Database: MySQL and BigQuery

#### **EXPERIENCE**

Research analyst, Hamilton College Neuroscience Department, Clinton, NY

May - June 2023

- Visualized participants' demographics using Python's Matplotlib
- Designed a correlational matrix among participants' responses to different questions from multiple psychological scales (MAIA, PVD, QCAE, TDD, and UCLA) using R's tidyverse ggplot2
- Customized an R script imported from college collaborator's Github repository to compile HRV data from 300+ participants
- Encoded participant's ambiguous responses into appropriate values using R's dplyr package
- Constructed an R script that compiles all responses from 4 different Qualtrics questionnaires
- Implemented additional R packages (lmerTest, lme4, and arsenal) to build multiple regression models, produce statistical summaries, as well as perform significant testings

## **PROJECTS**

# **Income prediction based on demographics**

January 2024 - March 2024

- Created a Logistic Regression model that predicts whether or not a person makes an income of over 50k
- Optimized the model using cross-validation, scaling, and regularization (L1 and L2)
- Visualized the data set's demographics (age, gender, occupation, etc.) using pie charts and bar charts
- Created a pipeline to preprocess data using Scikit-learn

# Is your banana ripe?

February 2024 - March 2024

- Utilized TensorFlow's Keras API to build a 2D multi-layered Convolutional Neural Network
- Visualized the model's training and validation loss and accuracy
- Integrated an Early Stopping Callback mechanism to prevent overfitting and enhancing model's accuracy from 80% to 95%
- Applied a 5-fold cross-validation technique, yielding an accuracy of 95.37% and a loss of 2%
- Tuned hyperparameters (L2 regularization rate, learning rate, and dropout rate, epochs) using GridSearchCV
- Evaluated the model's final performance using key metrics: precision (94.4%), recall (100%), and accuracy (96.68%)
- Leveraged OpenCV to download and preprocess over 3,000 images sourced from Google Images