# Ming (Jerry) Gao

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#### **EDUCATION**

New York University, Center for Data Science

New York City, NY

Master of Science in Data Science

Expected Graduation 05/2024

Relevant Coursework: Machine Learning, Deep Learning, Natural Language Processing with Representation Learning

Yunnan University, School of Mathematics and Statistics

Kunming, China

Bachelor of Engineering in Data Science and Big Data Technology | GPA: 3.83/4

09/2018-06/2022

Relevant Coursework: SQL, Data Mining, Data Visualization, Big Data Exploratory Analysis, Big Data Preprocessing

### **SKILLS**

- Programming: Python (Scikit-Learn, Pandas, Numpy, Matplotlib, Seaborn, etc.), R (dplyr, caret, ggplot2), SQL
- Data Science: Data Analysis, Machine Learning, NLP, Data Visualization, Statistics, Web Scraping, A/B Testing
- Other tools: PowerPoint, Excel, Jupyter Notebook, Pycharm, Rstudio

## PROFESSIONAL EXPERIENCES

Henan Junyou Digital Technology Co., Ltd. Data Scientist Intern

Zhengzhou, China

02/2022-04/2022

- Compared the similarity between Jingdong and Taobao category names and matched the Jingdong category to the Taobao category with the highest similarity layer-by-layer using the Alibaba Cloud word vector API interface and cosine similarity
- Applied the SMOTE algorithm to process the imbalanced 4M+ data of Jingdong in a certain month
- Constructed the **logistic regression**, **Naive Bayes**, **AdaBoost**, and **XGBoost classifiers** to predict price reduction for products, and found the optimal model with high interpretability and high accuracy of 87.6%
- Provided the company with a **project example** as it was the company's **first project** to use machine learning for data analysis during the transformation process

#### RELEVENT PROJECTS

Bank Credit Card Customer Churn Warning Based on Multi-class Logistic Regression Model

02/2022-05/2022

- Applied the **K-prototype clustering** algorithm to cluster **10k+** credit card customers data, dividing customers of a bank into three categories: loyal, churn and potential churn
- Conducted One-Hot encoding, performed feature engineering based on the Extra-Trees model
- Applied One-vs-Rest method to construct a three-classification logistic regression model, achieving 99.67% accuracy
- Explored the factors affecting the churn status of different credit card customer groups and delivered **recommendations** on customer churning prevention and improvement of customers' loyalty

## Research on the Accurate Discriminative Model for Heart Disease Patients

06/2021-07/2021

- Preprocessed 3k+ data, applied a C4.5 decision tree model, a back propagation neural network model, and a logistic regression model to construct different discriminant models for heart disease patients using Python
- Compared the 3 models and reached 73.33% accuracy via the logistic regression model, explored the factors affecting the likelihood of heart disease, which could improve the efficiency of predicting heart disease (by 30%)

## Natural Language Processing (NLP) Text Classification

12/2020-01/2021

- Led a team of 3 to scraped 1k+ pieces Covid-19 news in Chinese using Python, and labeled the news texts manually
- Segmented the texts using jieba library, removed non-text characters and stop words
- Performed a TF-IDF model and a PCA model for text feature extraction
- Constructed AdaBoost models and XGBoost models and optimized models with fine-tuning hyperparameters
- Compared the models, found that a certain AdaBoost model was the optimal model for classifying the scraped news with accuracy of 80%

#### Predictive Analysis of Car Dealer A's Customer Churn Warning and Car Return to Factory

05/2020-06/2020

- Preprocessed 430k+ data, used Python to connect the Microsoft SQL Server to store the data, and constructed discriminant analysis, logistic regression and k-nearest neighbor models to predict customer churn
- Constructed OLS Regression, Ridge Regression, Lasso Regression and Principal Component Regression models
  to predict the car dealer A's car return situations to the factory
- Compared the models and applied the optimal ones, founding the churn rate of the car dealer A's customers as 32.81%
- Explored the specific factors affecting the customer churn and delivered **recommendations** on reducing customer churn rate