# **Machine Learning Engineer Nanodegree**

Hang Guo

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### Domain Background

The coffee retail giant company Starbucks owns a consumer mobile application, which sent out rewards and offers for customers occasionally. The business team in Starbucks want to find out the pattern of a customer that takes the offer, providing us the data sets that include customer profiles, transaction history and offer type data.

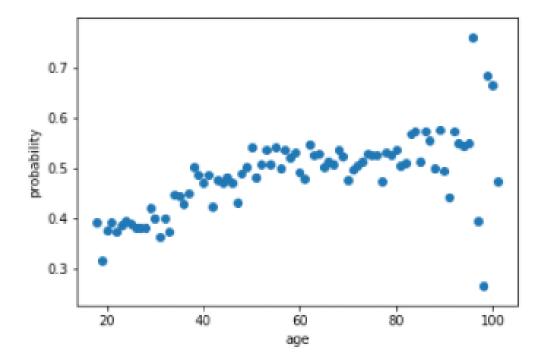
This project is designed to analyze the application data and designed and develop a machine learning model to predict the probability that a particular group of customers will take the offer. The model should test properly by using the testing data after deployed the machine learning model.

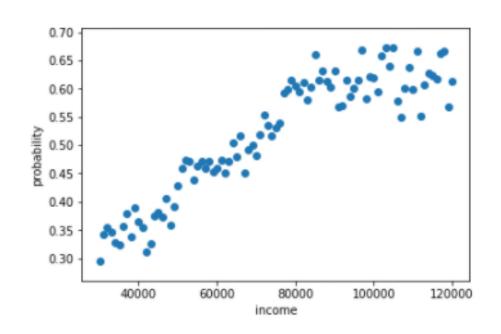
#### **Problem Statement**

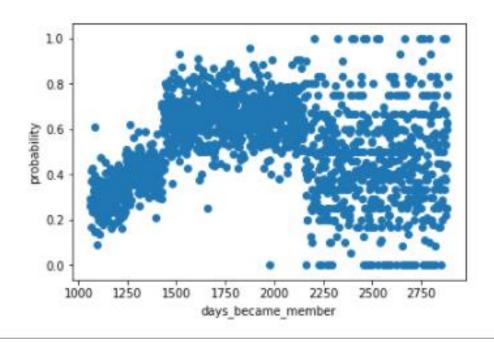
The business problem that we can tell from the data set is different group of customers responded differently to the offers. See below data:

| : |    |     |                  |        |                                  |          |             |                    |
|---|----|-----|------------------|--------|----------------------------------|----------|-------------|--------------------|
|   |    | age | became_member_on | gender | person                           | income   | probability | days_became_member |
|   | 1  | 55  | 20170715         | F      | 0610b486422d4921ae7d2bf64640c50b | 112000.0 | 0.500000    | 1438.0             |
|   | 3  | 75  | 20170509         | F      | 78afa995795e4d85b5d9ceeca43f5fef | 100000.0 | 0.750000    | 1505.0             |
|   | 5  | 68  | 20180426         | M      | e2127556f4f64592b11af22de27a7932 | 70000.0  | 0.500000    | 1153.0             |
|   | 8  | 65  | 20180209         | M      | 389bc3fa690240e798340f5a15918d5c | 53000.0  | 0.833333    | 1229.0             |
|   | 12 | 58  | 20171111         | M      | 2eeac8d8feae4a8cad5a6af0499a211d | 51000.0  | 0.333333    | 1319.0             |
|   | 13 | 61  | 20170911         | F      | aa4862eba776480b8bb9c68455b8c2e1 | 57000.0  | 0.600000    | 1380.0             |
|   | 14 | 26  | 20140213         | M      | e12aeaf2d47d42479ea1c4ac3d8286c6 | 46000.0  | 0.166667    | 2686.0             |
|   | 15 | 62  | 20160211         | F      | 31dda685af34476cad5bc968bdb01c53 | 71000.0  | 0.666667    | 1958.0             |
|   |    |     |                  |        |                                  |          |             |                    |

The field "probability" describes the probability that a customer is likely to take the offer, and it is calculated by joining profile table with the transaction table, divide the count of offer complete and count of offer received data. By further look into the correlation between the individual column and probability:







|        | probability |
|--------|-------------|
| gender |             |
| F      | 0.563502    |
| М      | 0.431158    |
| 0      | 0.549057    |

From the chart, fields like gender, income and age might be a good candidate to forecast the probability.

## **Datasets and Inputs**

The data is contained in three files:

- portfolio.json containing offer ids and meta data about each offer (duration, type, etc.)
- profile.json demographic data for each customer
- transcript.json records for transactions, offers received, offers viewed, and offers completed

Here is the schema and explanation of each variable in the files:

# portfolio.json

- id (string) offer id
- offer\_type (string) type of offer ie BOGO, discount, informational

- difficulty (int) minimum required spend to complete an offer
- reward (int) reward given for completing an offer
- duration (int) time for offer to be open, in days
- channels (list of strings)

### profile.json

- age (int) age of the customer
- became\_member\_on (int) date when customer created an app account
- gender (str) gender of the customer (note some entries contain 'O' for other rather than M or F)
- id (str) customer id
- income (float) customer's income

### transcript.json

- event (str) record description (ie transaction, offer received, offer viewed, etc.)
- person (str) customer id
- time (int) time in hours since start of test. The data begins at time t=0
- value (dict of strings) either an offer id or transaction amount depending on the record

#### **Solution Statement**

For this business problem, the solution would be design and train a linear regression model by using AWS Sagemaker. The model will take the input as customer's income, age, and gender and make prediction on the probability that the customer may accept the Starbuck's offer.

#### **Benchmark Model**

For this problem, the Benchmark model would also use the linear regression but without taking a hyper parameter tunning process. The expected winner of them should be the solution model.

#### **Evaluation Metrics**

The accuracy of the machine learning model will be used to evaluate the performance. The target accuracy matrix should be able to compare the prediction result with the labeled test data set. The same matrix will apply to the benchmark model to do the performance measuring.

### **Project Design**

The project workflow:

• Data cleansing by removing null or empty value for all data sets.

- Transform the data into a Pandas data frame with the fields gender, age, income and calculate the offer completion probability.
- Split the data into training, validation, and testing data set, and then output these data to csv files.
- Upload the csv files to AWS S3.
- Create hyperparameter tunning process.
- Call Sagemaker estimator object to train the model.
- Call transform method to predict the test data, display the scatter plot for prediction performance.
- Deploy the model to an end point and clean up the resources.