# Data Science Institute of Australia

Capstone Project - Maximizing Customer Retention at Instacart

Start Date: 08/06/2019

| Epic             | Predicting Customer Repeat Purchases  |
|------------------|---|
| Project owner    | Mai Anh Ly  |
| Project status   | Development   |
| Developers       | Mai Anh Ly  |
| Client           | Ravi Gupta (Instacart CFO & COO) David Hahn (Instacart Chief Product Officer)                                       |
| Stakeholders     | Instacart UX lead Instacart web/mobile development lead Data warehousing/engineering lead End users Retail partners |
| Design complete  | 23/06/2019  |
| Code complete    | 25/07/2019  |
| Testing complete | 30/07/2019  |
| Release date     | 15/08/2019  |

## **Background**

Instacart is a grocery delivery service that connects customers to grocers via a web/mobile app. Customers cite convenience as the primary reason for preferring Instacart vs. in-person shopping. Repeat customers also tend to repurchase products from their ordering history rather than via product search. Instacart will retain more customers and increase revenue over their competitors by capitalizing on these behaviours.

We can improve the Instacart shopping experience by predicting items a customer will repurchase and implementing a 'repurchase items' page on the app. The Instacart service will be more convenient to use, and encourage customers to make shopping with Instacart part of their routine.

## **Objectives**

| Objective                                  | Metric  |
|--|---|
| Predict products customers will repurchase | Accuracy of predictions is more than 75%              |
| Improve customer experience with app/site  | Majority of customers use new 'repurchase items' page |
| Improve customer retention                 | Monthly customer retention increases                  |

## **Assumptions**

#### Instacart business model:

- Instacart takes 20% of order value as revenue
- Month-to-month customer base is at least 200,000

#### **Instacart customers:**

- An average customer makes 3 orders in a month with each order worth \$100
- Majority of customers access service from the app
- Client has addressed customer and retail partner privacy concerns

#### Technical constraints and considerations:

- Instacart stores purchase data of all customers who have made at least 1 order
- Products have been ordered by multiple users at multiple retailers
- Modelling is performed on customers who have made at least 3 orders
- Each order, customer and product in Instacart's database has a unique ID number
- Coding, testing and deployment can be done in the cloud

# **User Stories and Requirements**

| Ref<br># | Name  | User Story   | Functional Requirements   |
|----------|---|--|---|
| 1        | Candidate products on single page                         | As a customer, I want to easily repurchase products without wasting my time navigating through multiple pages and choices  | <ul><li>Product carousel</li><li>Option to remove products</li><li>'Add all to cart' button</li></ul>   |
| 2        | Record number of correct predictions                      | As a retail partner, I want to see which of my products are popular with customers so I know where to maximise profit  | <ul> <li>Record result of each prediction in database</li> <li>Ability to aggregate value of correct predictions by product and retailer</li> </ul> |
| 3        | Record end<br>user clicks and<br>time spent on<br>feature | As a UX designer, I want to know how customers will respond to the 'repurchase item' feature so I can judge its effectiveness  | <ul> <li>Elements associated with feature logs clickstream data</li> <li>Site/app logs time spent on feature page</li> </ul>                        |
| 4        | Compatible with existing back-end infrastructure          | As a data engineer, I need to know the volume of predictions the model will make per user so I can manage the infrastructure between the server and the front-end app/site | - All code made available in company repository   |
| 5        | Compatible with existing front-end infrastructure         | As a web/mobile developer, I need to know what content is supplied by the new feature so I can code the app/site layout appropriately                                      |   |

## **Open Questions**

| Question   | Answer  | Date Answered |
|--|---|---------------|
| How does Instacart differ from online groceries?                               | Instacart operates a web/mobile app where customers order from Instacart retail partners, which are then fulfilled and delivered by Instacart personal shoppers. Retail partners handle all inventory management. | 25/06/2019    |
| How was retailer and user privacy and confidentiality dealt with?              | No retailer data was provided, user ID numbers have been anonymised and no personal information was released  | 23/06/2019    |
| How do we determine if a predictive model has an effect on customer retention? | Conduct an A/B test where model has been implemented in one case and measure customer retention/churn over time   | 23/06/2019    |

# **Out of Scope**

- UX design of "repurchase items" feature
- Creation of server infrastructure linking model to Instacart webpage/app
- Creation of dashboard for monitoring model data stream
- A/B testing of "repurchase items" feature

### **Risks**

| Risks  | Likelihood | Impact | Migation  |
|--|------------|--------|---|
| Memory limitations due to processing a very large production dataset | High       | High   | Test code using free Google<br>Compute Engine runtime via<br>Colab notebook<br>Use low-bit datatypes to<br>conserve memory  |
| Customer will churn if feature orders products they do not want      | Medium     | Medium | Refine model to have precision of 75% or higher Limit predictions to high probability of repurchase Include tutorial elements on first use of feature Add ability to remove items |

## User interaction and design

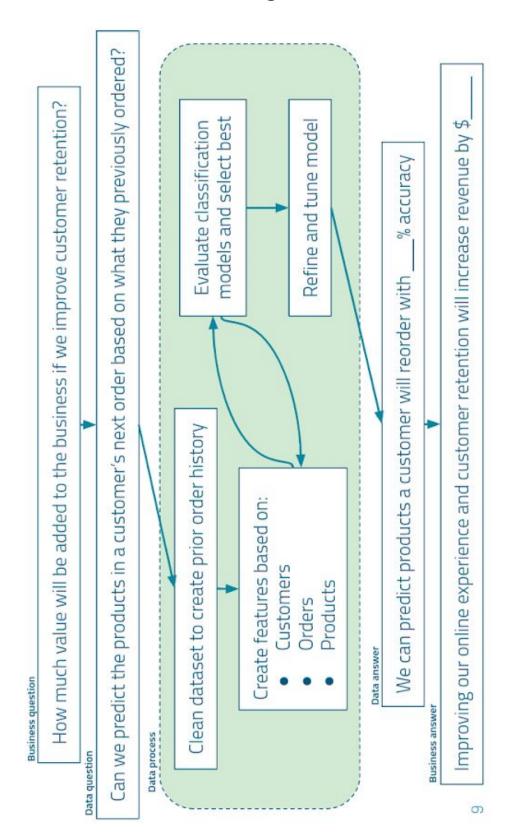


Figure 1. Project flowchart

# Tasks

| Task   | Name       | Due date   | % Time<br>allocated |
|--|------------|------------|---------------------|
| Acquire data   | Mai Anh Ly | 25/06/2019 | 5                   |
| Clean data   | Mai Anh Ly | 27/06/2019 | 10                  |
| Create features for predicting repurchases                               | Mai Anh Ly | 18/07/2019 | 30                  |
| Evaluate accuracy of classification models using all features/predictors | Mai Anh Ly | 21/07/2019 | 15                  |
| Tune hyperparameters of best evaluated model                             | Mai Anh Ly | 26/07/2019 | 20                  |
| Perform feature selection of tuned model                                 | Mai Anh Ly | 27/07/2019 | 10                  |
| Use tuned model to make final repurchasing predictions                   | Mai Anh Ly | 28/07/2019 | 5                   |
| Complete documentation   | Mai Anh Ly | 31/07/2019 | 5                   |

#### **Results and Conclusions**

We have built a model that predicts customer repurchases with 79% accuracy. This is a 4% improvement on the project objective for predictive accuracy. The model will also be correct in 75% of cases where it predicts an item repurchase. This will decrease the risk of the customer being shown items they do not want, or are not likely to repurchase.

We have yet to determine if the customer repurchase predictive model will improve the Instacart online experience. Subsequently, we do not yet know if the model will improve customer retention. However, we are confident the end-user model interface will add to the convenient shopping experience valued by Instacart customers. We estimate under the assumptions outlined in this document that a customer retention increase of 1% maintained over a year will deliver \$1.4 million in revenue. Therefore, a strategy that will increase customer retention is worth pursuing.

### **Key insights**

Our analysis from data cleaning and model development offers the following insights into customer preferences and behaviour:

- Customers tend to place more items they have purchased before in their cart the more recent the orders were made
- The top 10 repurchased items are all from the fresh fruits and vegetables aisle
- Six of the top 10 repurchased items contain the word "organic" in their name
- Item repurchases are strongly influenced by the number of purchases a customer has made from the same aisle as the target item

### **Next steps & recommendations**

- 1. Work with data engineers to deploy model and determine the frequency of live data input into model pipeline.
- 2. Work with UX and web/mobile development teams to develop "repurchase items" feature based on model predictions.
- 3. Conduct an A/B test and measure customer churn from month-to-month.
- 4. Based on A/B test results, make a decision on implementing feature site-wide.

#### **Supplementary recommendations**

- 1. As any changes to an app/site interface risks customer churn, we suggest holding a social media marketing campaign publicising an upcoming "repurchase items" feature. This may encourage customers to feel positively inclined towards the feature.
- 2. Our analysis of customer orders revealed that organic produce are the most popular repurchases among the customer base. We suggest working with retail partners to capitalize on this preference.