Dear [Client point-of-contact],

Thank you for providing us with the three datasets from Sprocket Central Pty Ltd. The below table highlights the summary statistics from the three datasets received. Please let us know if the figures are not aligned with your understanding.

Table Name	No. of records	Distinct Customer IDs	Date of Data Received
Customer Demographic	4,000	4,000	***
Transactions	20,000	3,495	***
Customer Address	3,999	3,999	***

Notable data quality issues that were encountered and the methods used to mitigate the identified data inconsistencies are as follows. Furthermore, recommendations have been provided to avoid the reoccurrence of data quality issues and improve the accuracy of the underlying data used to drive business decisions.

Additional Customer_ids in the "Transactions table" and "Customer Address table" but not in "Customer Demographic", which is the master table

Mitigation: Please ensure that all tables are from the same period. Only customers in the Customer Demographic table will be used as a training set for our model.

This indicates that the data received may not be in sync with each other which may skew the analysis results if there are missing data records.

Various columns, such as Brand, or Job Title, have empty values in certain records

Mitigation: If only a small number of rows are empty, filter out the record entirely from the training set for prediction. Else, if it is a core field, impute based on distribution in the training dataset. For key datasets, such as Transactions, less than 1% of transactions (totalling less than 0.1% of profit) have missing fields. These records have been removed from the training dataset.

Inconsistent values for the same attribute (e.g. New South Wales being represented as "NSW" and "New South Wales")

Mitigation: Use regular expression to replaced extended values into abbreviations to ensure consistency across addresses.

Recommendation: Enforce a drop-down list for the user entering the data rather than a free text field.

In order to construct meaningful variables for the model, the data has been cleaned to avoid multiple representations of the same values. Additionally, gender records where "U" have been replaced, and "Femal" has been corrected as "Female" to represent gender.

 Inconsistent data type for the same attribute (e.g. DOB involved in Customer Demographic table is numeric but the format of it is text (non-numeric) in NewCustomer table)

Mitigation: Convert selected records in characters to numeric. Remove non-numeric characters from string.

Recommendation: Ensure that fact tables in the given database have constraints on data types. Having different data types for a given field make it tricky to interpret results at the later stage. Therefore, appropriate data transformations are made to ensure consistent data types for a given field.

Moving forward, the team will continue with the data cleaning, standardisation and transformation process for the purpose of model analysis. Questions will be raised along the way and assumptions documented. After we have completed this, it would be great to spend some time with your data SME to ensure that all assumptions are aligned with Sprocket Central's understanding.

Sincerely yours, Heidi Qin