

KPMG Data Quality Assessment

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Introduction

This data quality assessment report aims to provide a comprehensive overview of our client's (Sprocket Central Pty Ltd) data quality landscape, addressing critical aspects such as Data Sources, Data Quality Framework, Data Quality Assessment Methodology, Data Quality Dimensions, and Findings & Recommendations. As we delve into each section, our objective is to assess the state of our data quality, identify potential areas of improvement, and offer actionable recommendations to enhance the overall reliability and utility of our data assets.

Data Sources

Our data quality assessment is centered around three primary datasets provided by the client:

- Customer Demographic Data
- Customer Addresses
- Transactions Data

Data Quality Framework

The Data Quality Framework is designed to uphold data quality across key dimensions:

Accuracy: We prioritize data precision through validation and cleansing.

Completeness: We ensure all relevant data is captured, addressing missing information.

Consistency: Standardized data entry and validation rules maintain uniformity.

Currency: Regular updates keep data current for real-time decisions.

Relevancy: Data sources are reviewed to maintain relevance to our operations.

Validity: Checks and validations enforce data conformity to defined criteria.

Uniqueness: Duplicate records are detected and resolved to preserve data integrity.

Tools and Software's

The data quality assessment is conducted using industry-standard tools and software, primarily Excel and Power BI. These widely recognized and accessible tools enable us to efficiently manage, analyze, and visualize large datasets, ensuring a thorough examination of data quality.

Findings & Recommendations

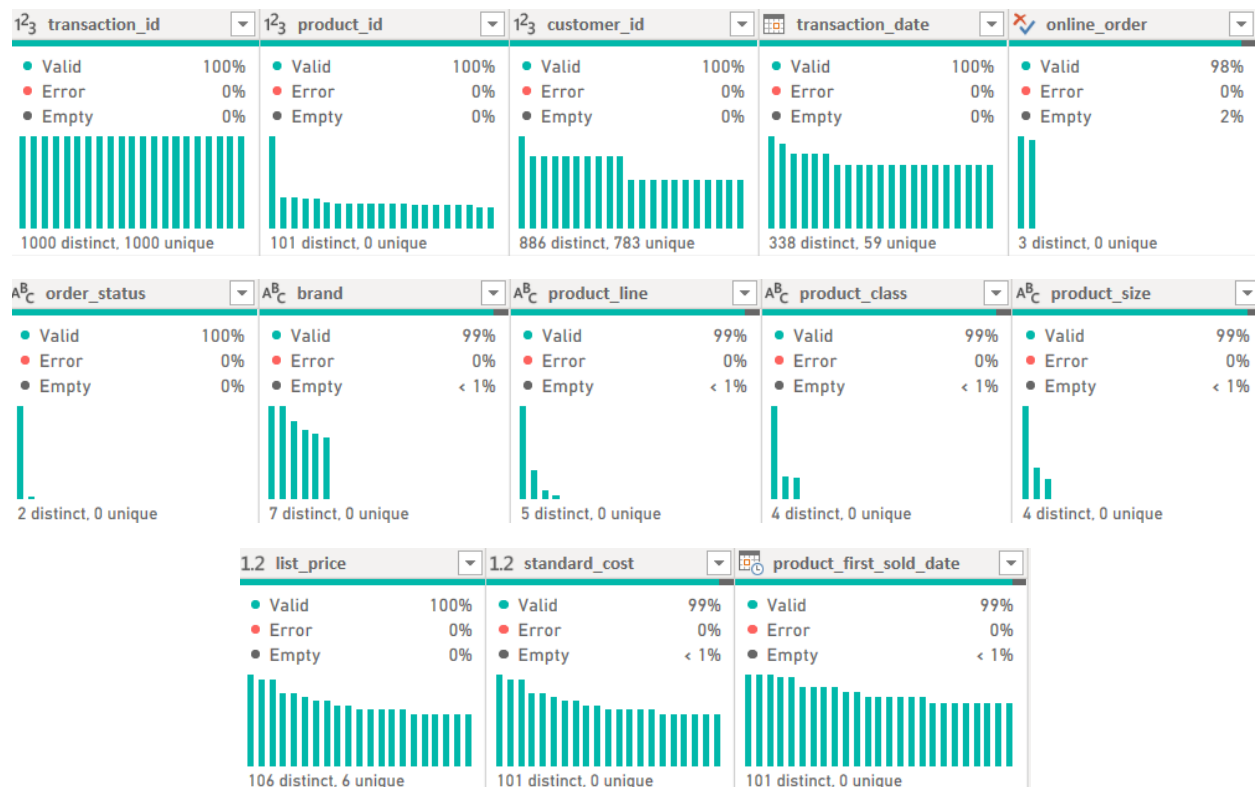
Transactions dataset:

- Dataset name confirmed

- After exploring the data type, it appears that (product_first_sold_date) column need to converted from integer to data-time

1 ² ₃ product_first_sold_date	product_first_sold_date
41245	12/2/2012 12:00:00 AM
41701	3/3/2014 12:00:00 AM
36361	7/20/1999 12:00:00 AM
36145	12/16/1998 12:00:00 AM
42226	8/10/2015 12:00:00 AM

- After exploring the columns, it appears that seven columns contain missing values, and their handling should align with the nature of the analysis, either through removal or appropriate treatment.



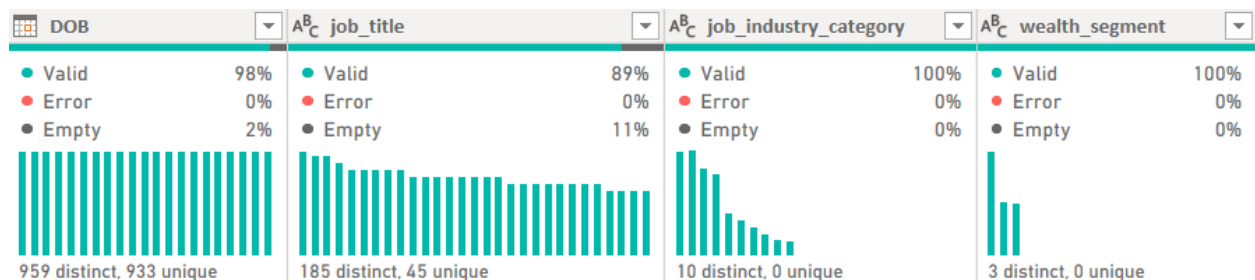
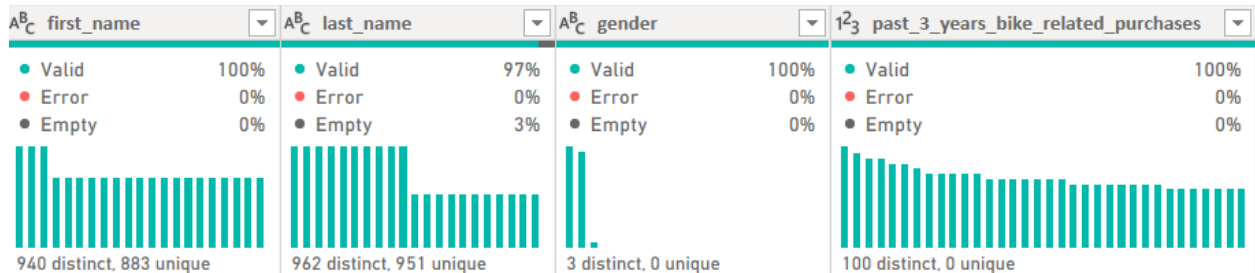
- There are no duplicate values.

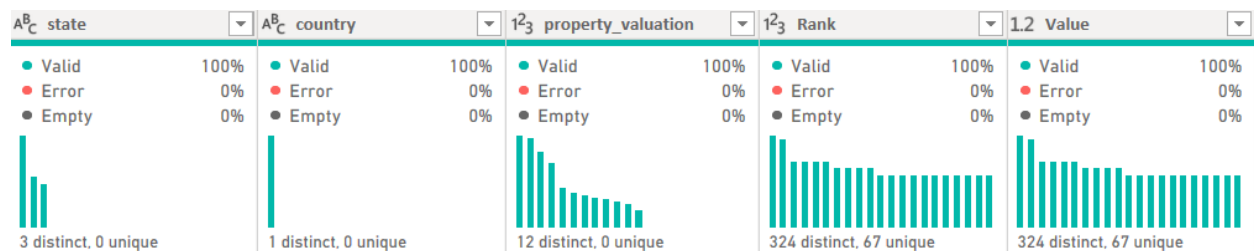
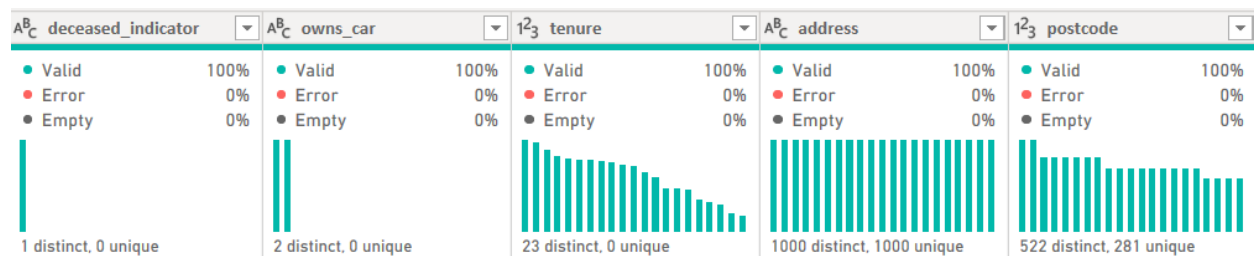
New Customer List dataset:

- Dataset name confirmed
- Column data type confirmed
- After exploring the column, it appears that columns 17 to 21 are unnamed. These columns will be removed

1.2 Column17	1.2 Column18	1.2 Column19	1.2 Column20	ABC 123 Column21
0.53	0.6625	0.828125	0.70390625	Error
0.94	0.94	1.175	0.99875	Error
0.86	0.86	0.86	0.86	Error
1.04	1.3	1.3	1.3	Error
0.92	0.92	1.15	1.15	Error
0.86	1.075	1.075	1.075	Error
1.02	1.02	1.02	0.867	Error
0.84	1.05	1.05	0.8925	Error

- After exploring the columns, it appears that four columns contain missing values, and their handling should align with the nature of the analysis, either through removal or appropriate treatment.





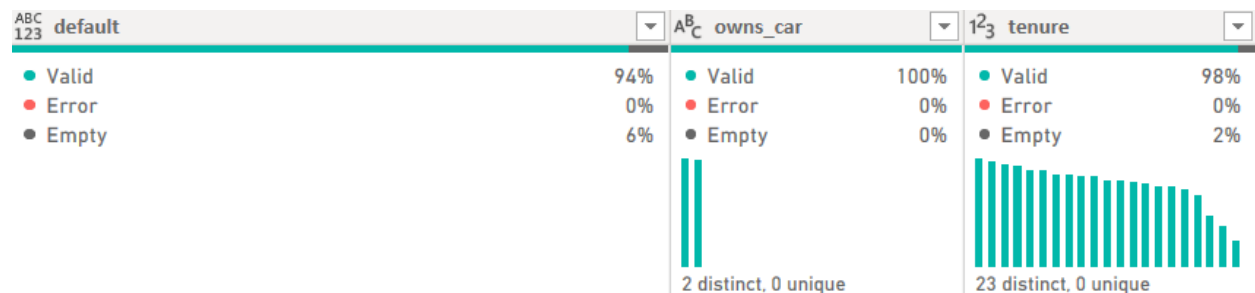
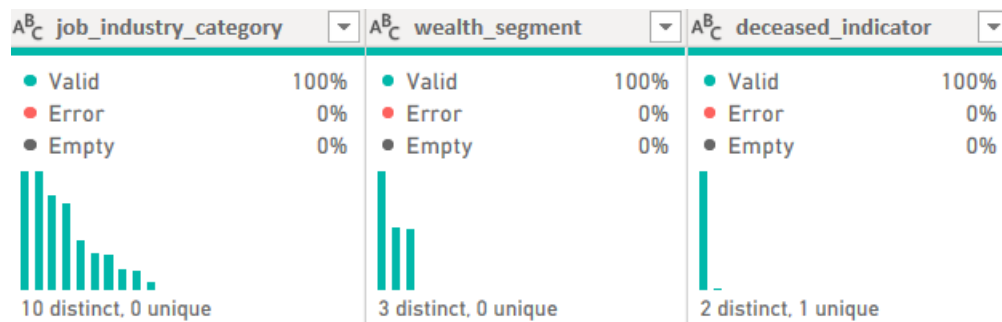
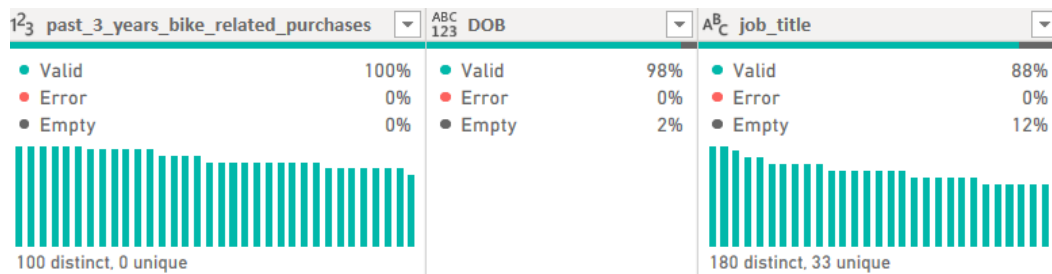
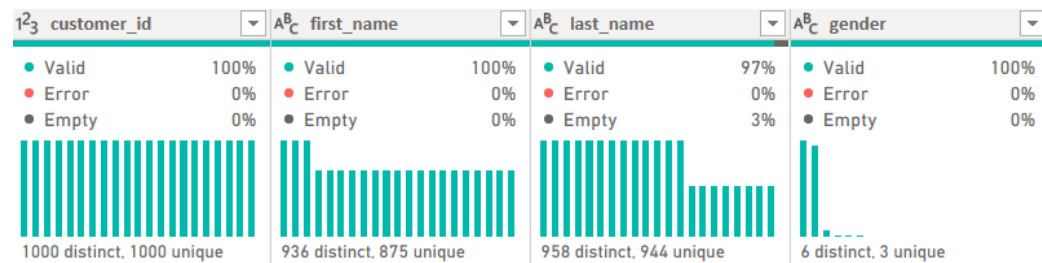
- There are no duplicate values.

- After exploring the Gender column, it appears that there are 17 records with unknown/unspecified gender (U).

A^B_C first_name	A^B_C last_name	A^B_C gender	I^2_3 past_3_years_bike_related_purchases
<ul style="list-style-type: none"> Valid 100% Error 0% Empty 0% <p>17 distinct, 17 unique</p>	<ul style="list-style-type: none"> Valid 100% Error 0% Empty 0% <p>17 distinct, 17 unique</p>	<ul style="list-style-type: none"> Valid 100% Error 0% Empty 0% <p>1 distinct, 0 unique</p>	<ul style="list-style-type: none"> Valid 100% Error 0% Empty 0% <p>16 distinct, 15 unique</p>
Normy	Goodinge	U	5
Hatti	Carletti	U	35
Rozamond	Turtle	U	69
Tamas	Swatman	U	65
Tracy	Andrejevic	U	71
Agneta	McAmish	U	66
Gregg	Aimeric	U	52
Johna	Bunker	U	93
Harlene	Nono	U	69
Gerianne	Kaysor	U	15

Customer Demographic Dataset:

- Dataset name confirmed
- Column data type confirmed
- After exploring the columns, it appears that five columns contain missing values, and their handling should align with the nature of the analysis, either through removal or appropriate treatment.



- After exploring the (default) columns, it appears that the data is inconsistent and the column will be removed

ABC 123	default		AB
	Valid	94%	
	Error	0%	
	Empty	6%	
			2
""		Ye	
<script>alert('hi')</script>		Ye	
	2/1/2018	Ye	
() { _; } >_[\$(\$())] { touch /tmp/blns.shellshock2.fail; }		Nc	
NIL		Ye	
øµ ø ø ø		Ye	
â°â´âµâââ		Ye	
(â´Â°â¡Â°î¼â´î,µ â»ââ»)		Nc	
0/0		Ye	
ø©ø½		Ye	
ÃãÃ´Ã°ã ÃÃ¨ÃÃÃ		Nc	
nil		Nc	
	-100	Ye	
â°â´âµ		Nc	
ø		Nc	

- There are no duplicate values.

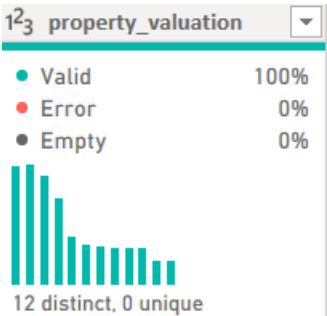
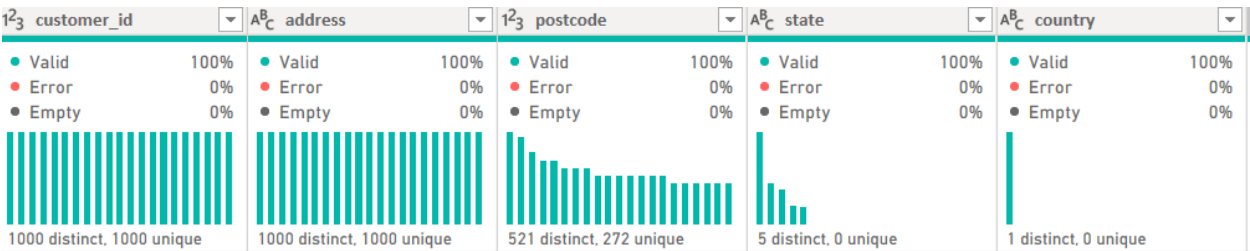
- The data in the (Gender) column is not consistent and will be standardized

- ✓ F
- ✓ Femal
- ✓ Female
- ✓ M
- ✓ Male
- ✓ U

- ✓ Female
- ✓ Male
- ✓ U

Customer Address dataset:

- Dataset name confirmed
- Column data type confirmed
- No null values



- There are no duplicate values.
- All the columns appear to have consistent and correct information

Conclusion

In conclusion, the data quality assessment has provided a comprehensive view of our data's current state. It has revealed both strengths and areas for improvement across various dimensions, including accuracy, completeness, consistency, currency, relevancy, validity, and uniqueness. Addressing these findings will be crucial to enhancing data reliability and integrity.