TheAnalyticsTeam

# Sprocket Central Pty Ltd

Data analytics approach

[Division Name] - [Engagement Manager], [Senior Consultant], [Junior Consultant]

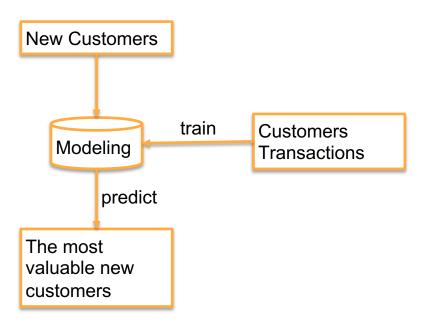
### Agenda

- 1. Introduction
- 2. Data Exploration
- 3. Model Development

#### Introduction

# Goal:

Recommend the most valuable customers from the 1000 new customers.



#### Introduction

#### Two Tables:

- Transactions
- Customers

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 19445 entries, 0 to 19999
Data columns (total 14 columns):
    Column
                              Non-Null Count
                                              Dtype
    transaction id
                              19445 non-null int64
    product id
                              19445 non-null
                                              int64
                              19445 non-null int64
    customer id
    transaction date
                              19445 non-null datetime64[ns]
    online order
                              19445 non-null
                                              float64
    order status
                              19445 non-null object
                              19445 non-null object
    brand
    product line
                              19445 non-null
                                              object
    product class
                              19445 non-null object
    product size
                              19445 non-null object
    list price
                              19445 non-null float64
 10
                             19445 non-null float64
 11
    standard cost
    product first sold date 19445 non-null datetime64[ns]
 13
    margins
                              19445 non-null float64
dtypes: datetime64[ns](2), float64(4), int64(3), object(5)
memory usage: 2.2+ MB
```

### Introduction

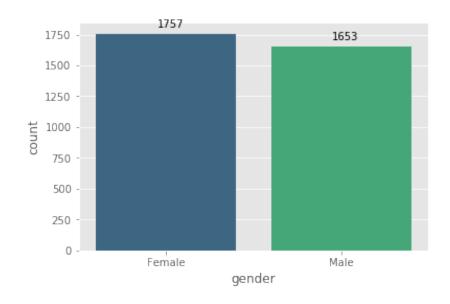
#### **Two Tables:**

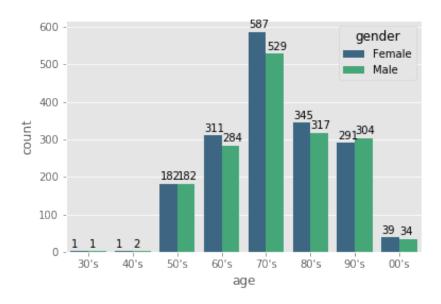
Transactions

#### Customers

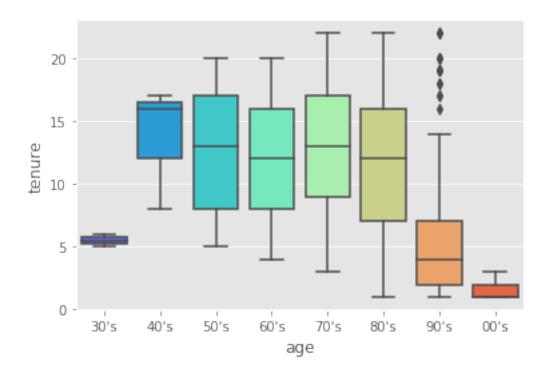
#	Column	Non-Null Count	Dtype
0	customer_id	3908 non-null	int64
1	gender	3908 non-null	object
2	past_3_years_bike_related_purchases	3908 non-null	int64
3	wealth_segment	3908 non-null	object
4	owns_car	3908 non-null	object
5	tenure	3908 non-null	int64
6	address	3908 non-null	object
7	postcode	3908 non-null	float64
8	state	3908 non-null	object
9	property_valuation	3908 non-null	float64
10	age	3908 non-null	category

### Is data skewed by gender or age?

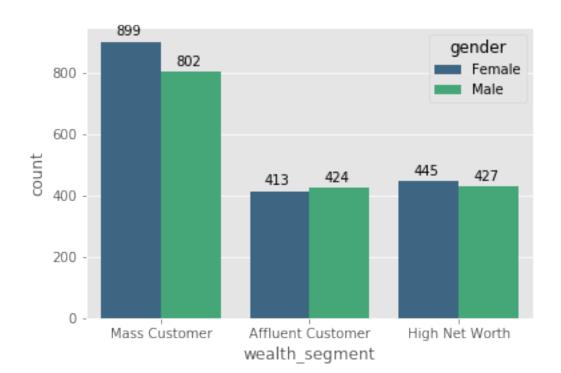




### tenure and age

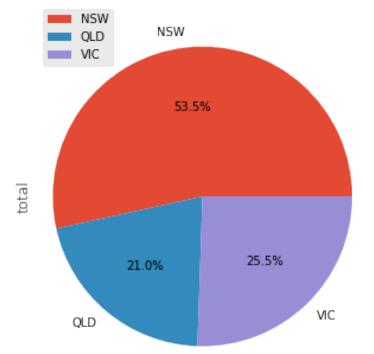


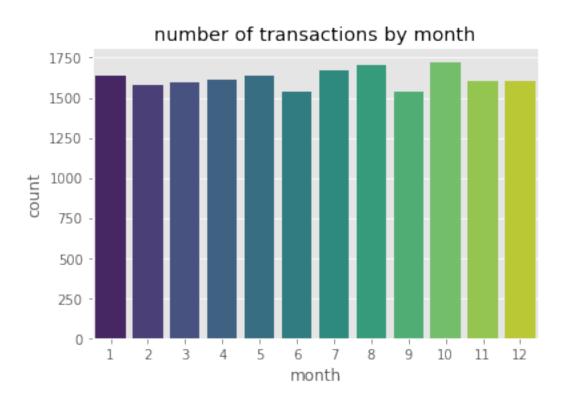
#### **Segment of wealth**



#### States where people from

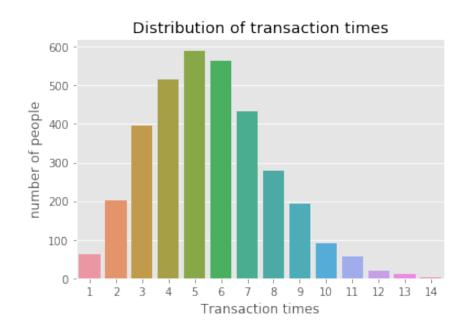
percentage of customers from different states

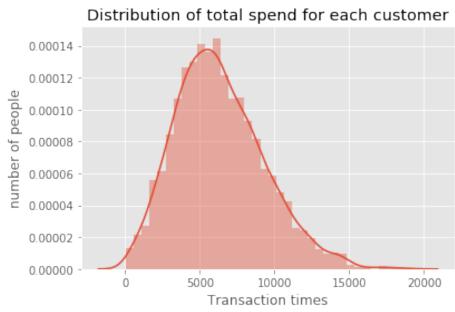




### **Model Development**

Labels valuable: transaction time >= 6 and total spend >=6000





#### **Model Development**

#### **Features:**

'gender', 'past\_3\_years\_bike\_related\_purchases', 'wealth\_segment', 'owns\_car', 'tenure', 'state', 'property\_valuation', 'age'

customer\_id
first\_name
last\_name
gender
past\_3\_years\_bike\_related\_purchases
DOB

19.6% missing
job\_title
job\_industry\_category
wealth\_segment
deceased\_indicator
owns\_car
tenure

address postcode

country

dtype: int64

property valuation

state

125

87

506

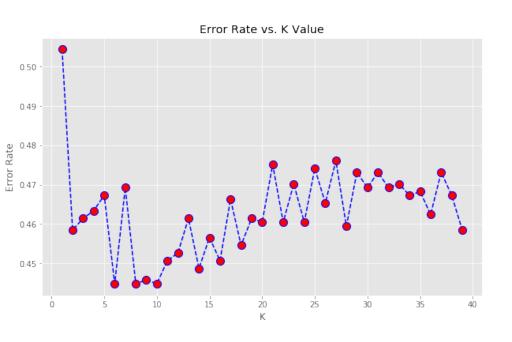
656

87

Further work:

- 1) job\_title and job\_industry\_category
- 2) Remoteness of areas

### **Model Development**



WITH K=6

[[472 113] [342 96]]

	precision	recall	f1-score	support
0	0.58	0.81	0.67	585
	0.46	0.22	0.30	438
accuracy			0.56	1023
macro avg	0.52	0.51	0.49	1023
veighted avg	0.53	0.56	0.51	1023