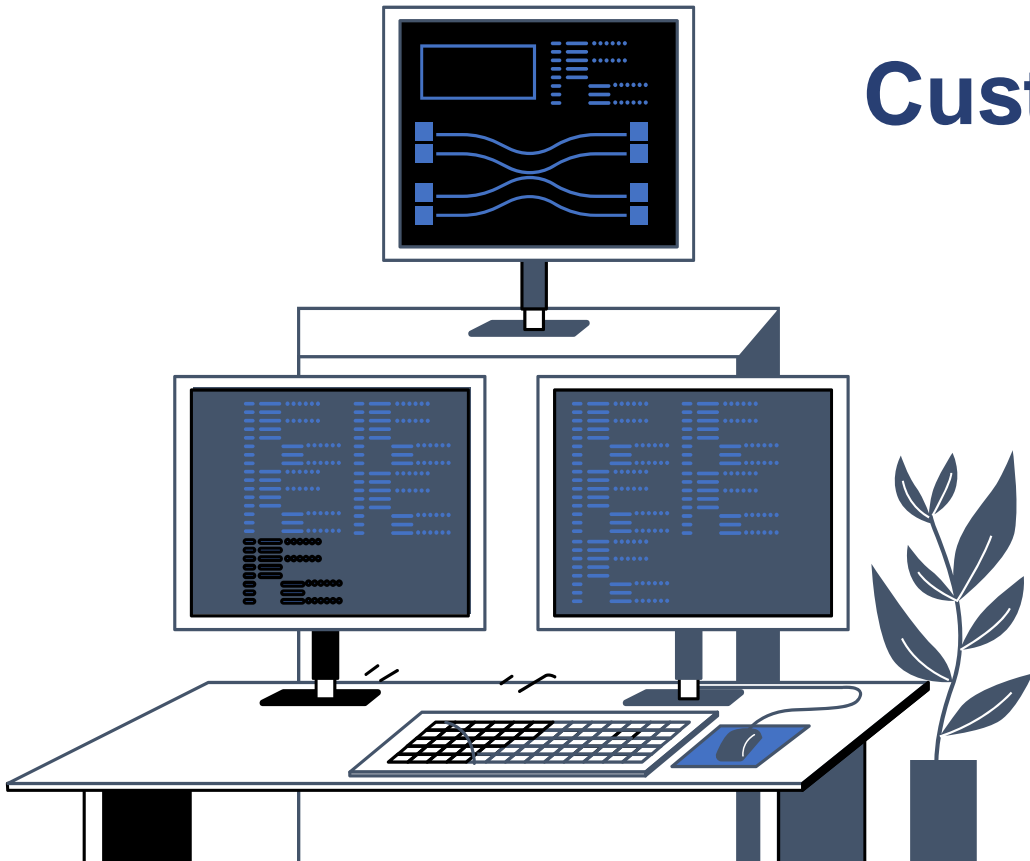


Customer Live Times Value Model



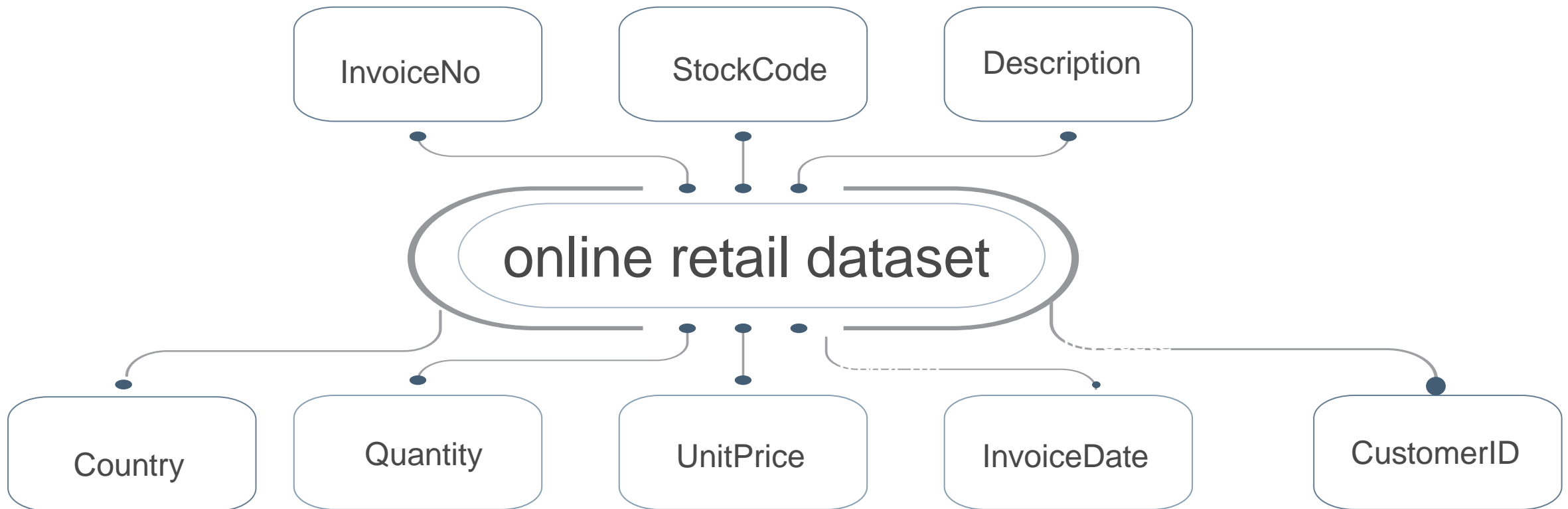
What is customer lifetime value (CLV) ?



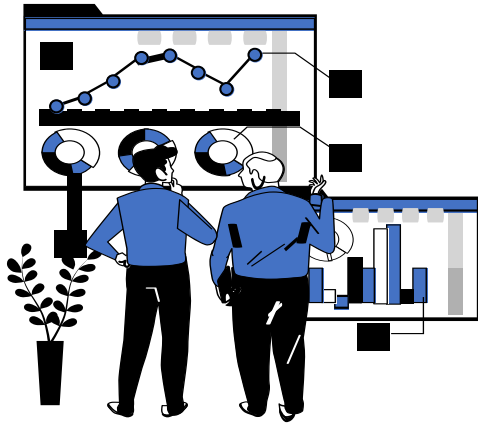
Data Description

Number of
Features: 8

Number of
Instance:
541909



The goal of this project was to use regression models to predict the CLTV model and analysis the data that answer the following questions:



Top 5 country transaction

Average Price by Country

Number of transactions for all countries

CLV for each customer

Transactions by Month and Year

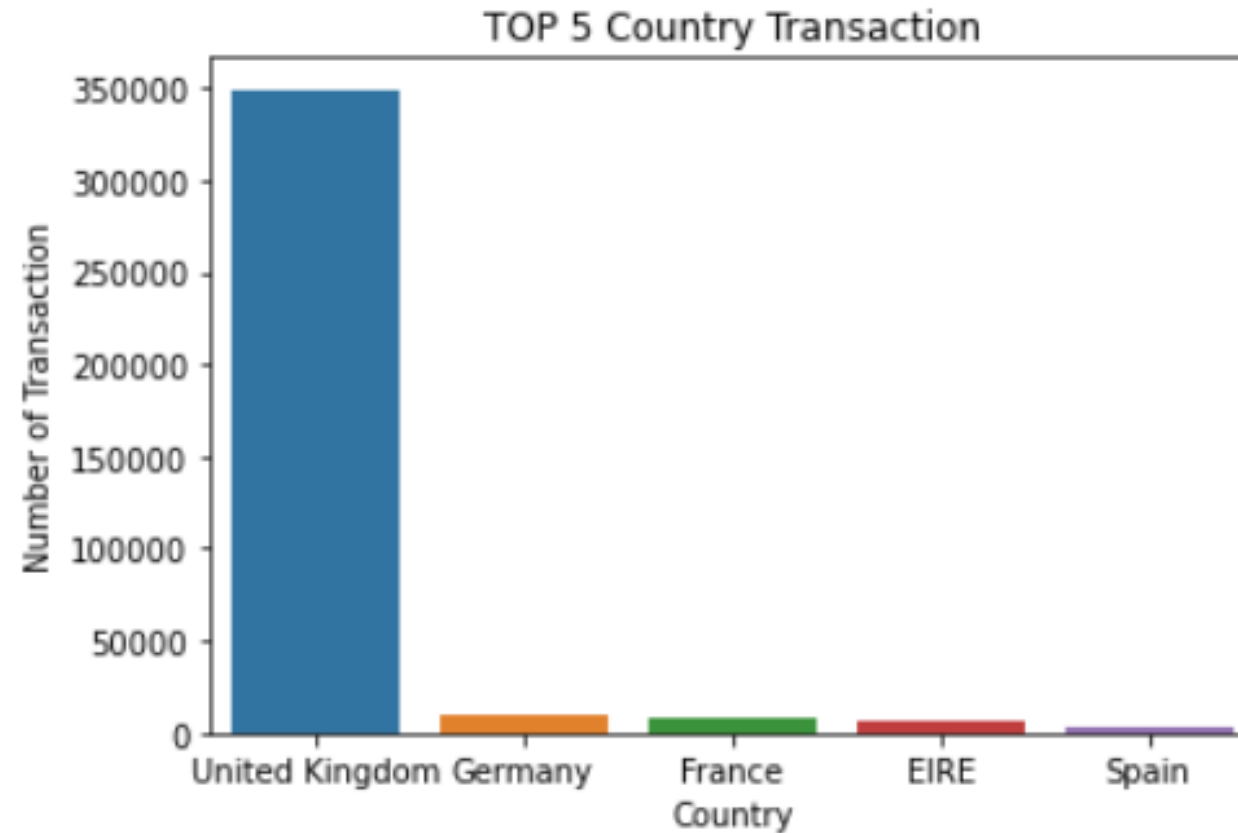
Products purchased more

Predict Purchases for 1 month for each customer

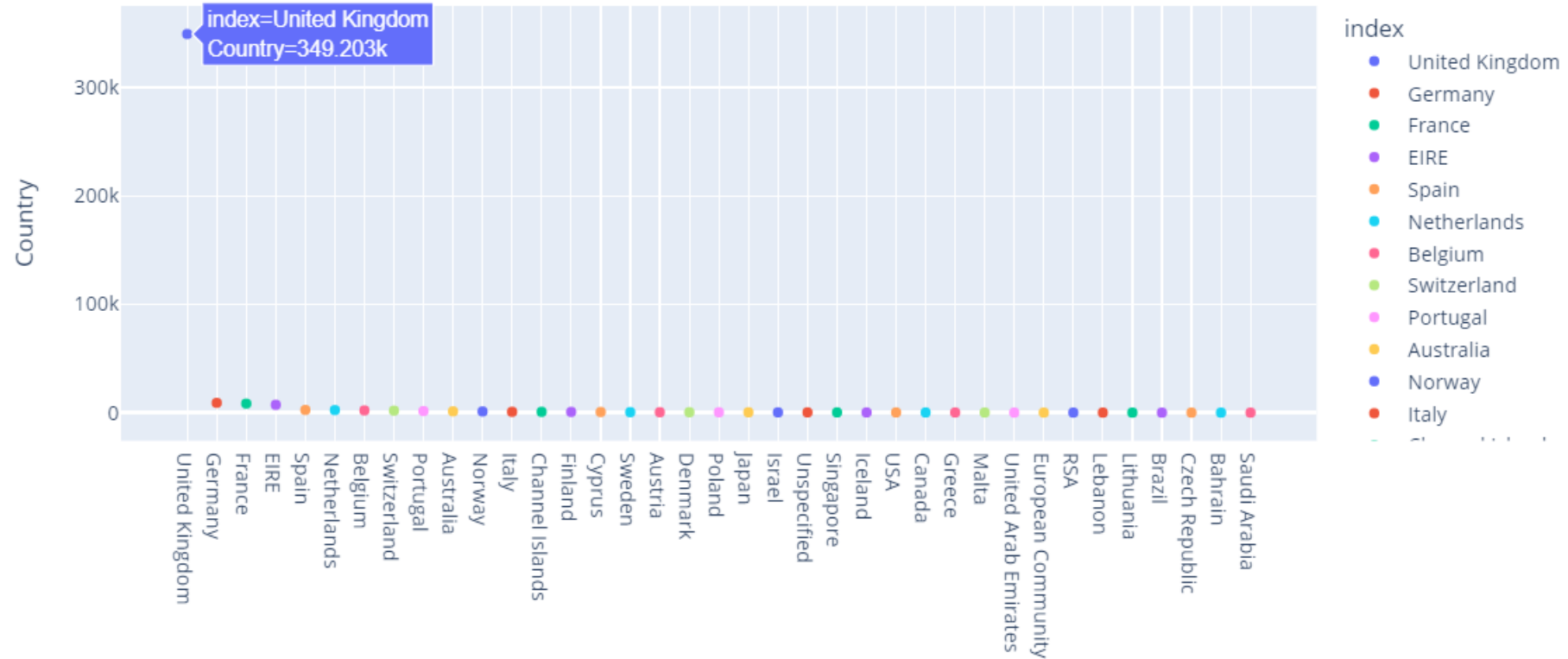
Predict CLTV using Linear Regression Model



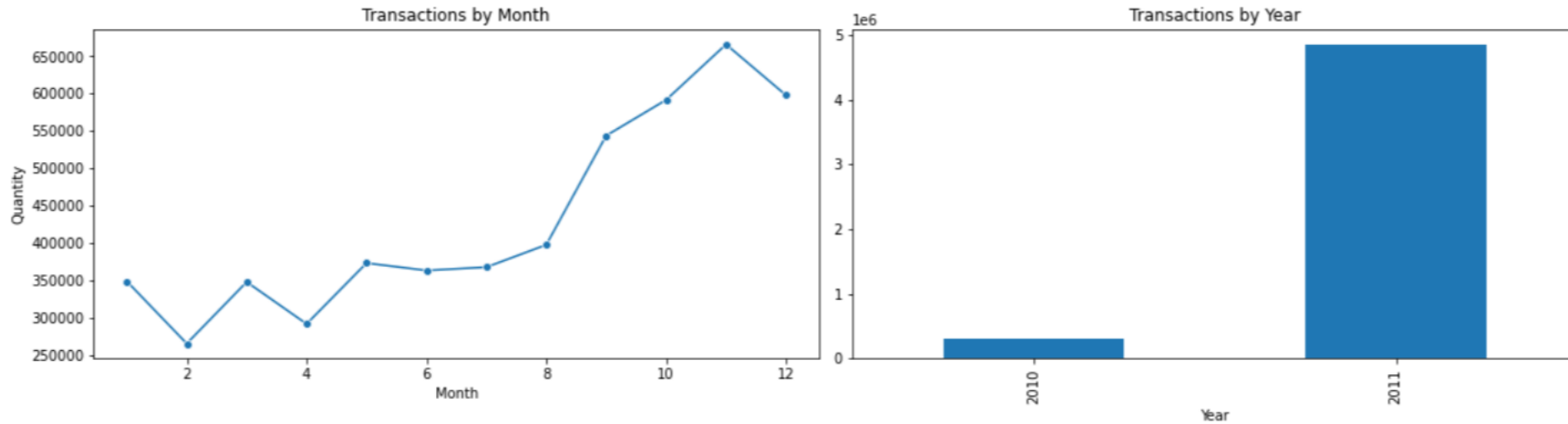
Top 5 country transaction



Number of transactions for all countries



Transactions by Month and year

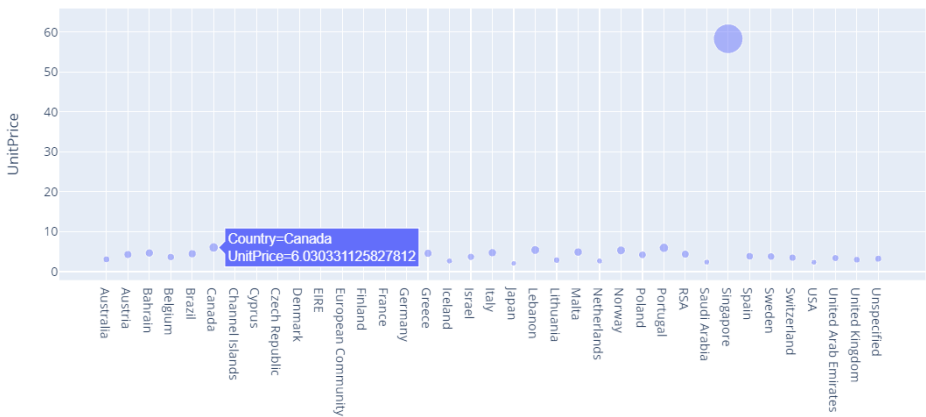
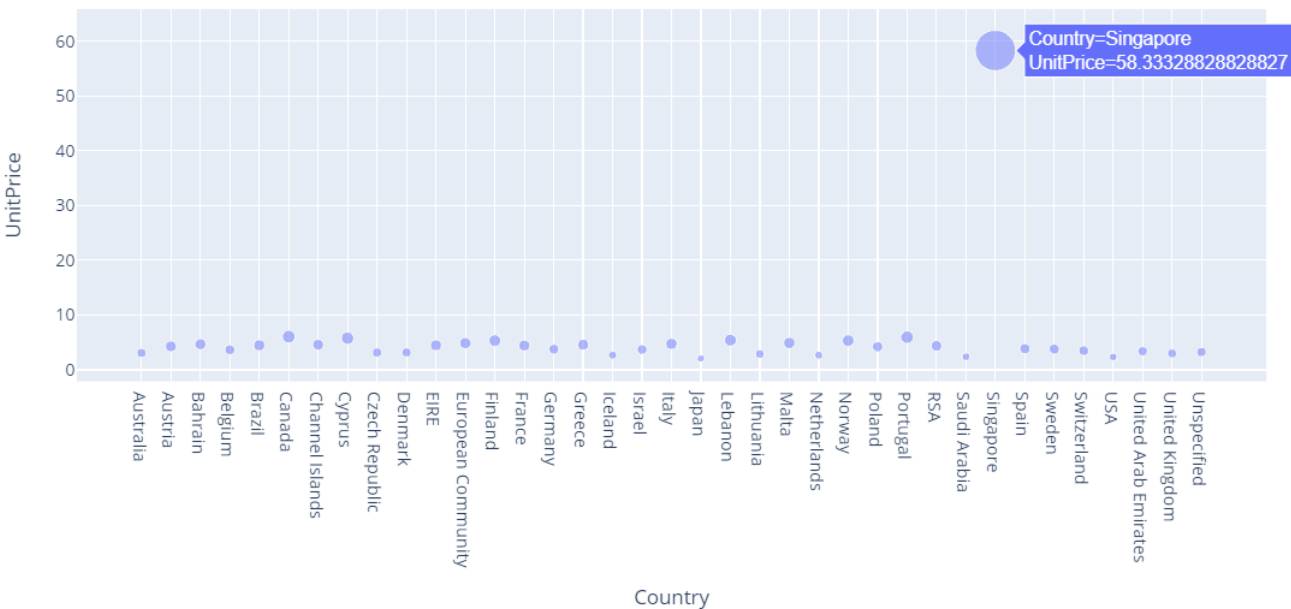




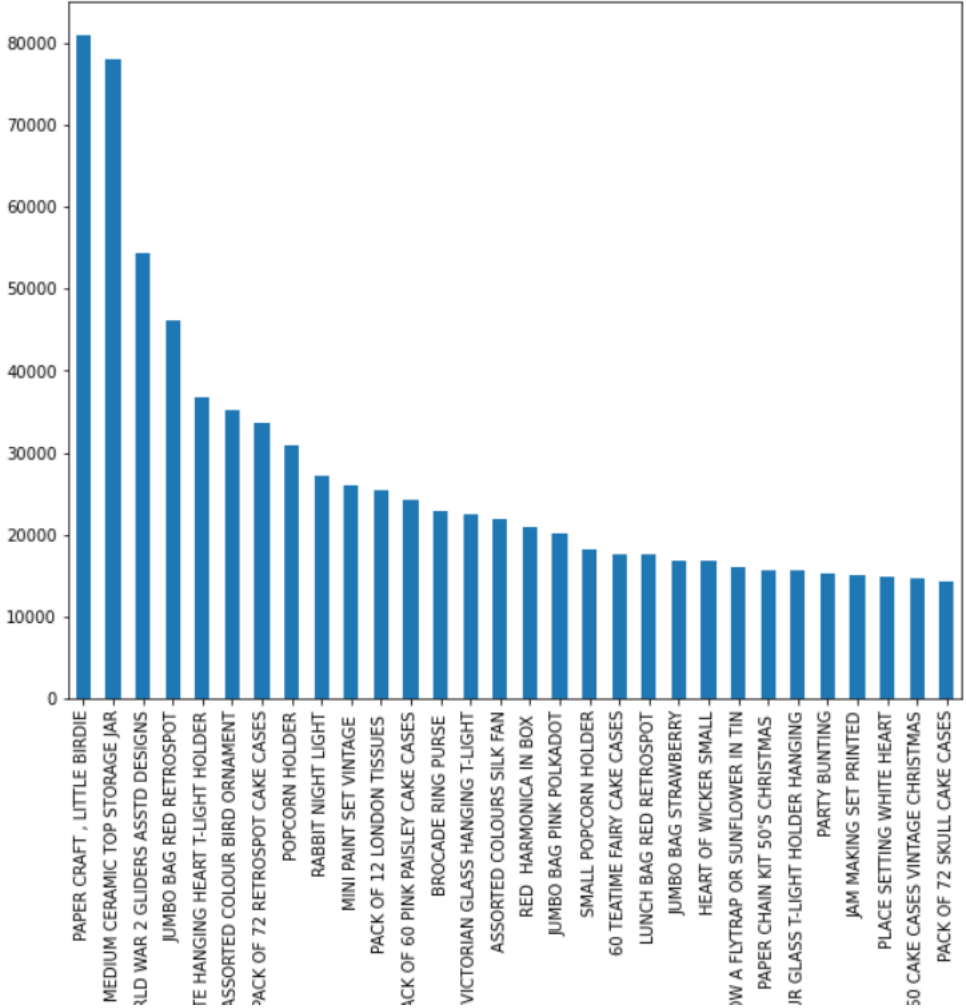
SDAIA

الهيئة السعودية للبيانات
والذكاء الاصطناعي
Saudi Data & AI Authority

Average price by country



Products purchased more



Predict Purchases for 1 month for each customer

```

result = rfm_summary.copy()

#Create instance
pareto_model = lifetimes.ParetoNBDFitter(penalizer_coef = 0.1)
pareto_model.fit(rfm_summary["frequency"], rfm_summary["recency"], rfm_summary["T"])

#Calculate the expected number of repeat purchases up to time t.

#t (array_like) - times to calculate the expectation for.
t = 30
result["predicted_purchases"] = pareto_model.conditional_expected_number_of_purchases_up_to_time(t, result["frequency"], result["T"])
  
```

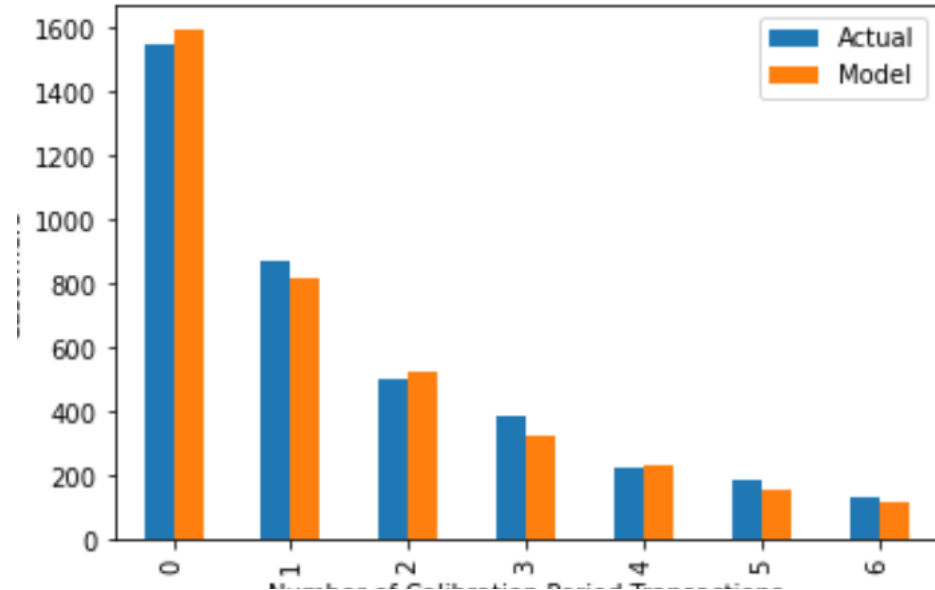
```

result["actual"] = (result["frequency"] / result["recency"]) * 30
result["actual"].fillna(0, inplace = True)
  
```

```

result["error"] = abs(result["actual"] - result["predicted_purchases"])
result.head()
  
```

CustomerID	frequency	recency	T	monetary_value	predicted_purchases	actual	error
12346.0	0.0	0.0	325.0	0.000000	0.063046	0.000000	0.063046
12347.0	6.0	365.0	367.0	599.701667	0.469673	0.493151	0.023477
12348.0	3.0	283.0	358.0	301.480000	0.268872	0.318021	0.049149
12349.0	0.0	0.0	18.0	0.000000	0.285125	0.000000	0.285125
12350.0	0.0	0.0	310.0	0.000000	0.065540	0.000000	0.065540



Customer Live Times Value

```
# Calculate Purchase Frequency
#Purchase Frequency: Purchase Frequency is the ratio of the total number of orders and the total number of customer. It represents the number of times a customer has purchased from the company.
purchase_frequency = sum(clv_grp['InvoiceNo'])/clv_grp.shape[0]

# Calculate Repeat and Churn Rate
#Repeat Rate: Repeat rate can be defined as the ratio of the number of customers with more than one order to the number of unique customers.
repeat_rate = clv_grp[clv_grp['InvoiceNo'] > 1].shape[0]/clv_grp.shape[0]
#Churn Rate: Percentage of customers who have not ordered again.
churn_rate = 1 - repeat_rate
```

```
]# Calculate Profit margin assuming gain of 5%
#Profit margin is the commonly used profitability ratio. It represents how much percentage of total sales has earned as the gain.
clv_grp['profit_margin'] = clv_grp['invoice_value'] * 0.05

# Calculate Customer Lifetime Value
clv_grp['CLV'] = ((clv_grp['AvgOdeVal'] * purchase_frequency)/churn_rate) * clv_grp['profit_margin']

CLV_F=clv_grp['CLV']
CLV_F.to_frame().tail()
#CLV_F.sort_index()
```

```
]#
```

CLV

CustomerID	
18280.0	9.019848e+05
18281.0	2.580497e+05
18282.0	7.305777e+05
18283.0	1.604871e+06
18287.0	1.333572e+07

Predict CLTV using Linear Regression Model

month_yr	CustomerID	Apr-2011	Aug-2011	Dec-2010	Dec-2011	Feb-2011	Jan-2011	Jul-2011	Jun-2011	Mar-2011	May-2011	Nov-2011	Oct-2011	Sep-2011	CLV
4333	18280.0	0.0	0.00	0.0	0.00	0.00	0.00	0.00	0.00	180.6	0.00	0.00	0.00	0.0	361.20
4334	18281.0	0.0	0.00	0.0	0.00	0.00	0.00	0.00	80.82	0.0	0.00	0.00	0.00	0.0	161.64
4335	18282.0	0.0	100.21	0.0	77.84	0.00	0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.0	356.10
4336	18283.0	115.6	0.00	0.0	208.00	100.95	213.75	139.89	296.52	0.0	85.22	637.71	112.99	134.9	3859.86
4337	18287.0	0.0	0.00	0.0	0.00	0.00	0.00	0.00	0.00	0.0	765.28	0.00	1072.00	0.0	3674.56

```

:
# now i will divide the given columns into two types: 1-target variable 2-feature variables
# 1- Latest 6 month as feature variables .
X = sale[['Dec-2011', 'Nov-2011', 'Oct-2011', 'Sep-2011', 'Aug-2011', 'Jul-2011']]
#2--target variable
y = sale[['CLV']]

:
X_train, X_test, y_train, y_test = train_test_split(X,y,train_size=0.8,test_size=0.2)
LR = LinearRegression()
LR.fit(X_train, y_train)
y_pred = LR.predict(X_test)

y_pred_score = LR.score(X_test, y_test)
print("Prediction score is: {}".format(y_pred_score))

Prediction score is: 0.9597429161719538

```

Thanks 😊