

CUSTOMER SHOPPING TRENDS

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INTRODUCTION

ABOUT THE DATASET

The dataset contains customer information with unique IDs, demographics (age, gender), purchase details (item, category, amount, location, size, color, season), review ratings, subscription status, shipping preferences, discount usage, promo codes, previous purchase counts, preferred payment methods, and purchase frequency. This comprehensive dataset enables the exploration of diverse customer shopping trends and behaviors.

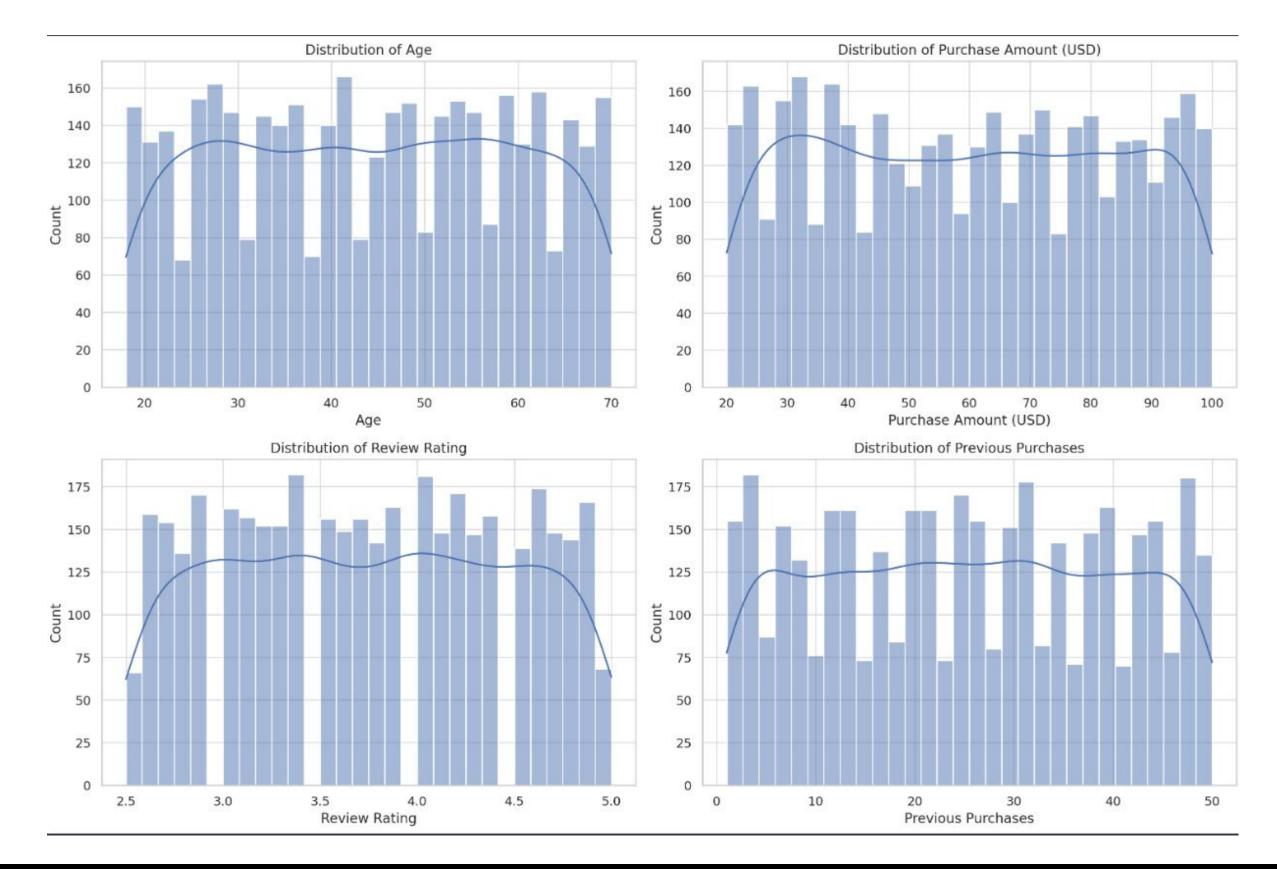
COMPRISING 4100 ROWS AND 18 COLUMNS,

DATASET

	Customer ID	000	e Gen	nder	Item Purchased	Category	Purchase Amount (USD)	Location	Size	Color	Season	Review Rating	Subscription Status	Shipping Type	Discount Applied	Promo Code Used	Previ Purcha
0	1	55	5 N	Male	Blouse	Clothing	53	Kentucky	L	Gray	Winter	3.1	Yes	Express	Yes	Yes	
1	2	19) N	Male	Sweater	Clothing	64	Maine	L	Maroon	Winter	3.1	Yes	Express	Yes	Yes	
2	3	50) N	Male	Jeans	Clothing	73	Massachusetts	s	Maroon	Spring	3.1	Yes	Free Shipping	Yes	Yes	
3	4	2	l N	Male	Sandals	Footwear	90	Rhode Island	М	Maroon	Spring	3.5	Yes	Next Day Air	Yes	Yes	
4	5	4	5 N	Male	Blouse	Clothing	49	Oregon	М	Turquoise	Spring	2.7	Yes	Free Shipping	Yes	Yes	

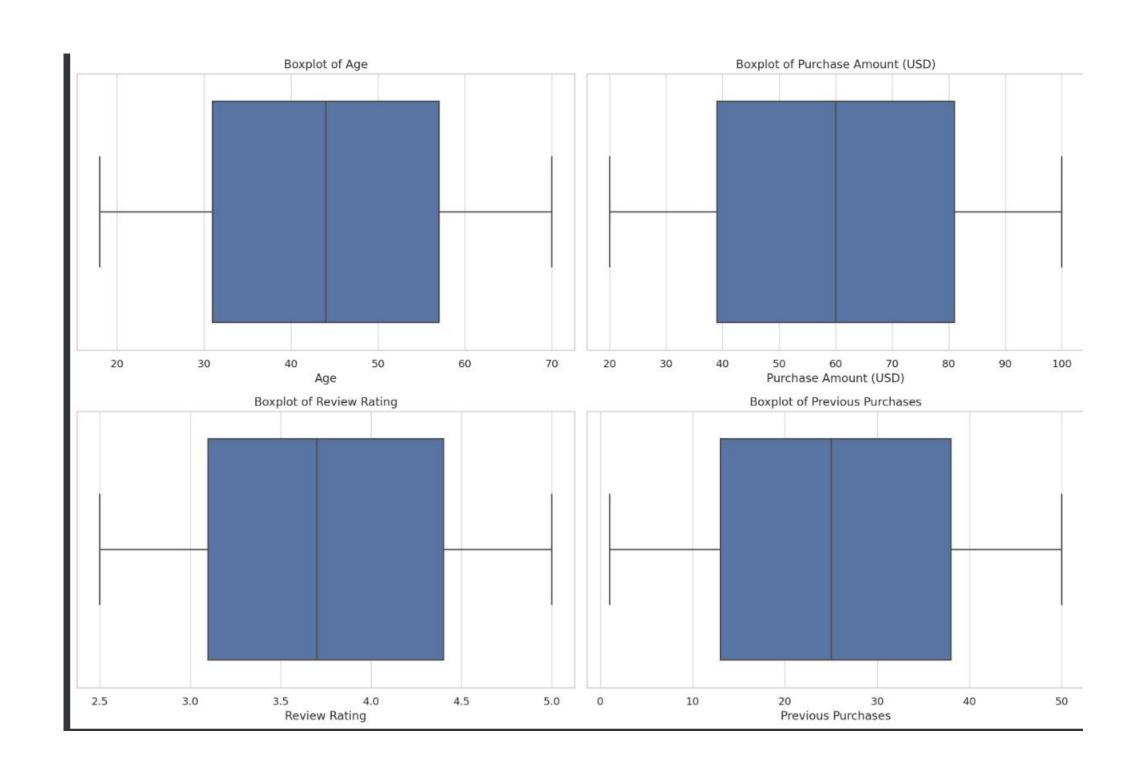
EXPLORATORY DATA ANALYSIS

HISTOGRAMS

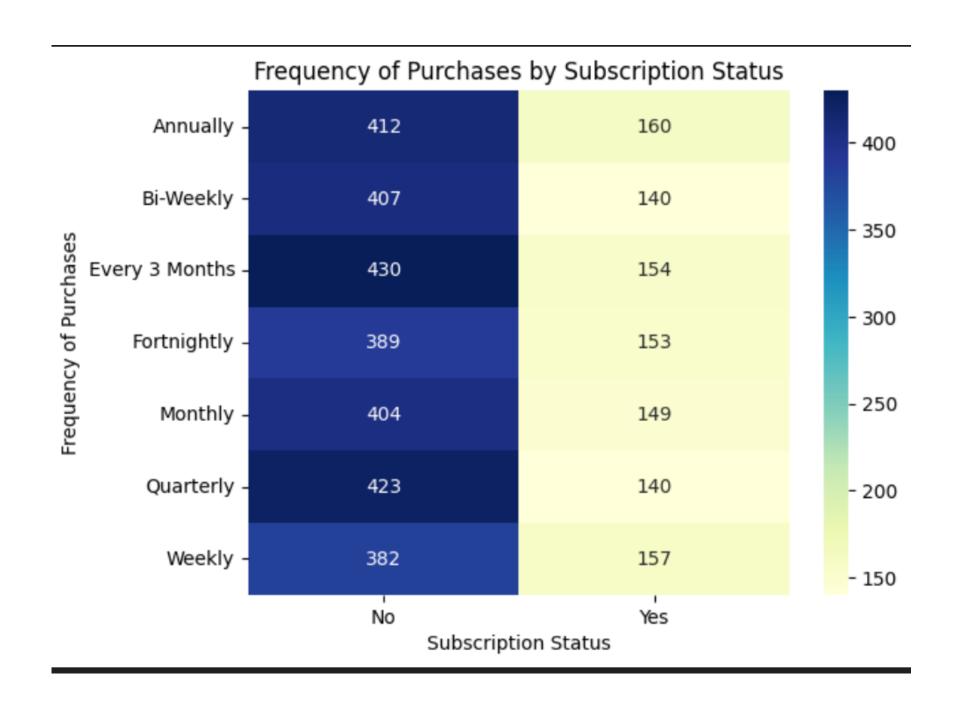


BOX PLOTS

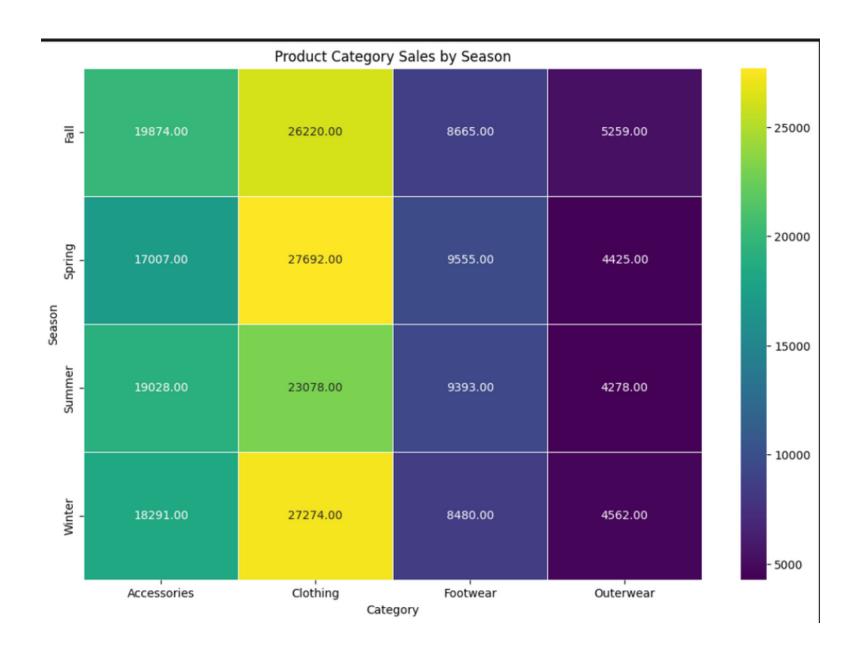
These boxplots indicating that there are no significant outliers in these numerical columns.



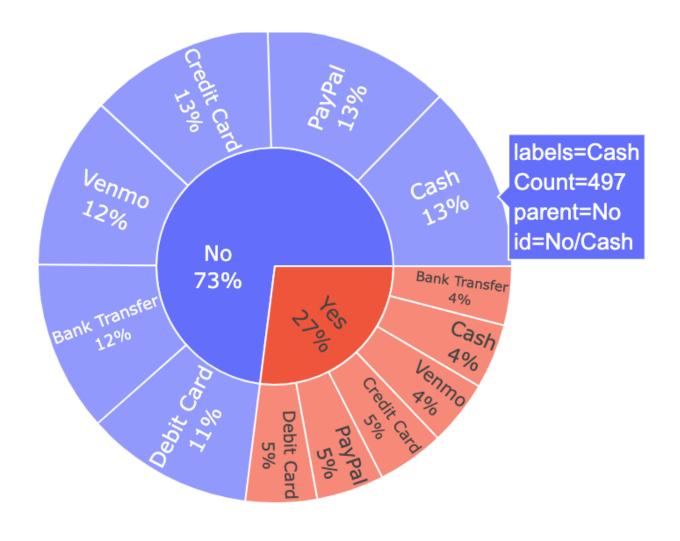
Heatmap of Frequency VS Subscription Status



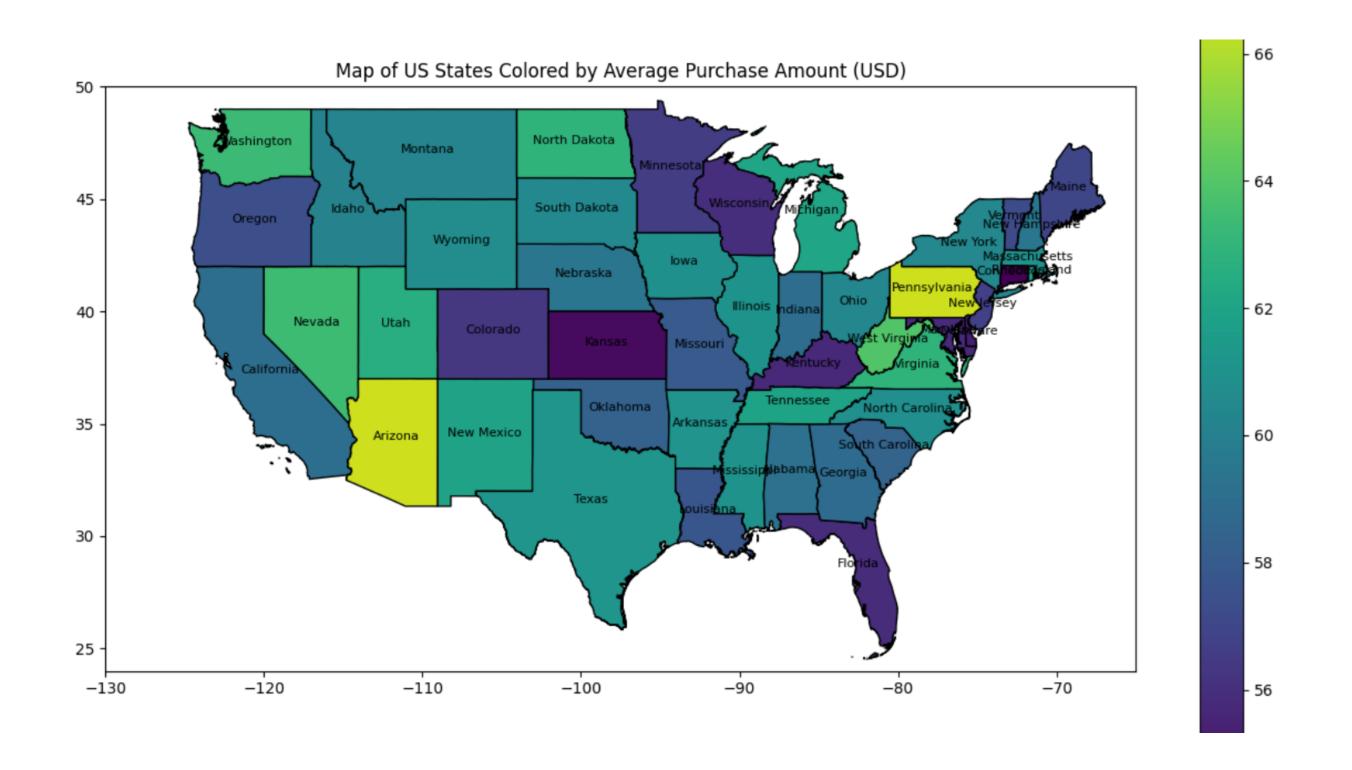
Product Category Sales by Season



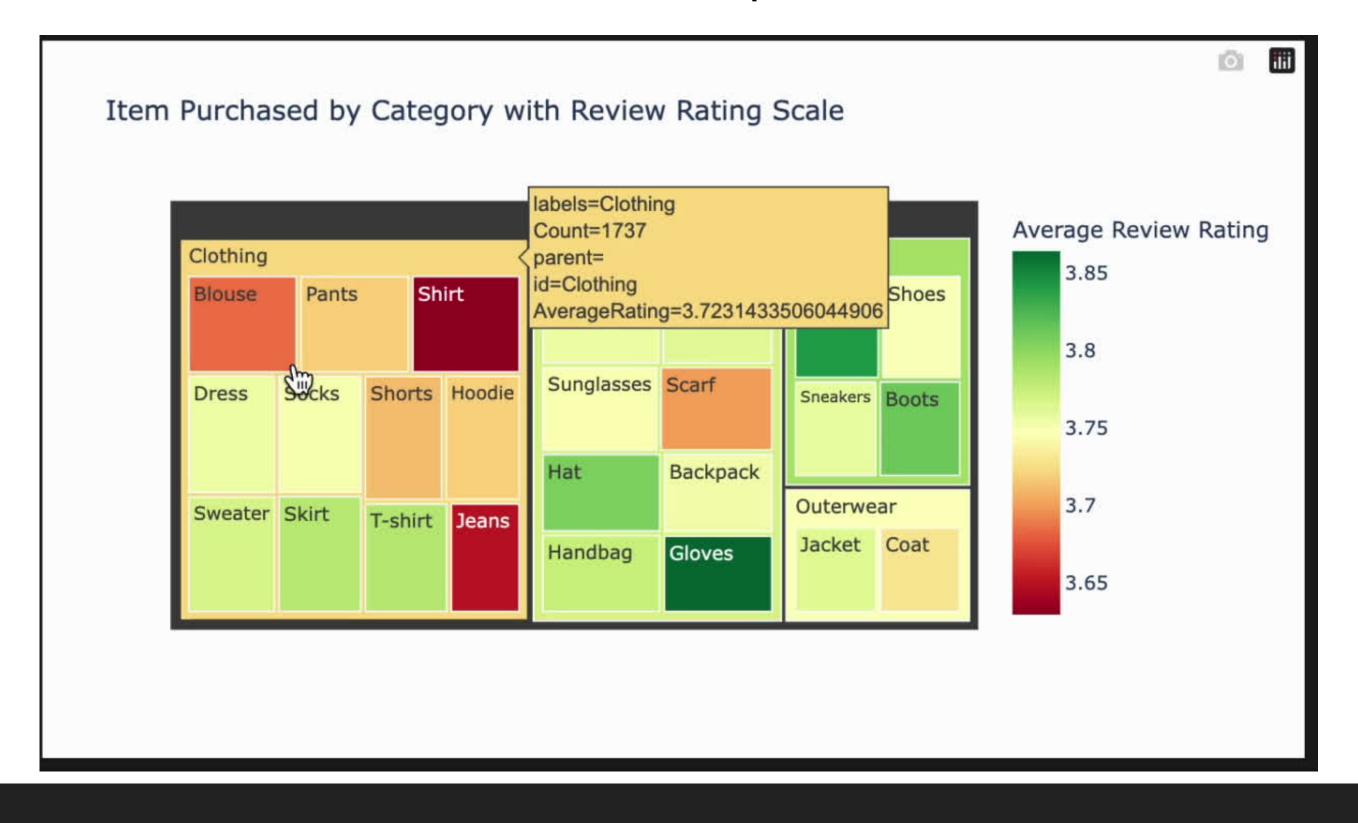
Subscription status vs Payment Method



Maps Of Location Colored By Average Purchase Amount

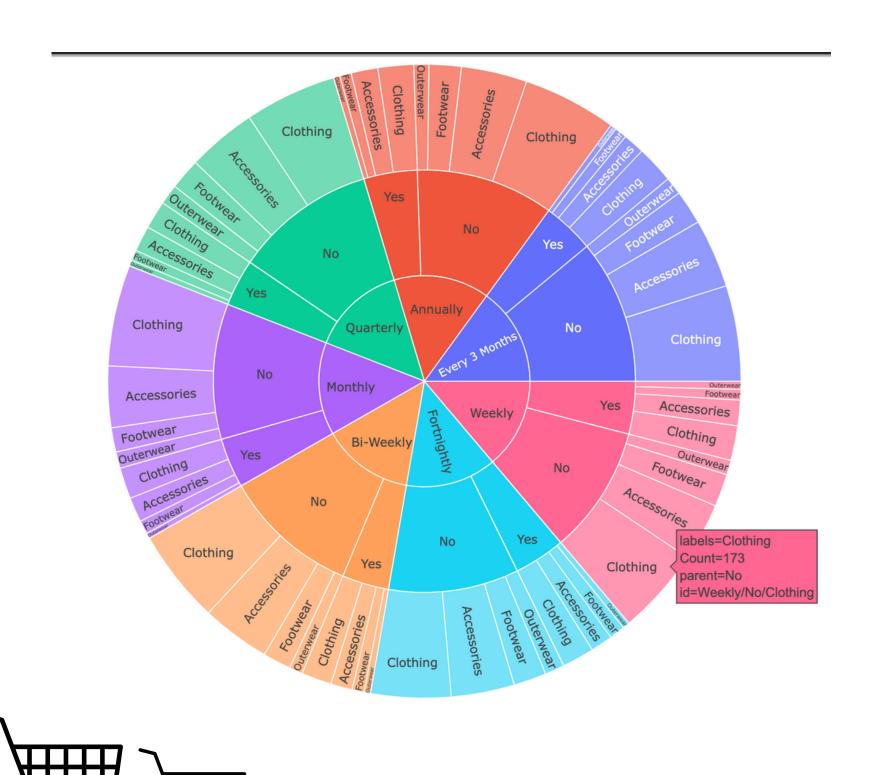


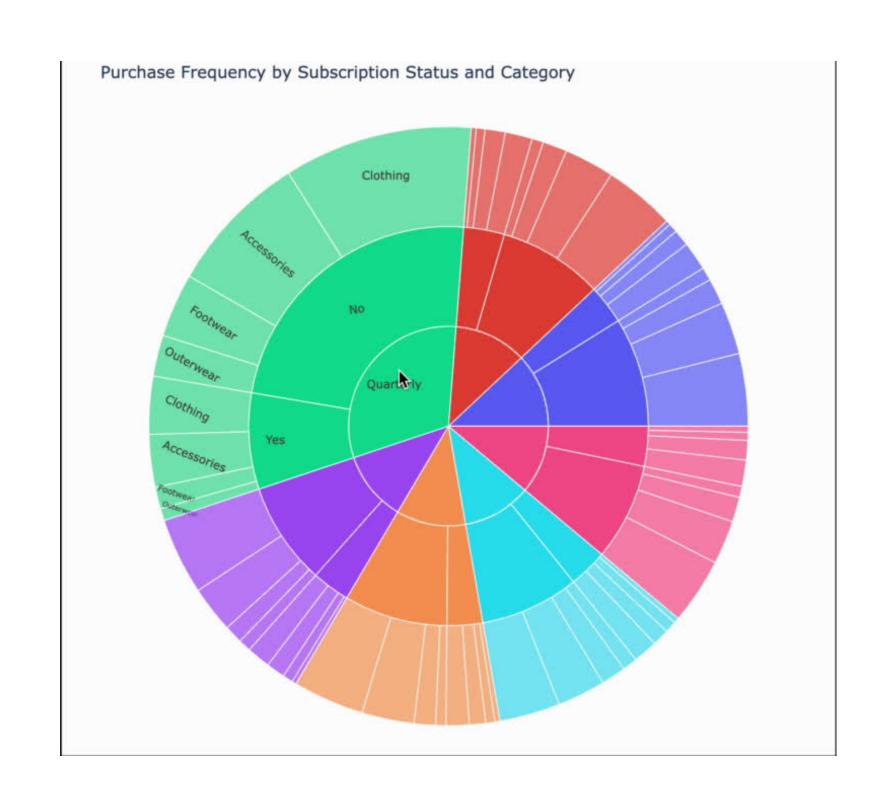
Tree Map



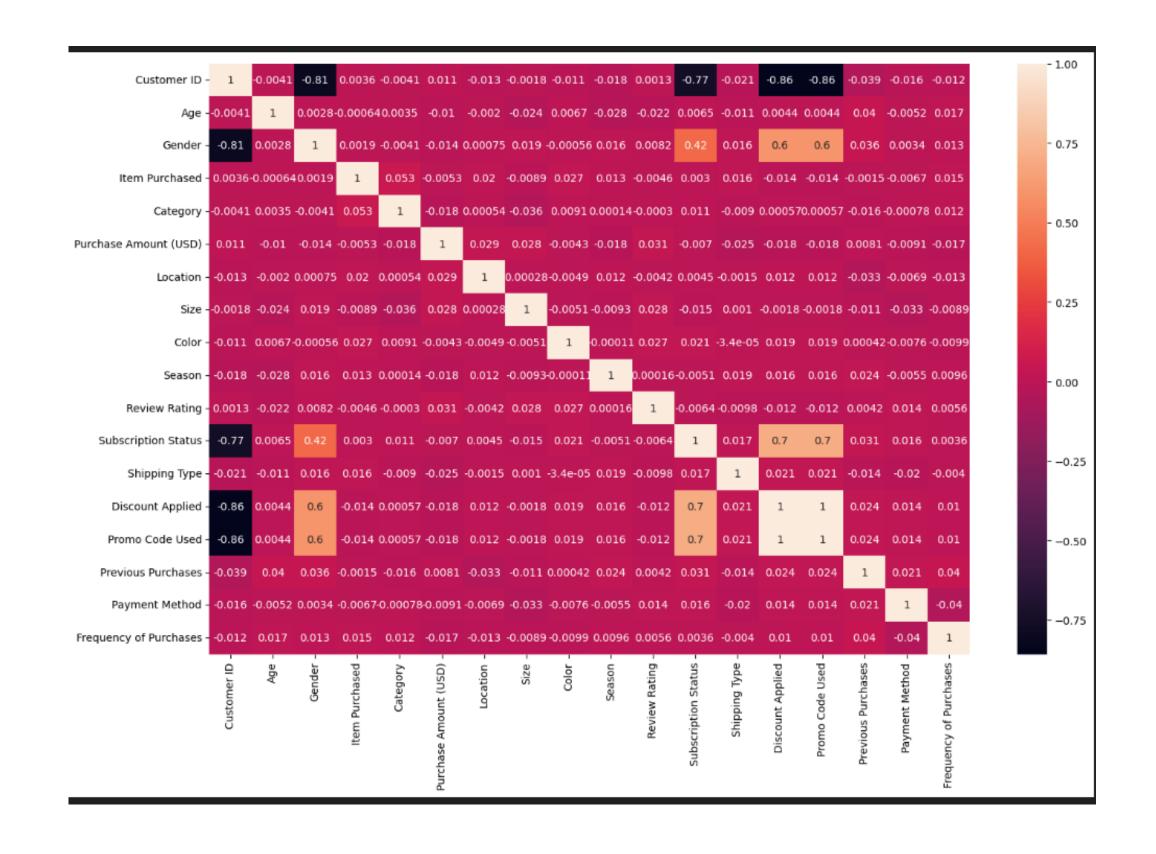
MULTIVARIANT ANALYSIS

Frequency Of Purchases VS. By Subscription VS Catrgory





Corelation between Purchases VS Subscription Status





MODELS

SMOTE ANALYSIS

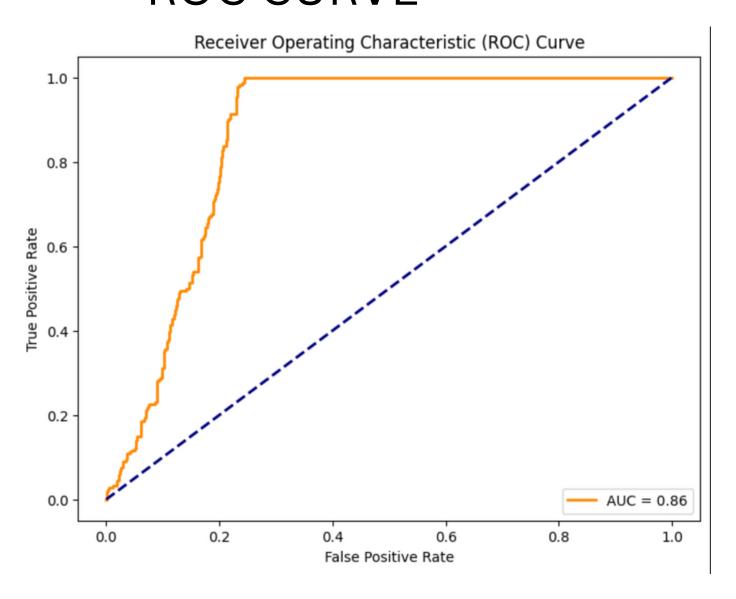
• SMOTE is a specific type of oversampling that generates synthetic examples for the minority class rather than duplicating existing instances.

• Using SMOTE with logistic regression can be particularly beneficial when dealing with imbalanced datasets, as it helps to alleviate issues related to biased models and poor generalization on the minority class.

• Variance Inflation Factor (VIF) and Multicollinearity:VIF quantifies the extent of multicollinearity in regression models by measuring how much the variance of an estimated regression coefficient increases if predictors are correlated.

Logistic Regression

ROC CURVE



Results

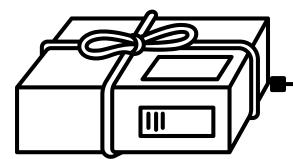
Accuracy: 0.8256410256410256 Classification Report:									
	precision	recall	f1-score	support					
0	0.99	0.77	0.86	558					
1	0.62	0.98	0.76	222					
accuracy			0.83	780					
macro avg	0.81	0.87	0.81	780					
weighted avg	0.88	0.83	0.83	780					
Confusion Matrix:									
[[427 131]									
[5 217]]									
AUC: 0.86									



Random Forest

RESULTS

Training Accuracy of the Random Forest model: 1.00 Testing Accuracy of the Random Forest model: 0.98 Accuracy of the Random Forest model: 0.98									
Classification Report:									
pr	ecision	recall	f1-score	support					
0	0.98	1.00	0.99	558					
1	1.00	0.94	0.97	222					
accuracy			0.98	780					
macro avg	0.99	0.97	0.98	780					
weighted avg	0.98	0.98	0.98	780					





CROSS VALIDATION

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✓ # # Cross Validation ...

Cross-Validation Accuracy Scores for 5 folds: [0.50512821 0.94230769 1. 1. 0.9474359 ]

Average CV Accuracy Score: 0.88
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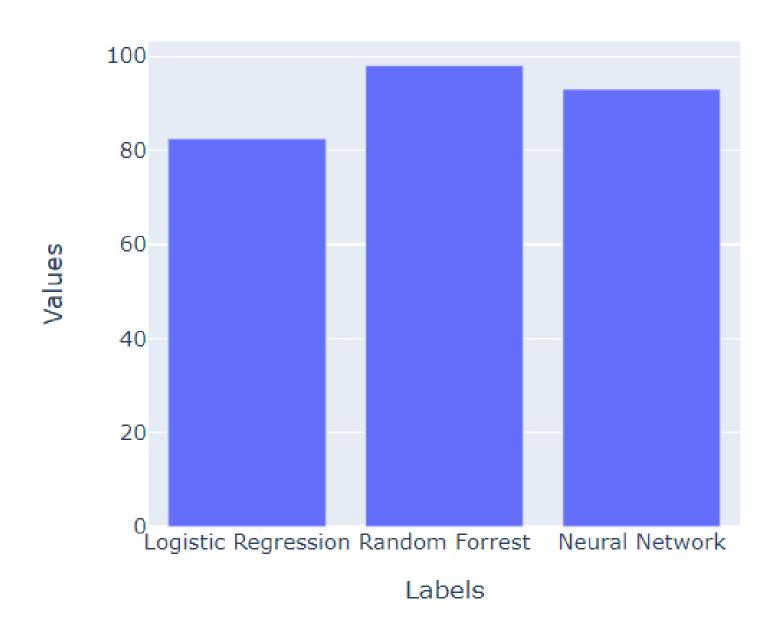


Neural Network

✓ # # Neural Network : MLPClassifier … Accuracy of the Neural Network: 0.93 Classification Report: recall f1-score precision support 0.96 0.95 834 No 0.94 0.87 0.90 0.88 336 Yes 0.93 1170 accuracy 0.91 0.92 1170 macro avg 0.92 weighted avg 0.93 0.93 0.93 1170

RESULTS

Bar Plot of accuracy



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