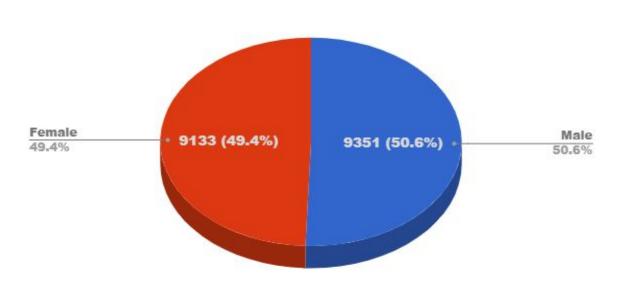
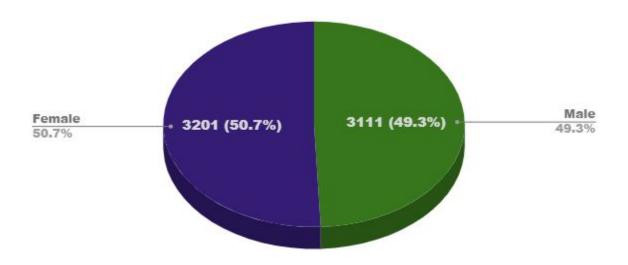
**Customer - Gender Information** 

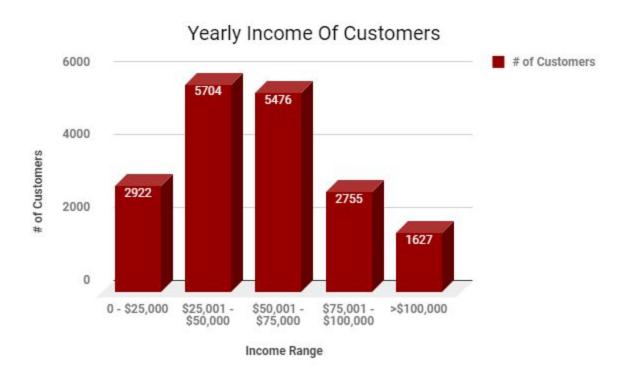
Gender Ratio



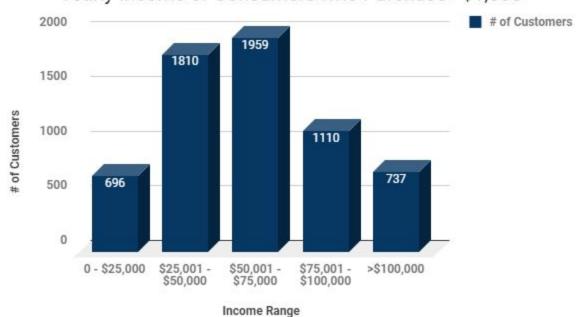
Gender Ratio with >\$1000 Spent



### **Yearly Income of Customers**

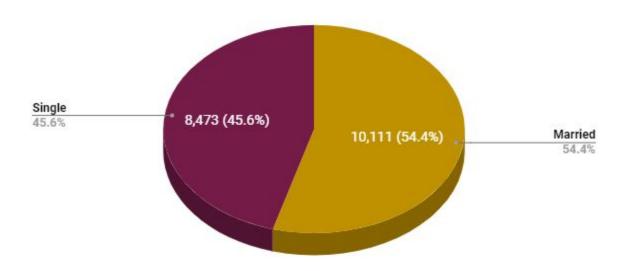


# Yearly Income of Consumers who Purchase >\$1,000

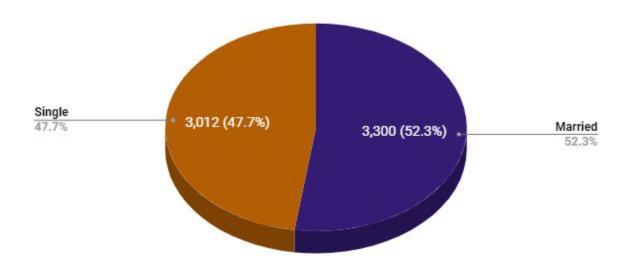


**Marital Status of Customers** 

### Marital Status of Customers



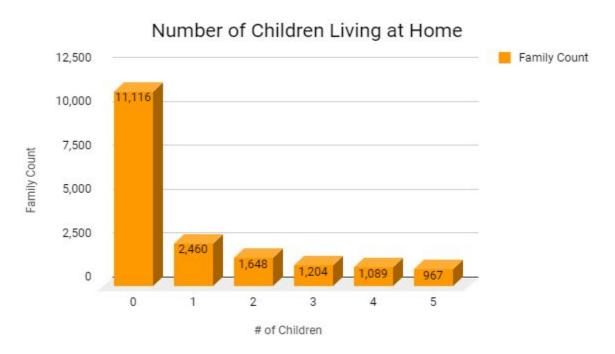
## Marital Status for Customers who Spent >\$1,000



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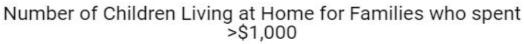
#### **Number of Children at Home**

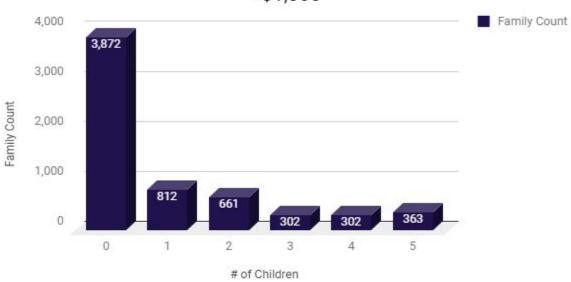
Basic Statistics:
Mean - 1.004058 Children
Std. Dev. - 1.52266



### Basic Statistics (>\$1,000 spent): Mean - 0.9605513 Children

Std. Dev. - 1.50303





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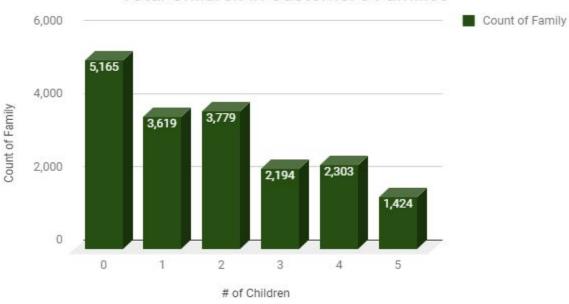
#### **Total Number of Children In Customer's Families**

#### **Basic Statistics:**

Mean - 1.844352 Children

Std. Dev. - 1.612408



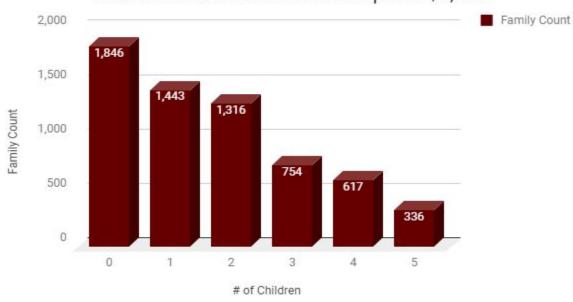


### Basic Statistics (>\$1,000 spent):

Mean - 1.661122 Children

Std. Dev. - 1.507817

## Total Children in Families who Spent >\$1,000



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#### **Total Purchase YTD**

Basic Statistics: Mean - \$1,137.655 Std. Dev. - \$1,768.029

For the Graph, because there were so many data points, I used an aggregate approach. So with the income range as the x-axis, I found the average of all of the money spent for each range, and that is the values you see on the graph. To help drive home the range of values, I included error bars that show the maximum and minimum amount of money spent by customers in the income range. On top of that, there is a trend line, which shows that the more income an individual has, the more money they spend.



Number of Customers who Spent >\$1,000

