

```
In [1]: import pandas as pd
import numpy as np
import plotly.express as px
import plotly.graph_objects as go
import plotly.io as pio
pio.templates.default = "plotly_white"

data = pd.read_csv("train.csv")
print(data.head())
```

	ID	Customer_ID	Month	Name	Age	SSN	Occupation	\
0	5634	3392	1	Aaron Maashoh	23.0	821000265.0	Scientist	
1	5635	3392	2	Aaron Maashoh	23.0	821000265.0	Scientist	
2	5636	3392	3	Aaron Maashoh	23.0	821000265.0	Scientist	
3	5637	3392	4	Aaron Maashoh	23.0	821000265.0	Scientist	
4	5638	3392	5	Aaron Maashoh	23.0	821000265.0	Scientist	

	Annual_Income	Monthly_Inhand_Salary	Num_Bank_Accounts	...	Credit_Mix	\
0	19114.12	1824.843333	3.0	...	Good	
1	19114.12	1824.843333	3.0	...	Good	
2	19114.12	1824.843333	3.0	...	Good	
3	19114.12	1824.843333	3.0	...	Good	
4	19114.12	1824.843333	3.0	...	Good	

	Outstanding_Debt	Credit_Utilization_Ratio	Credit_History_Age	\
0	809.98	26.822620	265.0	
1	809.98	31.944960	266.0	
2	809.98	28.609352	267.0	
3	809.98	31.377862	268.0	
4	809.98	24.797347	269.0	

	Payment_of_Min_Amount	Total_EMI_per_month	Amount_invested_monthly	\
0	No	49.574949	21.46538	
1	No	49.574949	21.46538	
2	No	49.574949	21.46538	
3	No	49.574949	21.46538	
4	No	49.574949	21.46538	

	Payment_Behaviour	Monthly_Balance	Credit_Score
0	High_spent_Small_value_payments	312.494089	Good
1	Low_spent_Large_value_payments	284.629162	Good
2	Low_spent_Medium_value_payments	331.209863	Good
3	Low_spent_Small_value_payments	223.451310	Good
4	High_spent_Medium_value_payments	341.489231	Good

[5 rows x 28 columns]

```
In [2]: print(data.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100000 entries, 0 to 99999
Data columns (total 28 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   ID                                     100000 non-null  int64
1   Customer_ID                           100000 non-null  int64
2   Month                                  100000 non-null  int64
3   Name                                   100000 non-null  object
4   Age                                    100000 non-null  float64
5   SSN                                    100000 non-null  float64
6   Occupation                             100000 non-null  object
7   Annual_Income                           100000 non-null  float64
8   Monthly_Inhand_Salary                   100000 non-null  float64
9   Num_Bank_Accounts                       100000 non-null  float64
10  Num_Credit_Card                         100000 non-null  float64
11  Interest_Rate                           100000 non-null  float64
12  Num_of_Loan                             100000 non-null  float64
13  Type_of_Loan                             100000 non-null  object
14  Delay_from_due_date                     100000 non-null  float64
15  Num_of_Delayed_Payment                   100000 non-null  float64
16  Changed_Credit_Limit                     100000 non-null  float64
17  Num_Credit_Inquiries                     100000 non-null  float64
18  Credit_Mix                               100000 non-null  object
19  Outstanding_Debt                         100000 non-null  float64
20  Credit_Utilization_Ratio                 100000 non-null  float64
21  Credit_History_Age                       100000 non-null  float64
22  Payment_of_Min_Amount                     100000 non-null  object
23  Total_EMI_per_month                       100000 non-null  float64
24  Amount_invested_monthly                   100000 non-null  float64
25  Payment_Behaviour                         100000 non-null  object
26  Monthly_Balance                           100000 non-null  float64
27  Credit_Score                             100000 non-null  object
dtypes: float64(18), int64(3), object(7)
memory usage: 21.4+ MB
None
```

```
In [3]: print(data.isnull().sum())
```

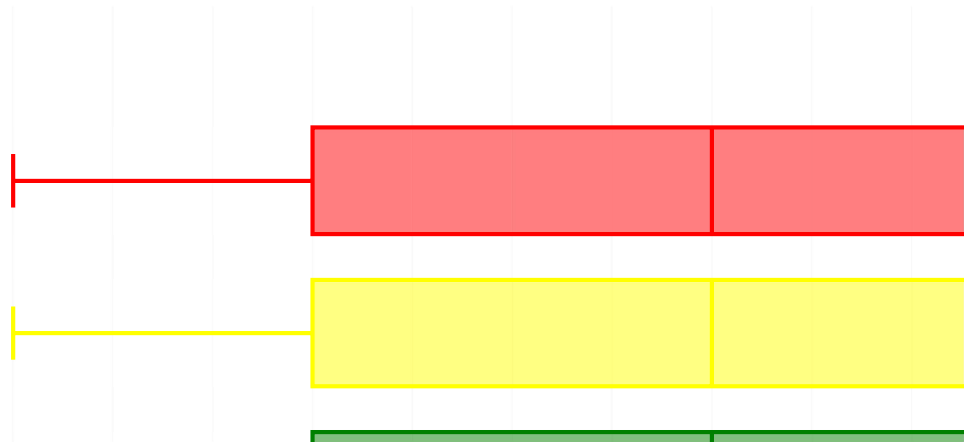
ID	0
Customer_ID	0
Month	0
Name	0
Age	0
SSN	0
Occupation	0
Annual_Income	0
Monthly_Inhand_Salary	0
Num_Bank_Accounts	0
Num_Credit_Card	0
Interest_Rate	0
Num_of_Loan	0
Type_of_Loan	0
Delay_from_due_date	0
Num_of_Delayed_Payment	0
Changed_Credit_Limit	0
Num_Credit_Inquiries	0
Credit_Mix	0
Outstanding_Debt	0
Credit_Utilization_Ratio	0
Credit_History_Age	0
Payment_of_Min_Amount	0
Total_EMI_per_month	0
Amount_invested_monthly	0
Payment_Behaviour	0
Monthly_Balance	0
Credit_Score	0
dtype:	int64

```
In [4]: data["Credit_Score"].value_counts()
```

```
Out[4]: Credit_Score
Standard    53174
Poor        28998
Good        17828
Name: count, dtype: int64
```

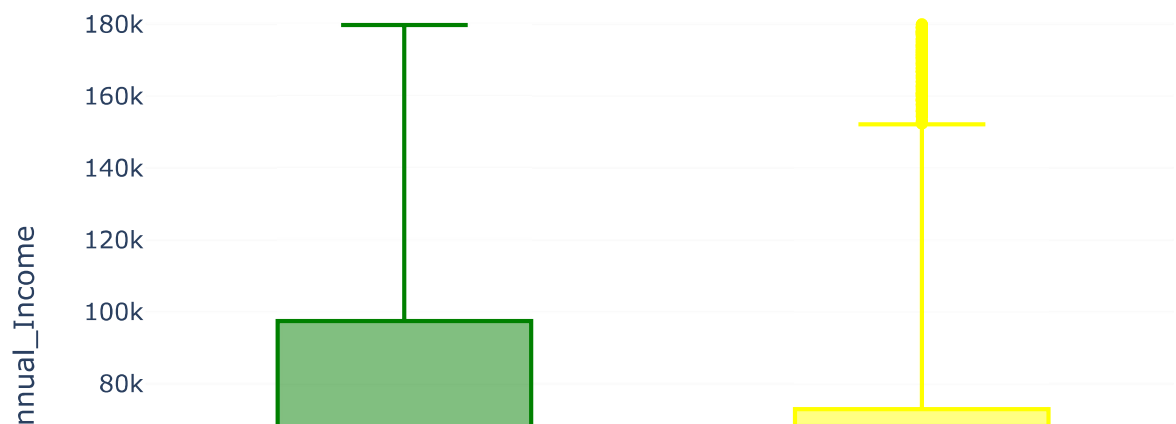
```
In [5]: fig = px.box(data,  
                    x="Occupation",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Occupation",  
                    color_discrete_map={'Poor':'red',  
                                         'Standard':'yellow',  
                                         'Good':'green'})  
  
fig.show()
```

Credit Scores Based on Occupation



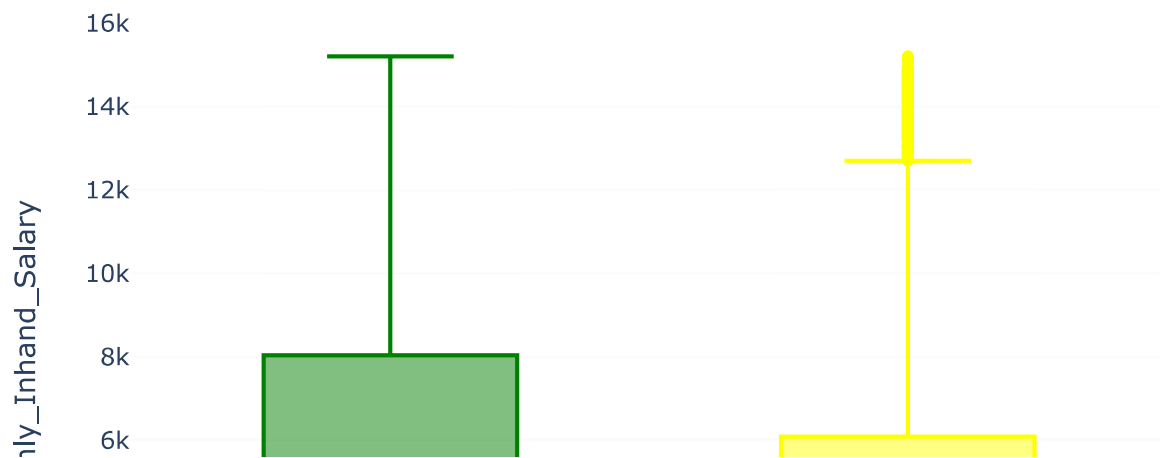
```
In [6]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Annual_Income",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Annual Income",  
                    color_discrete_map={'Poor':'red',  
                                         'Standard':'yellow',  
                                         'Good':'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on Annual Income



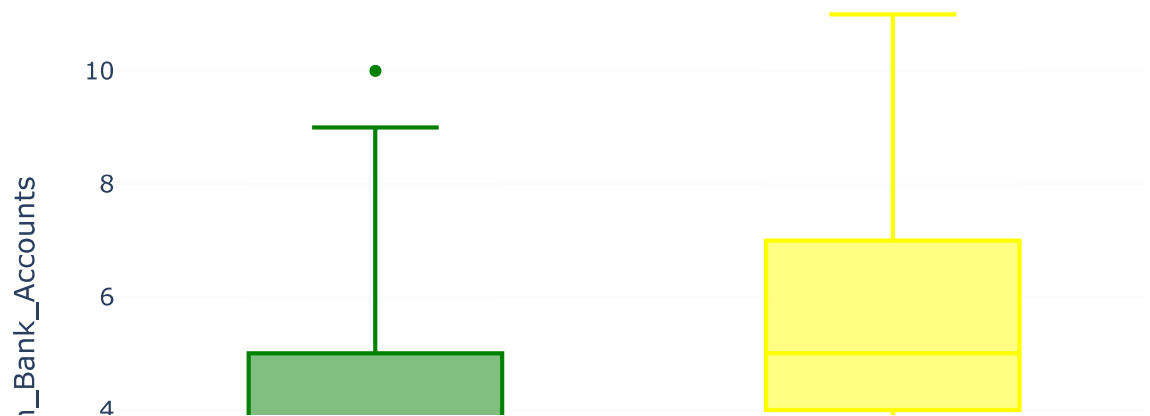
```
In [7]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Monthly_Inhand_Salary",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Monthly Inhand Salary",  
                    color_discrete_map={'Poor':'red',  
                                         'Standard':'yellow',  
                                         'Good':'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on Monthly Inhand Salary



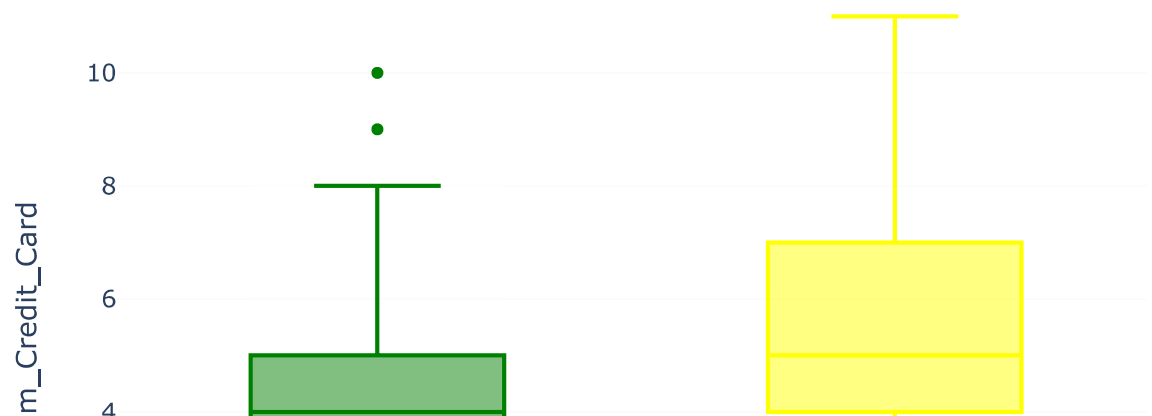
```
In [8]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Num_Bank_Accounts",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Number of Bank Accounts",  
                    color_discrete_map={'Poor':'red',  
                                         'Standard':'yellow',  
                                         'Good':'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on Number of Bank Accounts



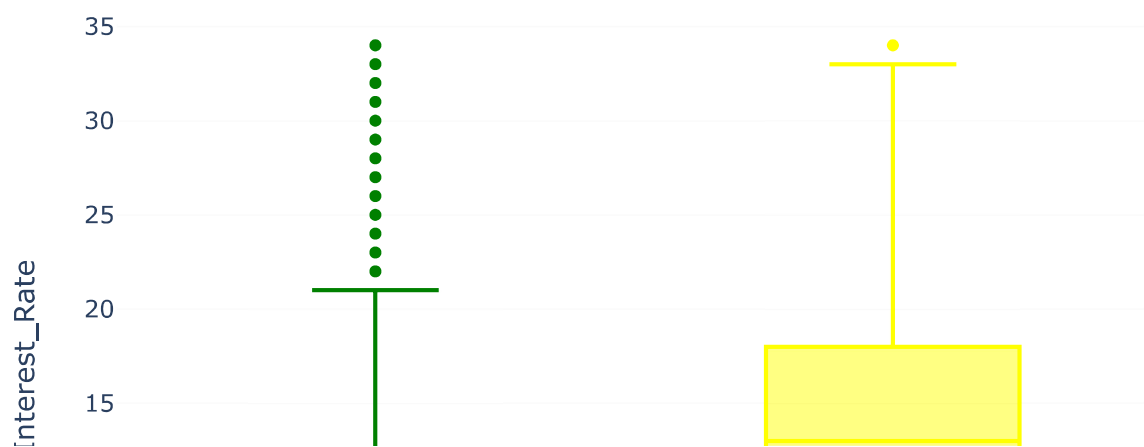
```
In [9]: fig = px.box(data,
                    x="Credit_Score",
                    y="Num_Credit_Card",
                    color="Credit_Score",
                    title="Credit Scores Based on Number of Credit cards",
                    color_discrete_map={'Poor':'red',
                                         'Standard':'yellow',
                                         'Good':'green'})
fig.update_traces(quartilemethod="exclusive")
fig.show()
```

Credit Scores Based on Number of Credit cards




```
In [10]: fig = px.box(data,  
                      x="Credit_Score",  
                      y="Interest_Rate",  
                      color="Credit_Score",  
                      title="Credit Scores Based on the Average Interest rates",  
                      color_discrete_map={'Poor':'red',  
                                          'Standard':'yellow',  
                                          'Good':'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on the Average Interest rates



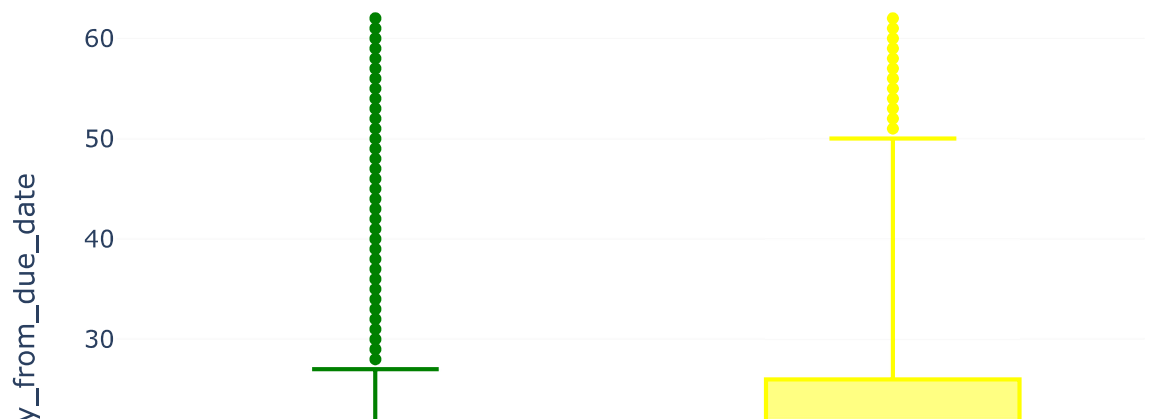
```
In [11]: fig = px.box(data,
                    x="Credit_Score",
                    y="Num_of_Loan",
                    color="Credit_Score",
                    title="Credit Scores Based on Number of Loans Taken by the Person",
                    color_discrete_map={'Poor':'red',
                                         'Standard':'yellow',
                                         'Good':'green'})
fig.update_traces(quartilemethod="exclusive")
fig.show()
```

Credit Scores Based on Number of Loans Taken by the Person



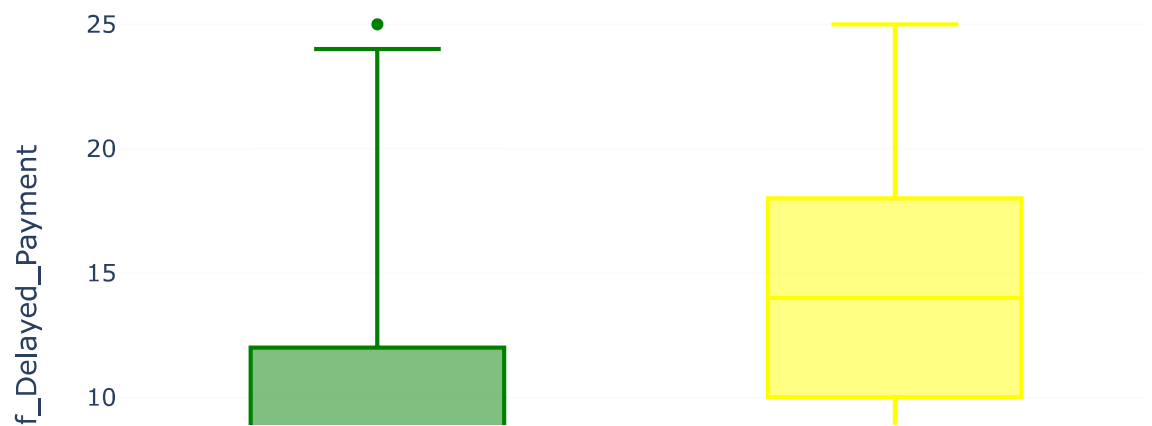
```
In [12]: fig = px.box(data,
                    x="Credit_Score",
                    y="Delay_from_due_date",
                    color="Credit_Score",
                    title="Credit Scores Based on Average Number of Days Delayed for Cr",
                    color_discrete_map={'Poor':'red',
                                         'Standard':'yellow',
                                         'Good':'green'})
fig.update_traces(quartilemethod="exclusive")
fig.show()
```

Credit Scores Based on Average Number of Days Delayed for Cr



```
In [13]: fig = px.box(data,  
                      x="Credit_Score",  
                      y="Num_of_Delayed_Payment",  
                      color="Credit_Score",  
                      title="Credit Scores Based on Number of Delayed Payments",  
                      color_discrete_map={'Poor':'red',  
                                          'Standard':'yellow',  
                                          'Good':'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on Number of Delayed Payments



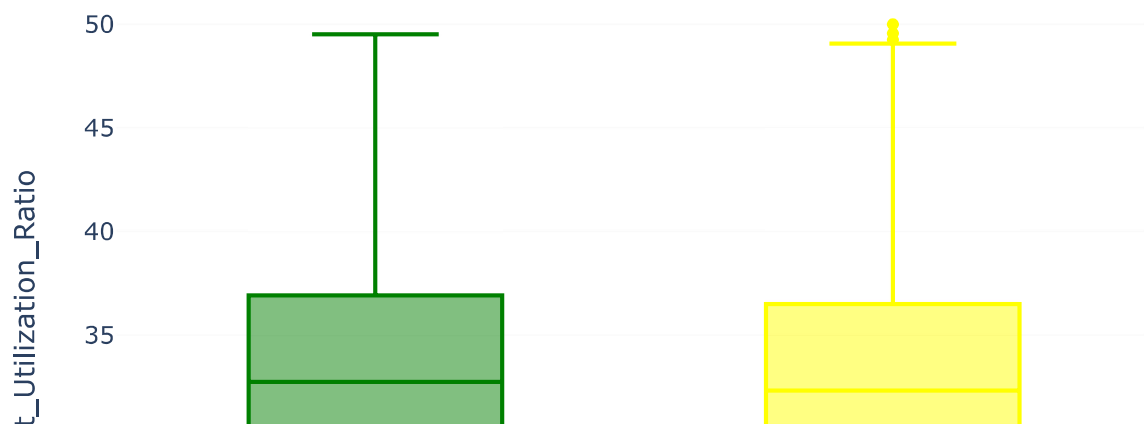
```
In [14]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Outstanding_Debt",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Outstanding Debt",  
                    color_discrete_map={'Poor':'red',  
                                         'Standard':'yellow',  
                                         'Good':'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on Outstanding Debt



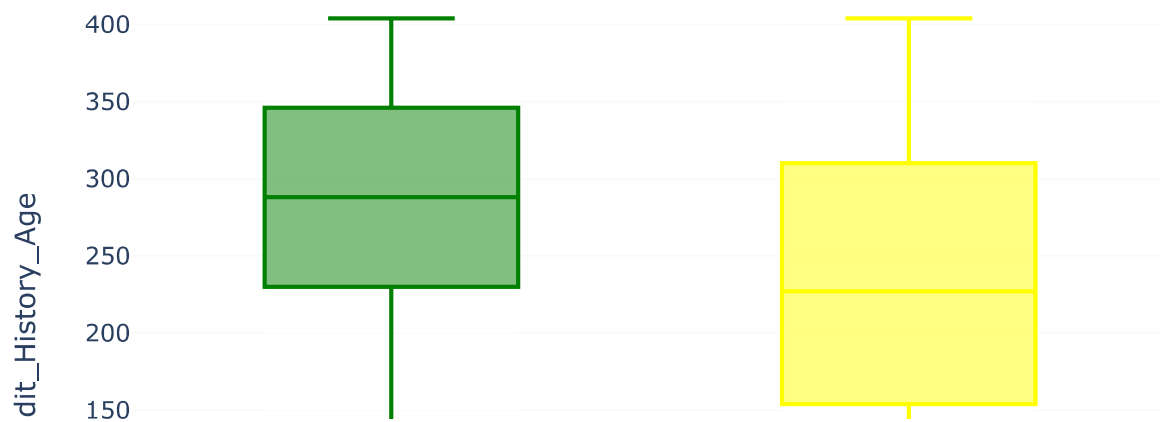
```
In [15]: fig = px.box(data,  
                      x="Credit_Score",  
                      y="Credit_Utilization_Ratio",  
                      color="Credit_Score",  
                      title="Credit Scores Based on Credit Utilization Ratio",  
                      color_discrete_map={'Poor':'red',  
                                          'Standard':'yellow',  
                                          'Good':'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on Credit Utilization Ratio



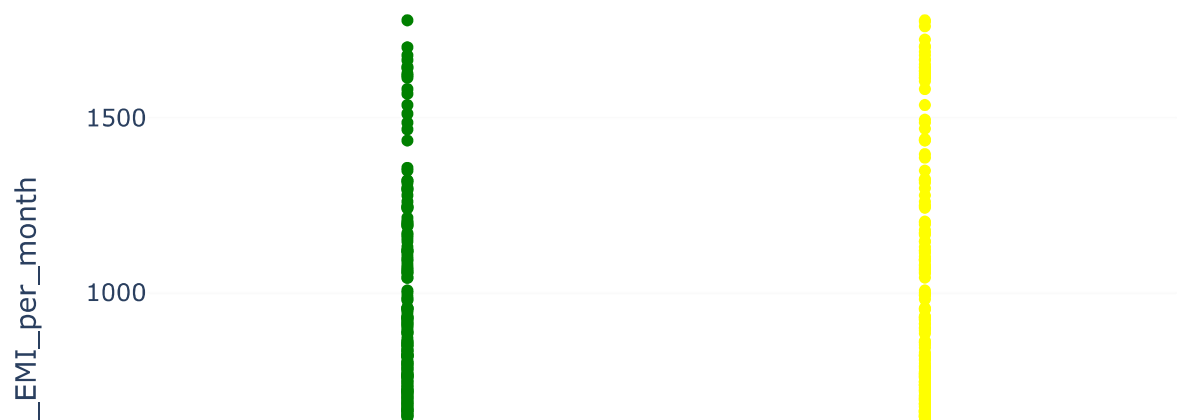
```
In [16]: fig = px.box(data,  
                      x="Credit_Score",  
                      y="Credit_History_Age",  
                      color="Credit_Score",  
                      title="Credit Scores Based on Credit History Age",  
                      color_discrete_map={'Poor':'red',  
                                           'Standard':'yellow',  
                                           'Good':'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on Credit History Age



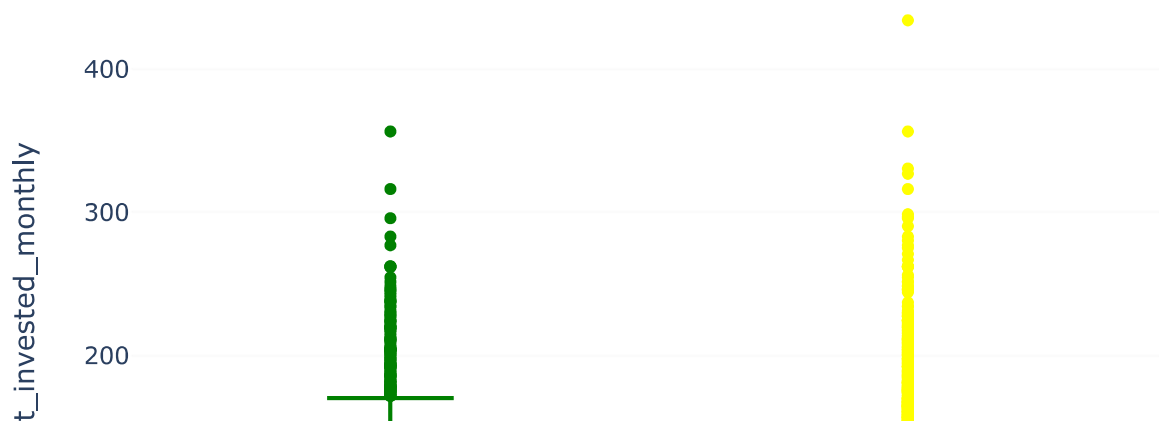
```
In [17]: fig = px.box(data,
                    x="Credit_Score",
                    y="Total_EMI_per_month",
                    color="Credit_Score",
                    title="Credit Scores Based on Total Number of EMIs per Month",
                    color_discrete_map={'Poor': 'red',
                                         'Standard': 'yellow',
                                         'Good': 'green'})
fig.update_traces(quartilemethod="exclusive")
fig.show()
```

Credit Scores Based on Total Number of EMIs per Month



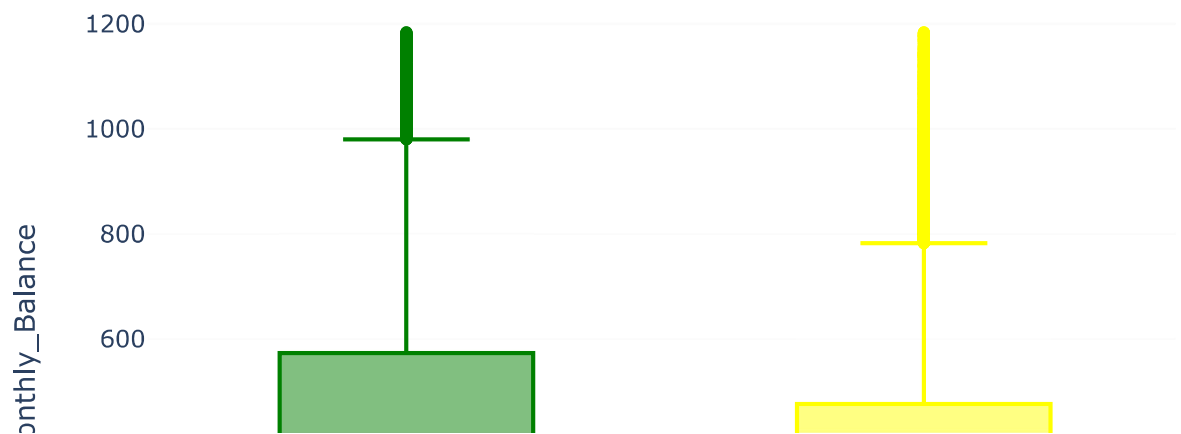

```
In [18]: fig = px.box(data,
                    x="Credit_Score",
                    y="Amount_invested_monthly",
                    color="Credit_Score",
                    title="Credit Scores Based on Amount Invested Monthly",
                    color_discrete_map={'Poor':'red',
                                       'Standard':'yellow',
                                       'Good':'green'})
fig.update_traces(quartilemethod="exclusive")
fig.show()
```

Credit Scores Based on Amount Invested Monthly



```
In [19]: fig = px.box(data,
                    x="Credit_Score",
                    y="Monthly_Balance",
                    color="Credit_Score",
                    title="Credit Scores Based on Monthly Balance Left",
                    color_discrete_map={'Poor': 'red',
                                        'Standard': 'yellow',
                                        'Good': 'green'})
fig.update_traces(quartilemethod="exclusive")
fig.show()
```

Credit Scores Based on Monthly Balance Left



```
In [20]: data["Credit_Mix"] = data["Credit_Mix"].map({"Standard": 1,
                                                       "Good": 2,
                                                       "Bad": 0})
```

```
In [21]: from sklearn.model_selection import train_test_split
x = np.array(data[["Annual_Income", "Monthly_Inhand_Salary",
                  "Num_Bank_Accounts", "Num_Credit_Card",
                  "Interest_Rate", "Num_of_Loan",
                  "Delay_from_due_date", "Num_of_Delayed_Payment",
                  "Credit_Mix", "Outstanding_Debt",
                  "Credit_History_Age", "Monthly_Balance"]])
y = np.array(data[["Credit_Score"]])
```

```
In [22]: xtrain, xtest, ytrain, ytest = train_test_split(x, y,
                                                         test_size=0.33,
                                                         random_state=42)

from sklearn.ensemble import RandomForestClassifier
model = RandomForestClassifier()
model.fit(xtrain, ytrain)
```

C:\Users\Sarang Kale\anaconda3\Lib\site-packages\sklearn\base.py:1151: DataConversionWarning:

A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().

```
Out[22]: ▼ RandomForestClassifier
RandomForestClassifier()
```

```
In [23]: print("Credit Score Prediction : ")
a = float(input("Annual Income: "))
b = float(input("Monthly Inhand Salary: "))
c = float(input("Number of Bank Accounts: "))
d = float(input("Number of Credit cards: "))
e = float(input("Interest rate: "))
f = float(input("Number of Loans: "))
g = float(input("Average number of days delayed by the person: "))
h = float(input("Number of delayed payments: "))
i = input("Credit Mix (Bad: 0, Standard: 1, Good: 3) : ")
j = float(input("Outstanding Debt: "))
k = float(input("Credit History Age: "))
l = float(input("Monthly Balance: "))

features = np.array([[a, b, c, d, e, f, g, h, i, j, k, l]])
print("Predicted Credit Score = ", model.predict(features))
```

```
Credit Score Prediction :
Annual Income: 19114.12
Monthly Inhand Salary: 1824.843333
Number of Bank Accounts: 3
Number of Credit cards: 4
Interest rate: 3
Number of Loans: 4
Average number of days delayed by the person: 3
Number of delayed payments: 7
Credit Mix (Bad: 0, Standard: 1, Good: 3) : 3
Outstanding Debt: 809.98
Credit History Age: 265
Monthly Balance: 312.4940887
Predicted Credit Score = ['Good']
```

In []:

In []:

In []: