

```
In [2]: 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 [3]: 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 [4]: print(data.isnull().sum())
```

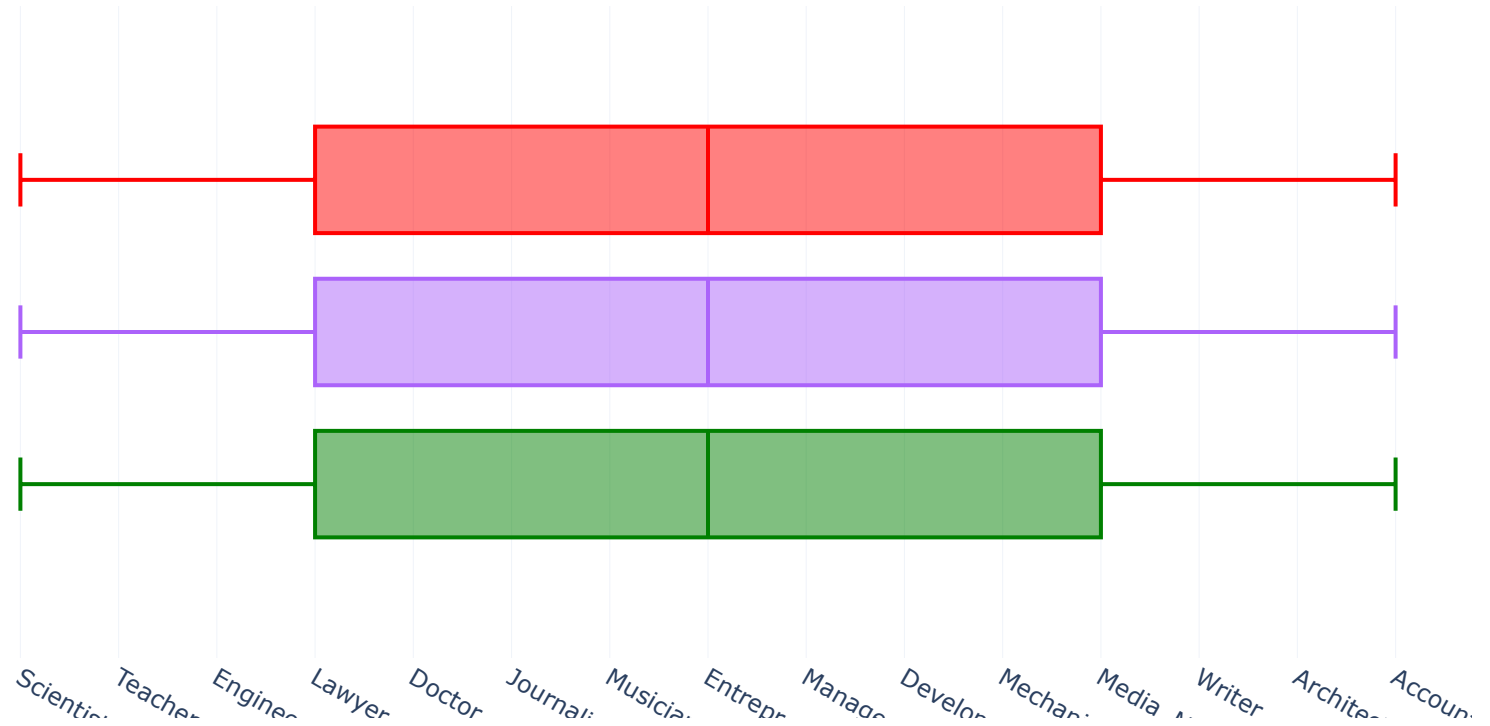
```
ID          0
Customer_ID 0
Month        0
Name         0
Age          0
SSN          0
Occupation   0
Annual_Income 0
Monthly_Inhand_Salary 0
Num_Bank_Accounts 0
Num_Credit_Card 0
Interest_Rate 0
Num_of_Loan  0
Type_of_Loan 0
Delay_from_due_date 0
Num_of_Delayed_Payment 0
Changed_Credit_Limit 0
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 [5]: data["Credit_Score"].value_counts()
```

```
Out[5]: Credit_Score
Standard    53174
Poor        28998
Good        17828
Name: count, dtype: int64
```

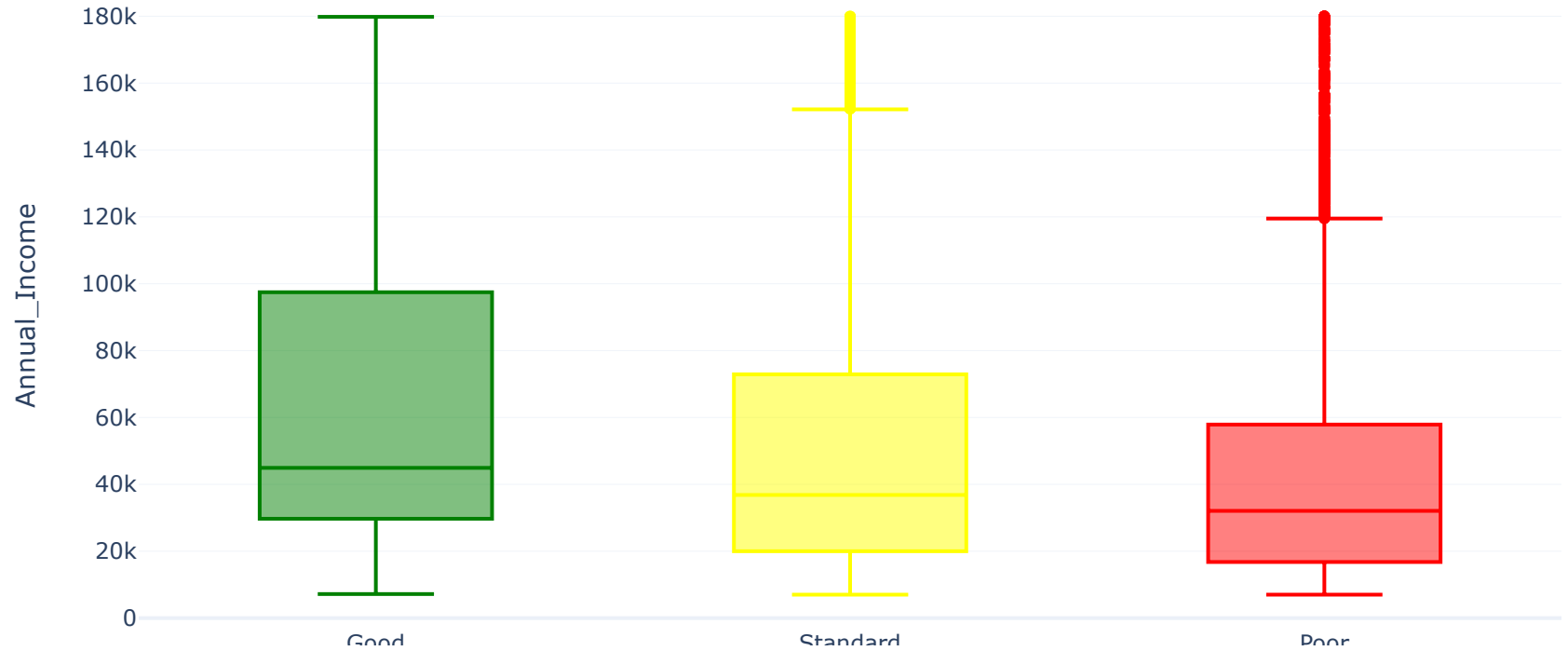
```
In [6]: fig = px.box(data,  
                    x="Occupation",  
                    color="Credit_Score",  
                    title="Credit Score Based on Occupation",  
                    color_discrete_map={'Poor' : 'red',  
                                       'Standard' : 'yellow',  
                                       'Good' : 'green'})  
  
fig.show()
```

Credit Score Based on Occupation



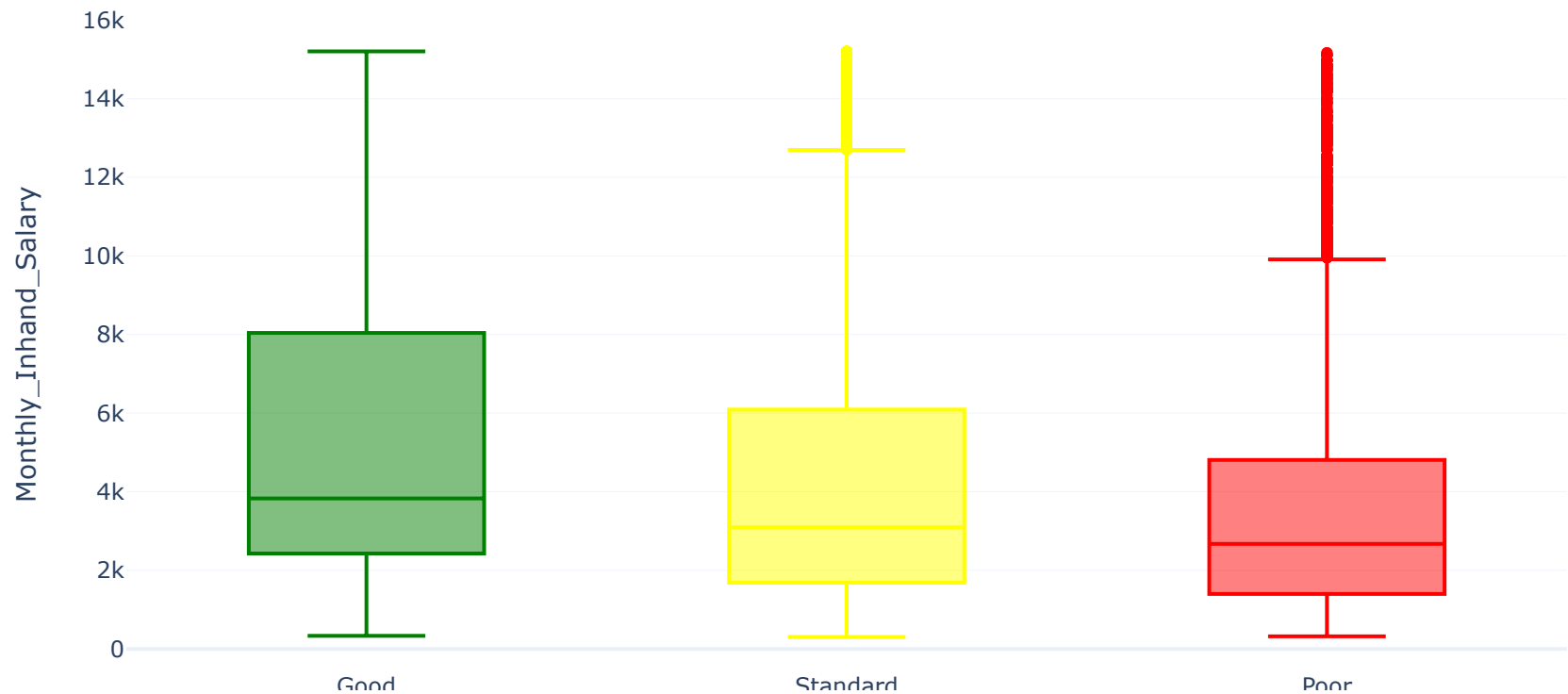
```
In [7]: fig = px.box(data,  
                    x='Credit_Score',  
                    y='Annual_Income',  
                    color='Credit_Score',  
                    title="Credit Score Based on Annual Income",  
                    color_discrete_map={'Poor':'red',  
                   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

Credit Score Based on Annual Income



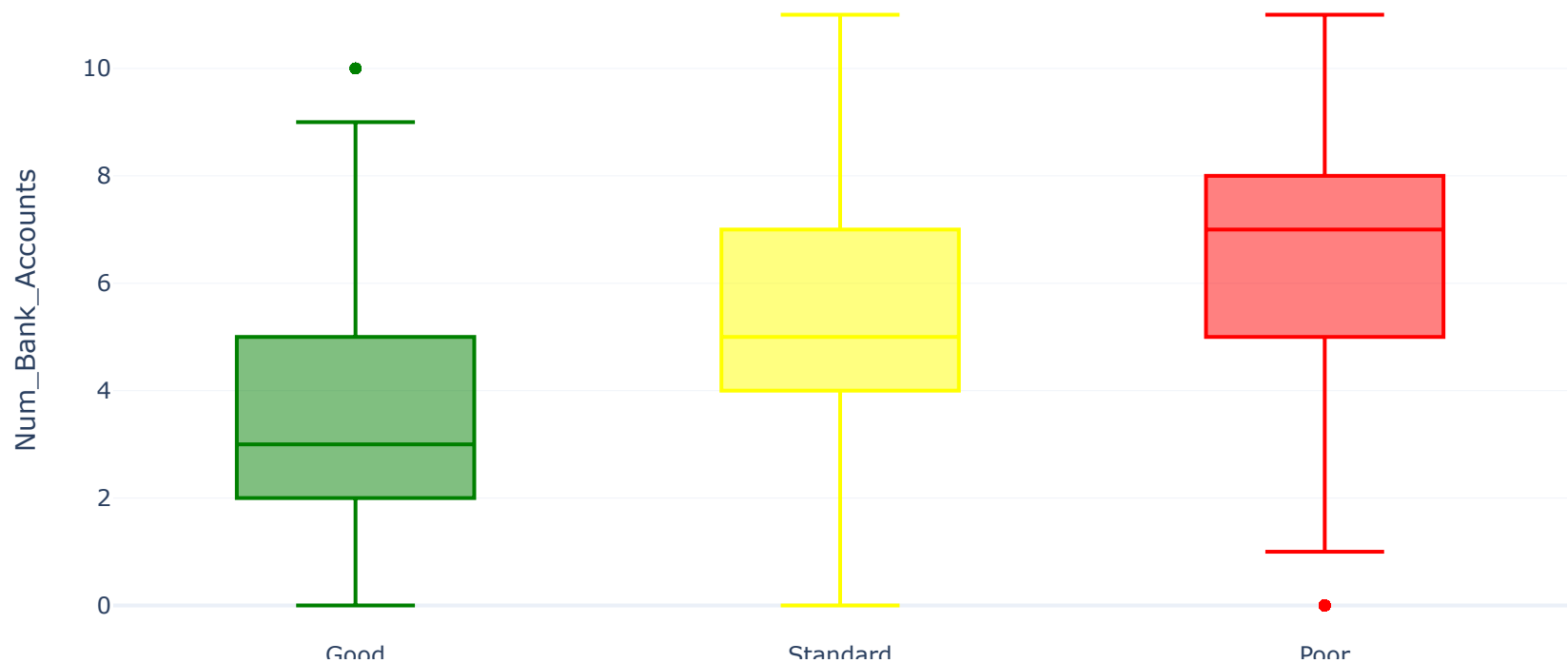
```
In [8]: fig = px.box(data,  
                    x='Credit_Score',  
                    y='Monthly_Inhand_Salary',  
                    color='Credit_Score',  
                    title='Credit Score based on Inhand Slary',  
                    color_discrete_map={'Poor':'red',  
                                       'Standard':'yellow',  
                                       'Good':'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Score based on Inhand Slary



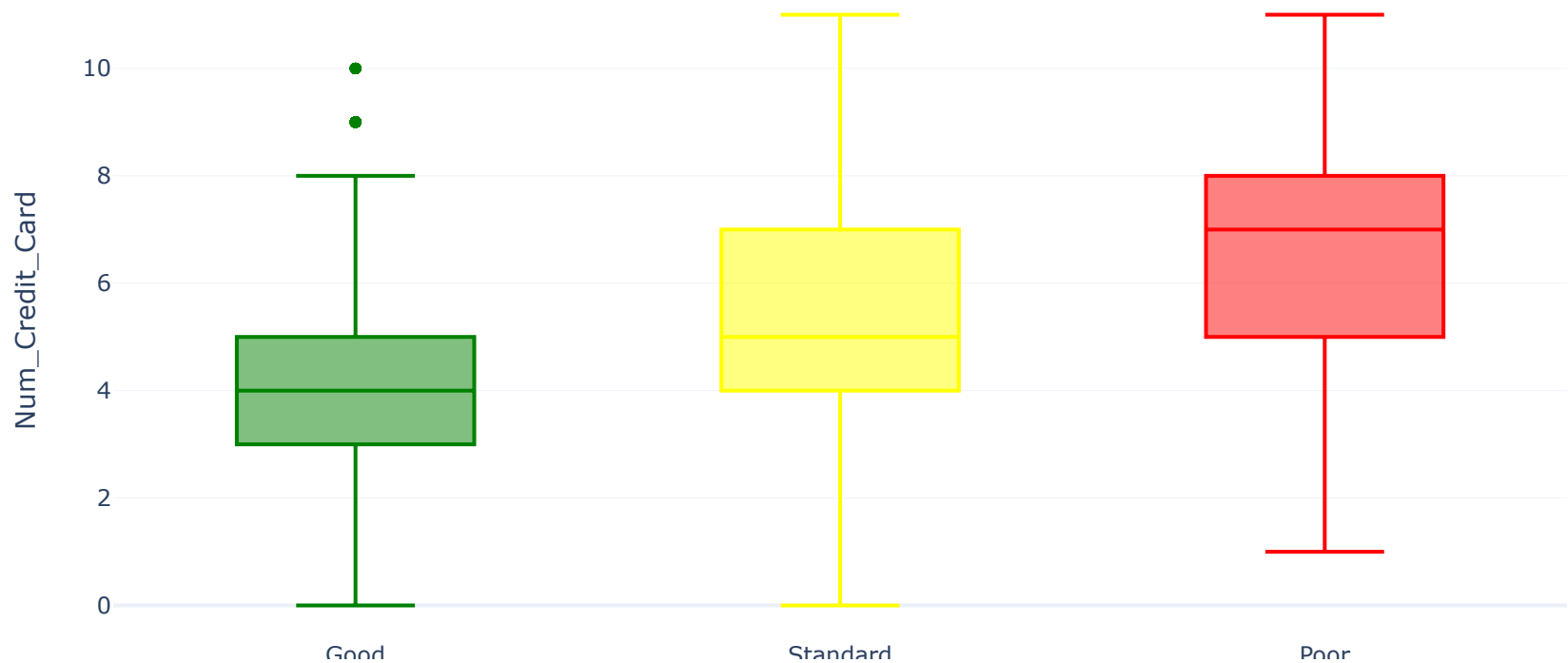

```
In [9]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Num_Bank_Accounts",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Number of Bank Accounts",  
                    color_discrete_map={'Poor': 'red',  
   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

Credit Scores Based on Number of Bank Accounts



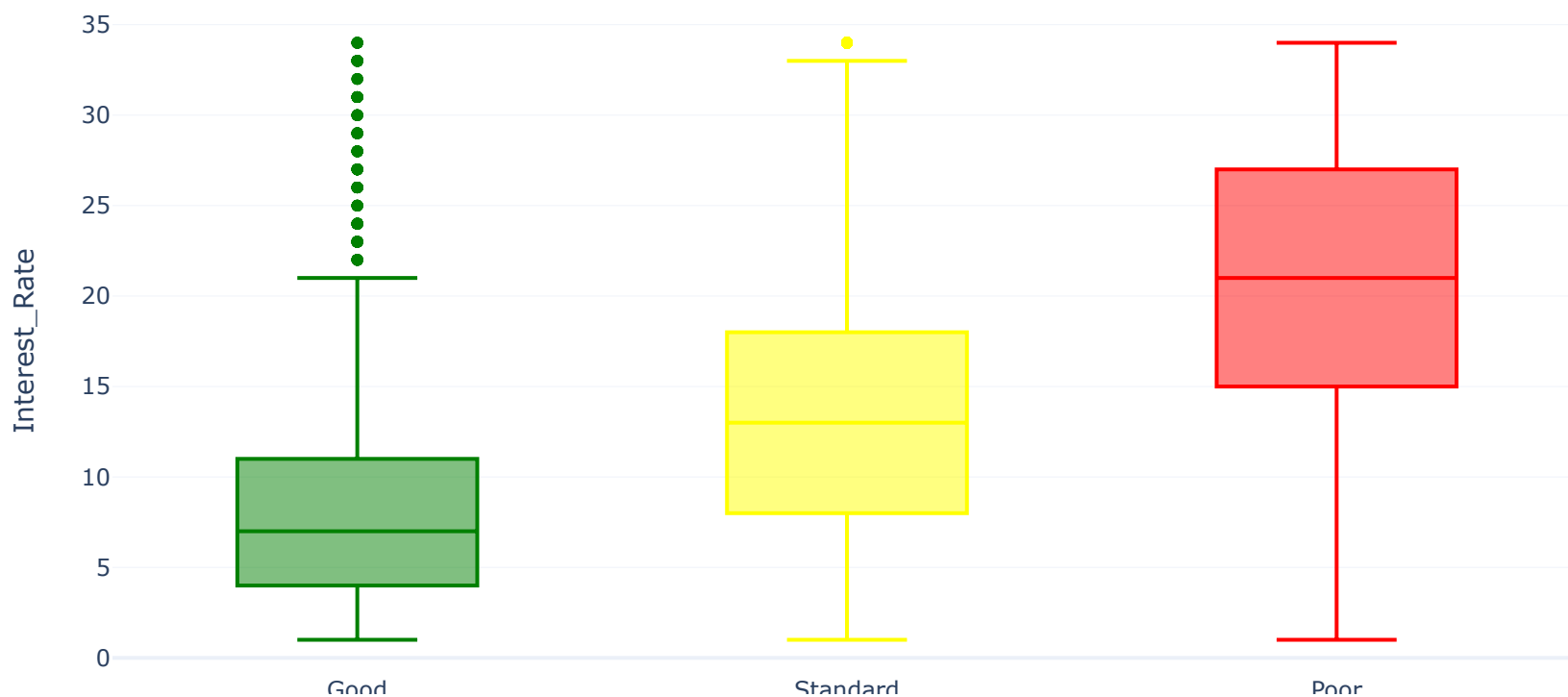
```
In [10]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Num_Credit_Card",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Number of Credit cards",  
                    color_discrete_map={'Poor': 'red',  
     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

Credit Scores Based on Number of Credit cards



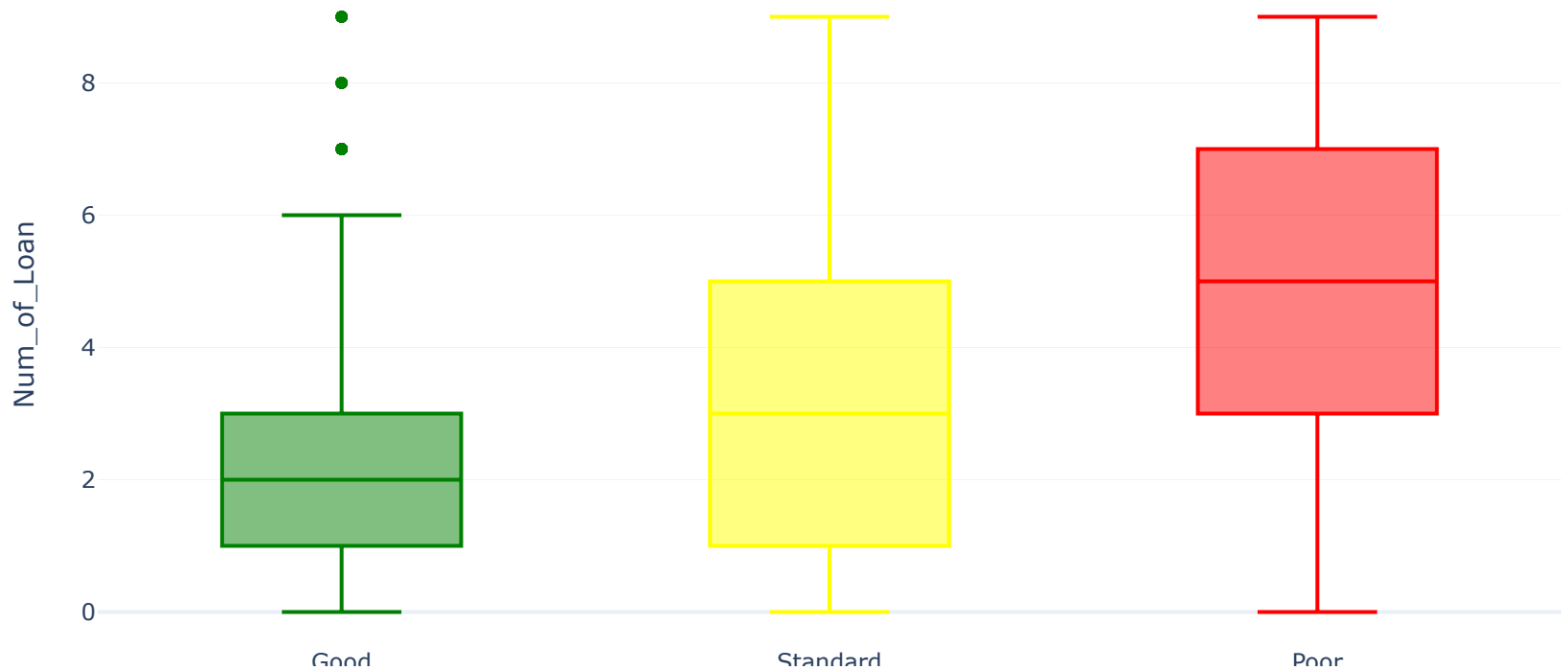
```
In [11]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Interest_Rate",  
                    color="Credit_Score",  
                    title="Credit Scores Based on The Avarage Interest Rates",  
                    color_discrete_map={'Poor': 'red',  
                                         'Standard': 'yellow',  
                                         'Good': 'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on The Avarage Interest Rates



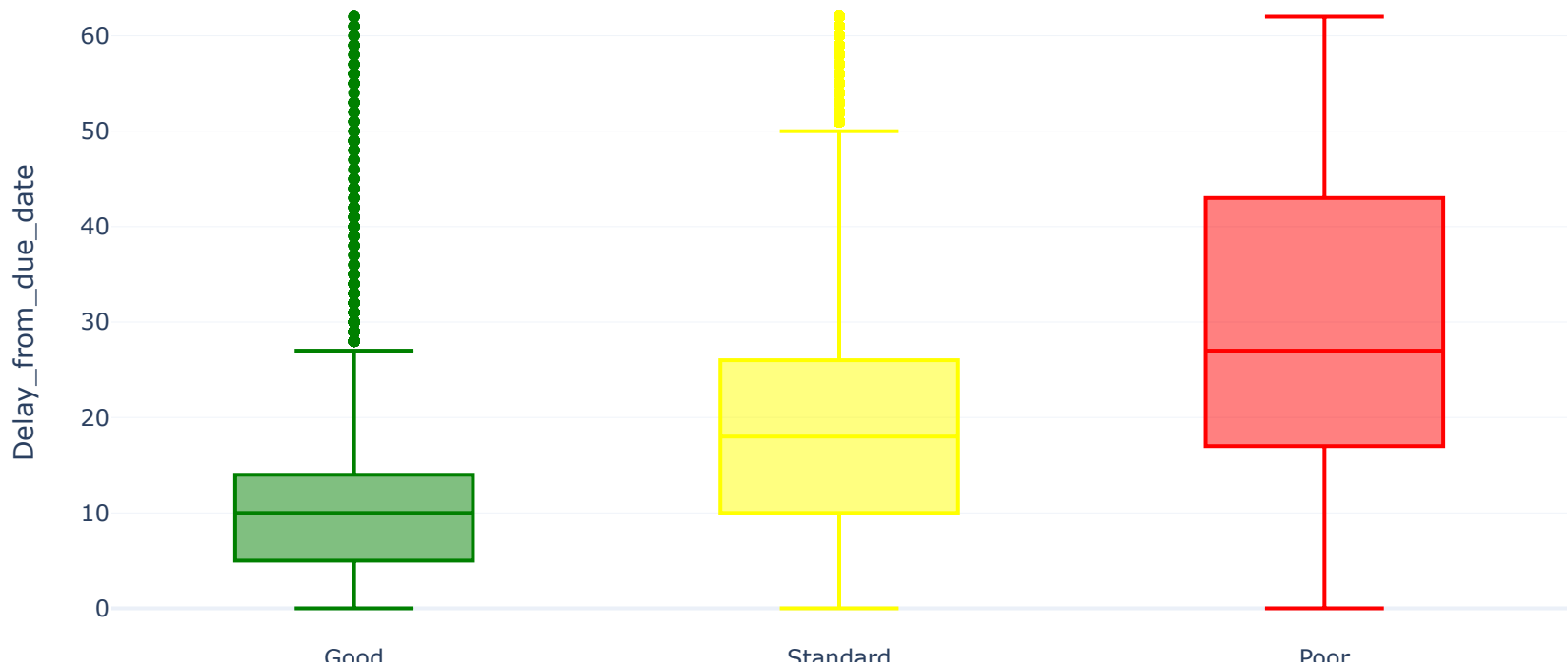
```
In [12]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Num_of_Loan",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Number of Loan",  
                    color_discrete_map={'Poor':'red',  
                  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

Credit Scores Based on Number of Loan



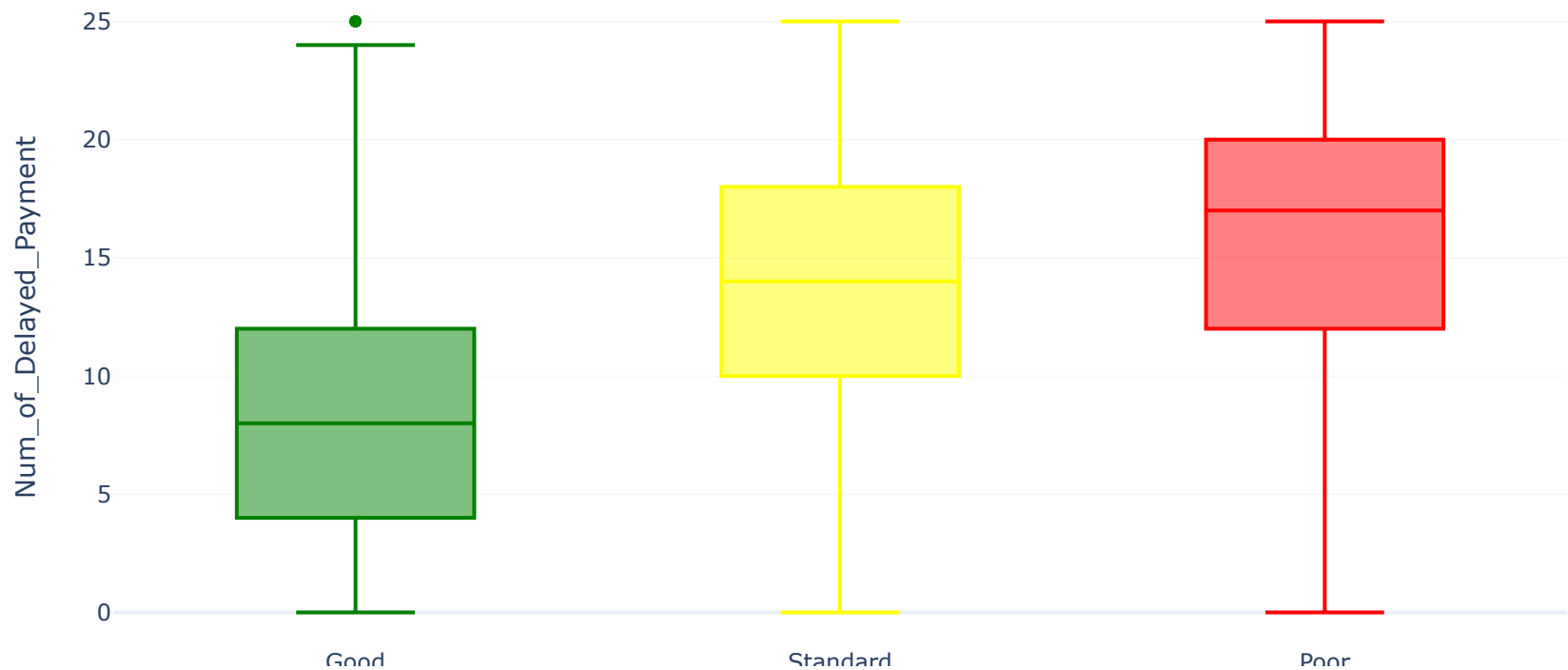
```
In [13]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Delay_from_due_date",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Delay from due date",  
                    color_discrete_map={'Poor': 'red',  
                                         'Standard': 'yellow',  
                                         'Good': 'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on Delay from due date



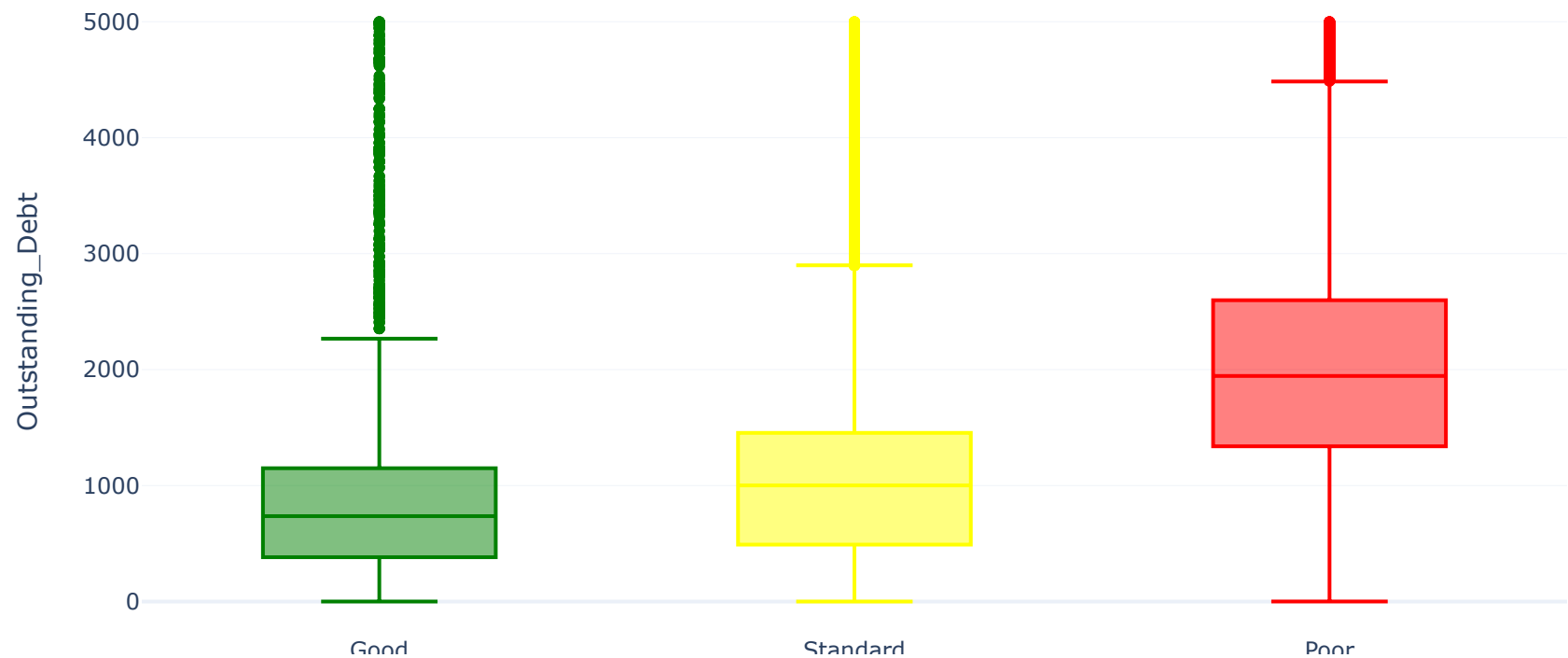
```
In [14]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Num_of_Delayed_Payment",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Number of Delayed Payments",  
                    color_discrete_map={'Poor': 'red',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

Credit Scores Based on Number of Delayed Payments



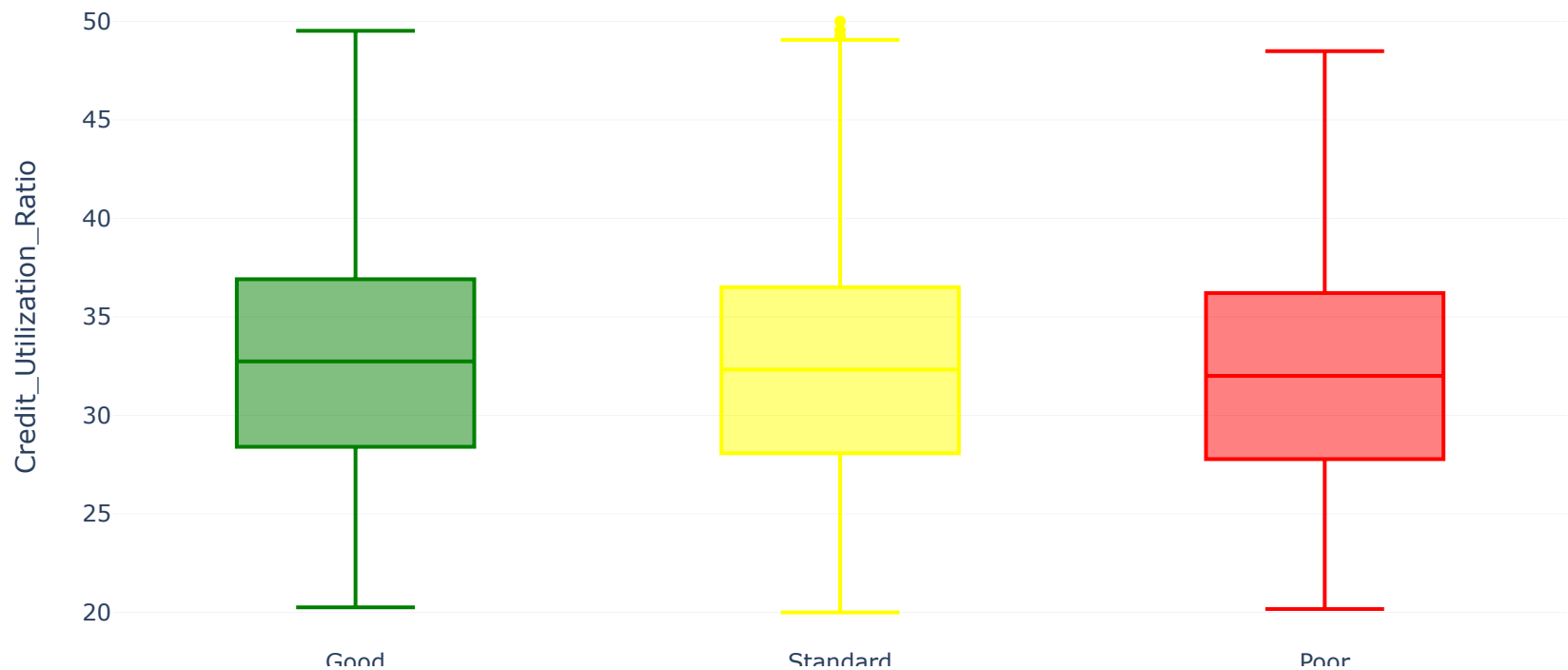
```
In [15]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Outstanding_Debt",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Outstanding Debt",  
                    color_discrete_map={'Poor': 'red',  
          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

Credit Scores Based on Outstanding Debt

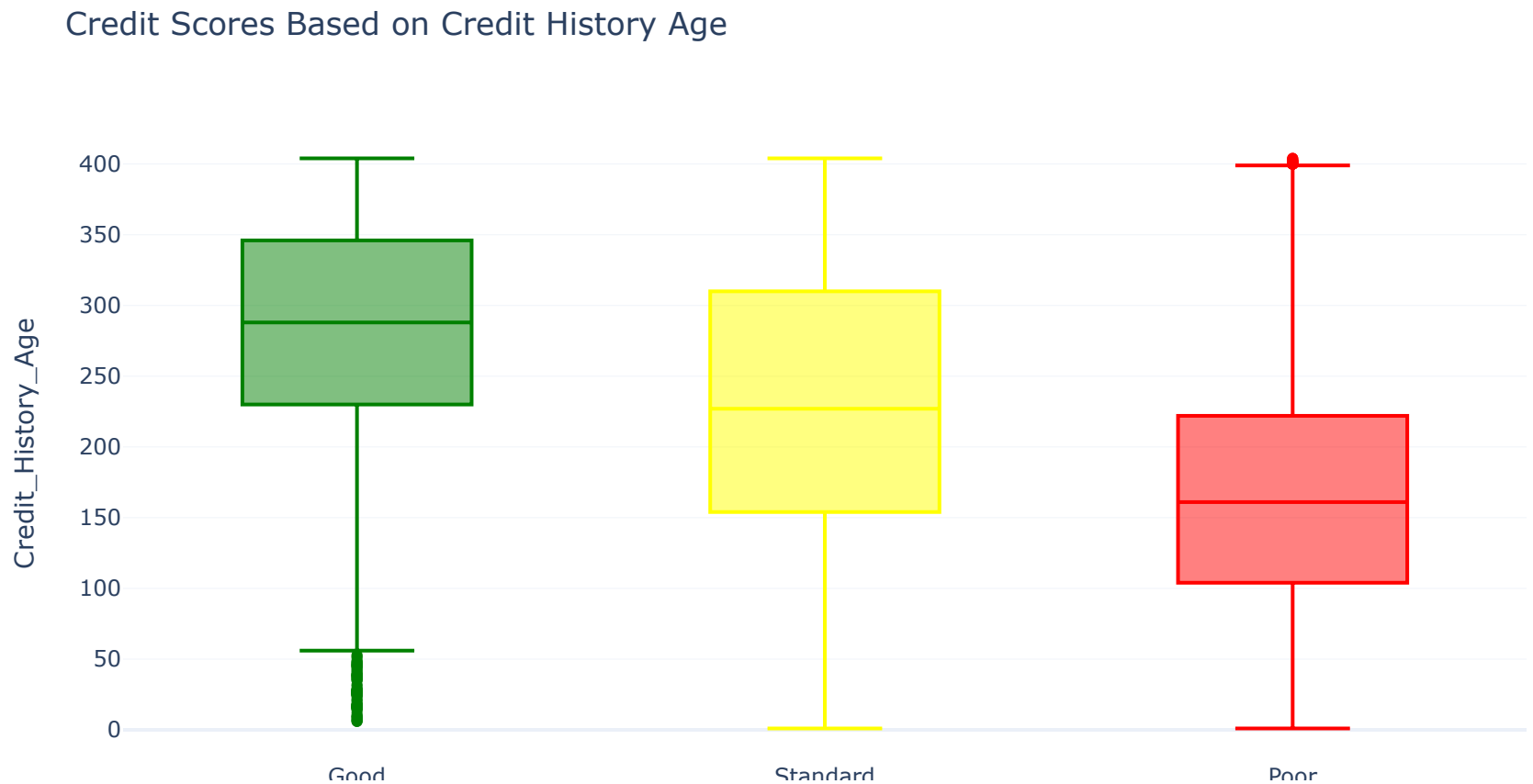


```
In [16]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Credit_Utilization_Ratio",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Credit Utilization Ratio",  
                    color_discrete_map={'Poor':'red',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

Credit Scores Based on Credit Utilization Ratio

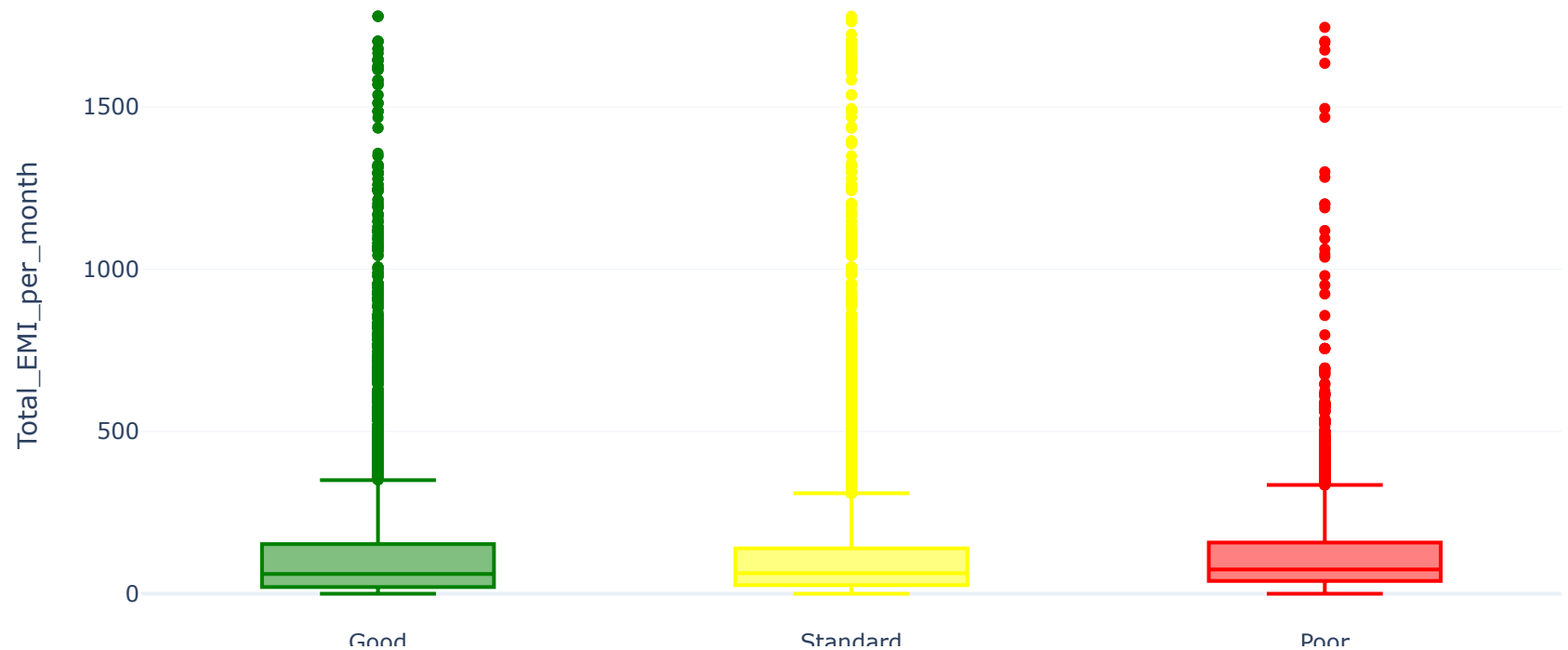



```
In [17]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Credit_History_Age",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Credit History Age",  
                    color_discrete_map={'Poor': 'red',  
                                         'Standard': 'yellow',  
                                         'Good': 'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

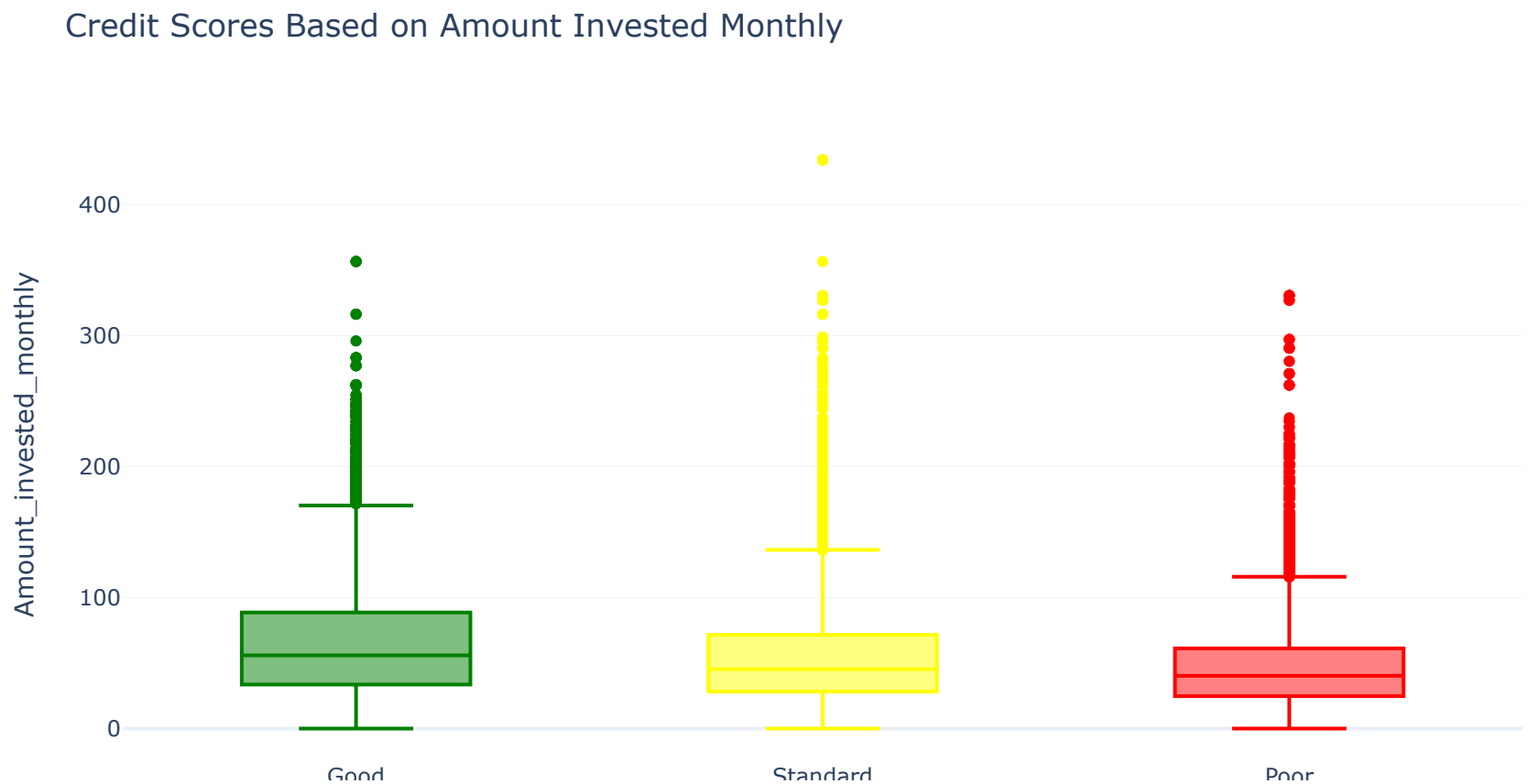


```
In [18]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Total_EMI_per_month",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Total Number of EMIs per Month",  
                    color_discrete_map={'Poor': 'red',  
                                         'Standard': 'yellow',  
                                         'Good': 'green'})  
fig.update_traces(quartilemethod="exclusive")  
fig.show()
```

Credit Scores Based on Total Number of EMIs per Month

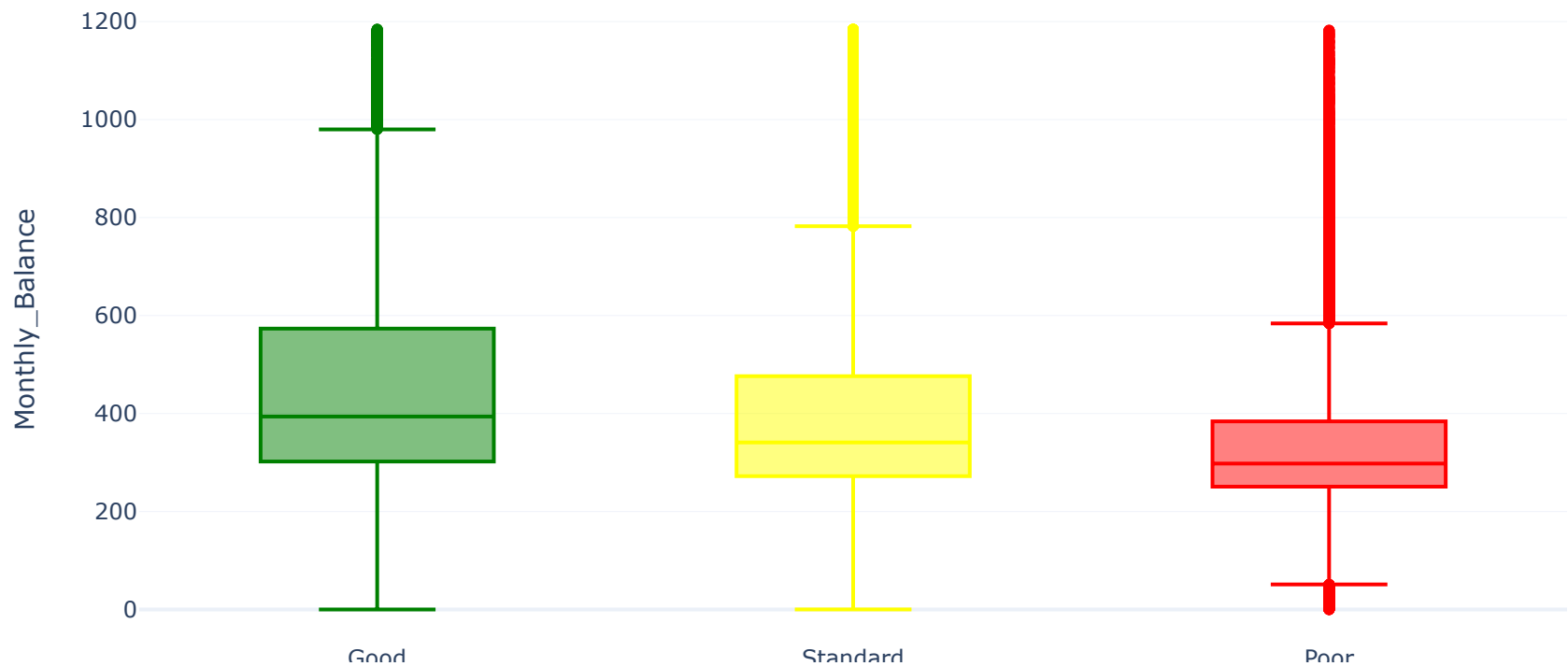


```
In [19]: fig = px.box(data,  
                    x="Credit_Score",  
                    y="Amount_invested_monthly",  
                    color="Credit_Score",  
                    title="Credit Scores Based on Amount Invested Monthly",  
                    color_discrete_map={'Poor':'red',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



```
In [20]: 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 [21]: data["Credit_Mix"] = data["Credit_Mix"].map({'Standard':1,  
                                                    'Good':2,  
                                                    "Bad":0})
```

```
In [22]: 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"]])  
y = np.ravel(y)
```

```
In [23]: 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)
```

```
Out[23]: ▼ RandomForestClassifier  
RandomForestClassifier()
```

```
In [*]: 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))
```

In []:

In []: