

```
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("/content/sample_data/train.csv")
print(data.head())
```

	ID	Customer_ID	Month	Name	Age	SSN	Occupation	Annual_Income	\
0	5634	3392	1	SIDHU	32	821000265	Scientist	19114.12	
1	5635	3392	2	SIDHU	32	821000265	Scientist	19114.12	
2	5636	3392	3	SIDHU	32	821000265	Scientist	19114.12	
3	5637	3392	4	SIDHU	32	821000265	Scientist	19114.12	
4	5638	3392	5	SIDHU	32	821000265	Scientist	19114.12	

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

	Outstanding_Debt	Credit_Utilization_Ratio	Credit_History_Age	\
0	809.98	26.822620	265	
1	809.98	31.944960	266	
2	809.98	28.609352	267	
3	809.98	31.377862	268	
4	809.98	24.797347	269	

	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	284.629163	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]

```
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  int64
5   SSN                                    100000 non-null  int64
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  int64
10  Num_Credit_Card                       100000 non-null  int64
11  Interest_Rate                         100000 non-null  int64
12  Num_of_Loan                           100000 non-null  int64
13  Type_of_Loan                           100000 non-null  object
14  Delay_from_due_date                    100000 non-null  int64
15  Num_of_Delayed_Payment                 100000 non-null  int64
16  Changed_Credit_Limit                   100000 non-null  float64
17  Num_Credit_Inquiries                   100000 non-null  int64
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  int64
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(8), int64(13), object(7)
memory usage: 21.4+ MB
None
```

```
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
```

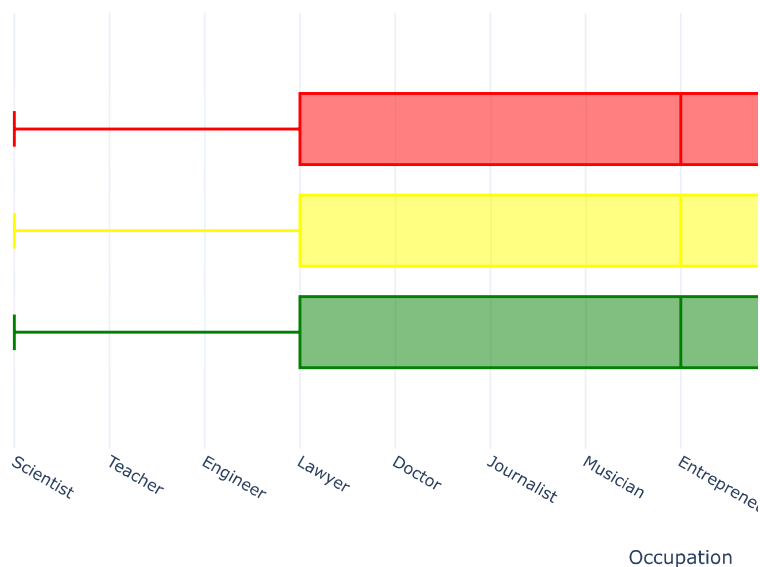
```
data["Credit_Score"].value_counts()
```

```
Standard    53174
Poor         28998
Good         17828
Name: Credit_Score, dtype: int64
```

```
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



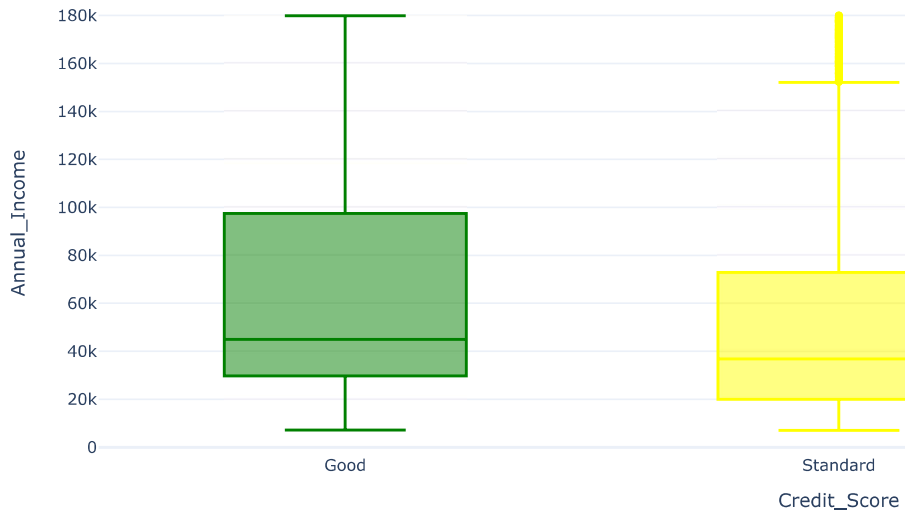
```
fig = px.box(data,
              x="Credit_Score",
              y="Annual_Income",
              color="Credit_Score",
```

```

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

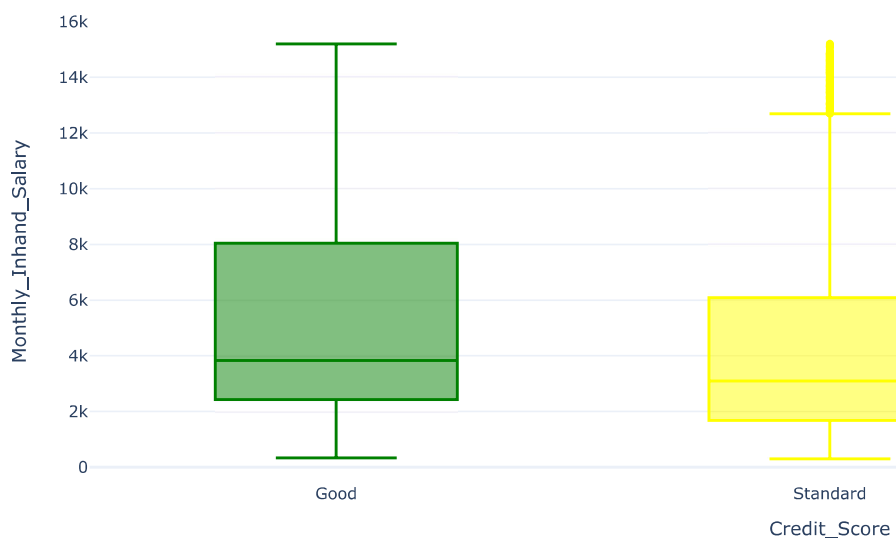


```

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

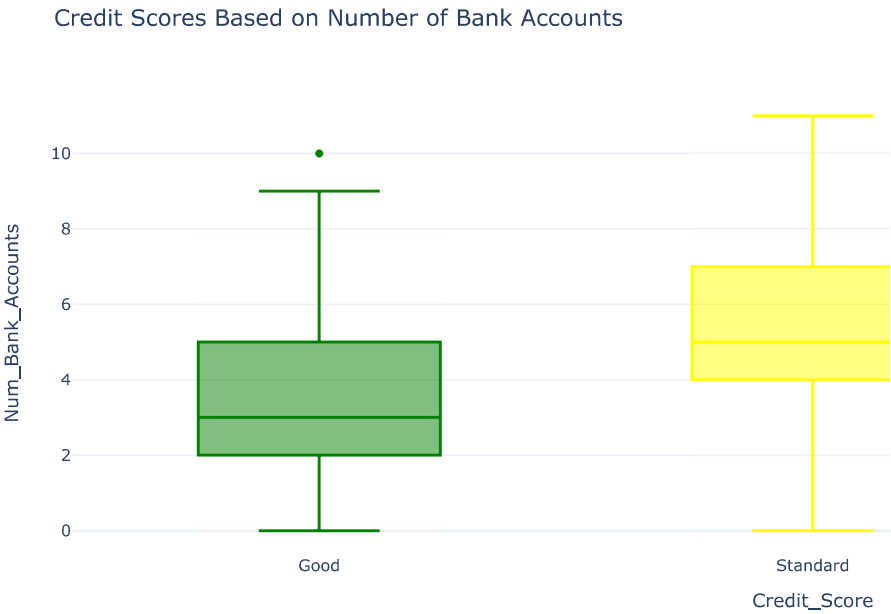


```

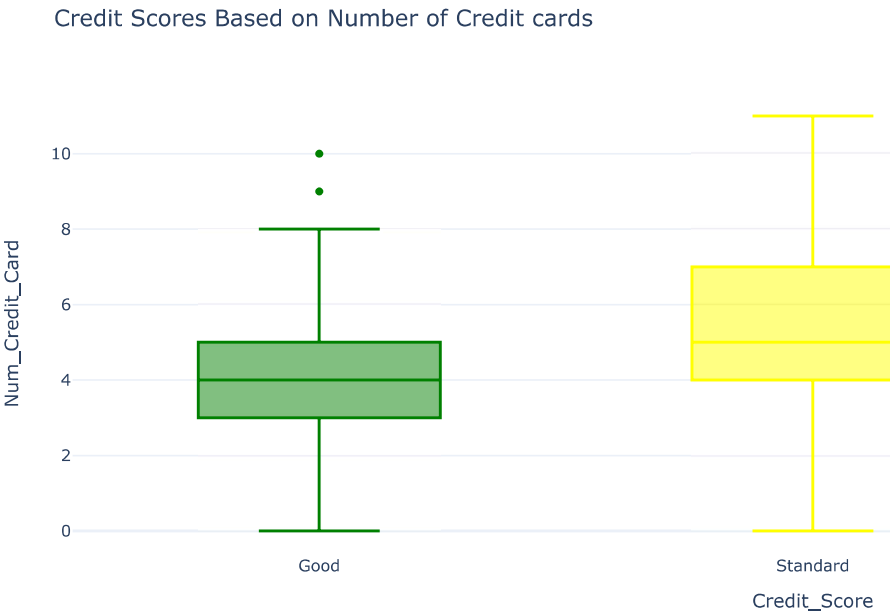
fig = px.box(data,
             x="Credit_Score",
             y="Num_Bank_Accounts",
             color="Credit_Score",
             title="Credit Scores Based on Number of Bank Accounts",
             color_discrete_map={'Poor':'red',
                                 'Standard':'yellow',
                                 'Good':'green'})

```

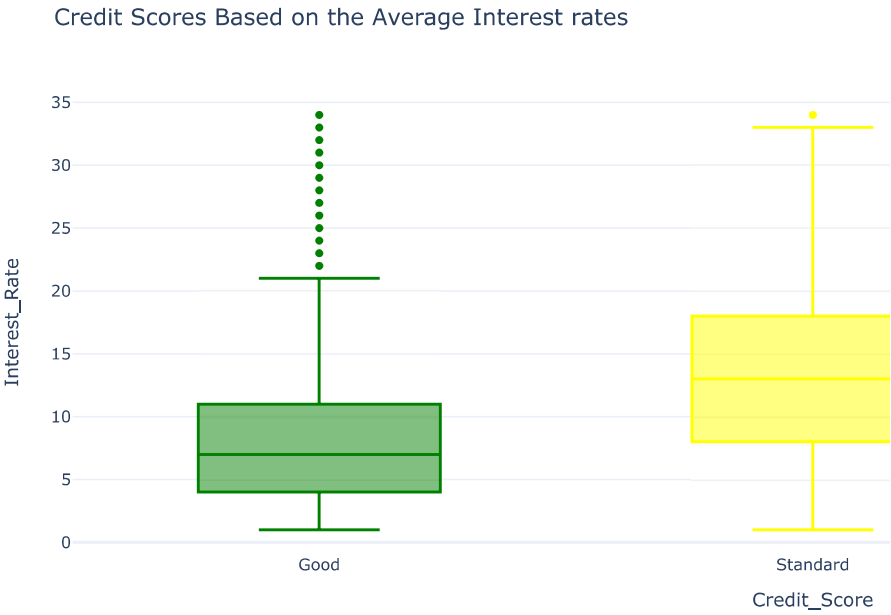
```
'Standard':'yellow',
'Good':'green'})
fig.update_traces(quartilemethod="exclusive")
fig.show()
```



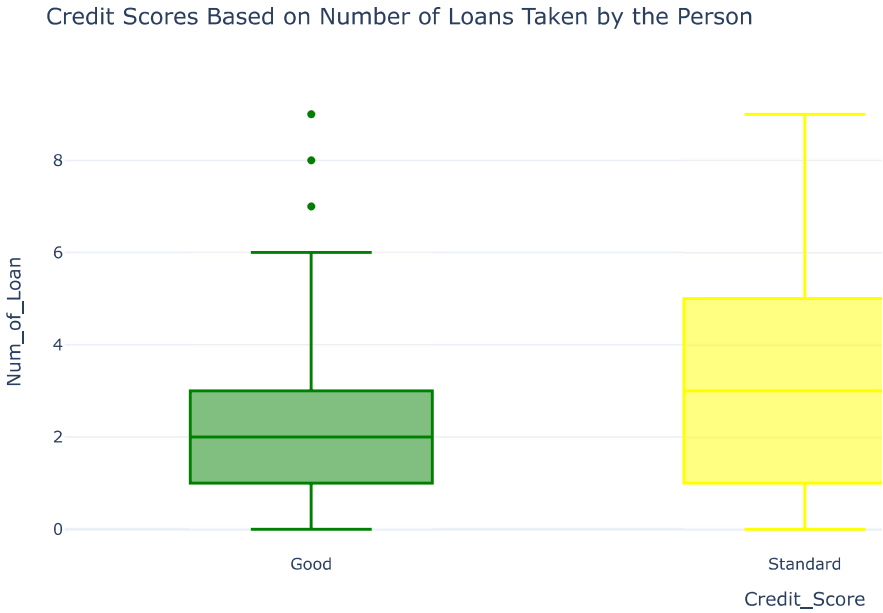
```
fig = px.box(data,
              x="Credit_Score",
              y="Num_Credit_Card",
              color="Credit_Score",
              title="Credit Scores Based on Number of Credit cards",
              color_discrete_map={'Poor':'red',
                                  'Standard':'yellow',
  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



```
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 = 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()
```



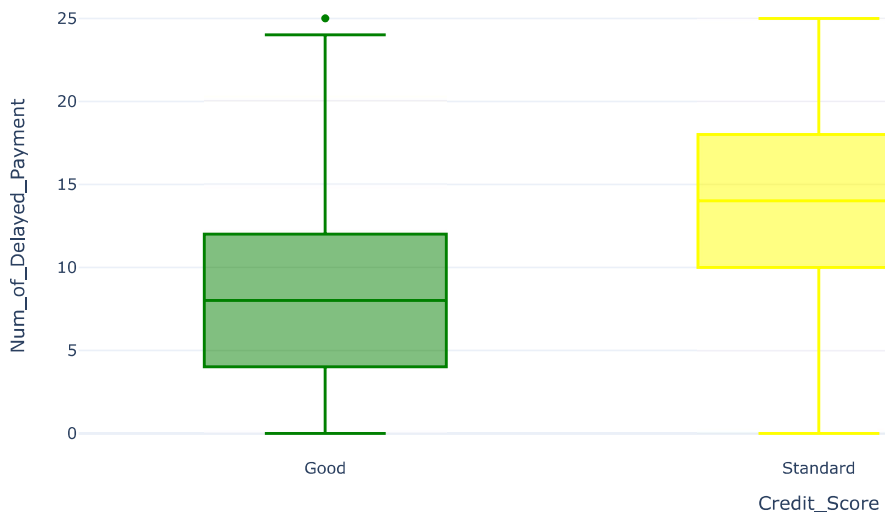
```
fig = px.box(data,
             x="Credit_Score",
             y="Delay_from_due_date",
             color="Credit_Score",
             title="Credit Scores Based on Average Number of Days Delayed for Credit card Payments",
             color_discrete_map={'Poor':'red',
                         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

Credit Scores Based on Average Number of Days Delayed for Credit card



```
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



```
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

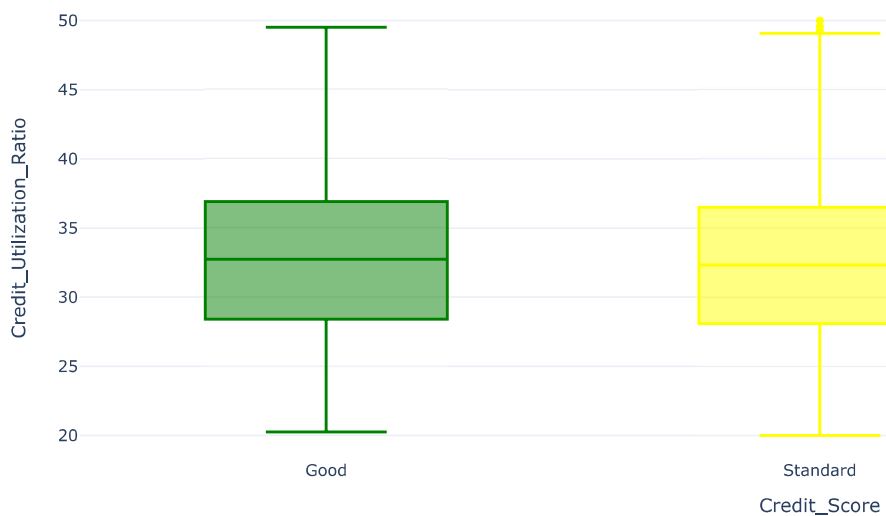


```

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



```

fig = px.box(data,
x="Credit_Score",
y="Credit_History_Age",
color="Credit_Score",

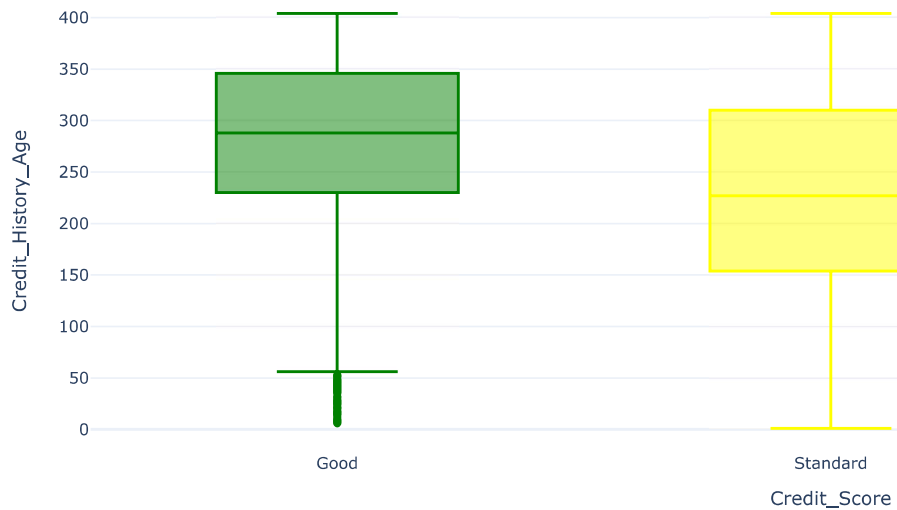
```

```

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

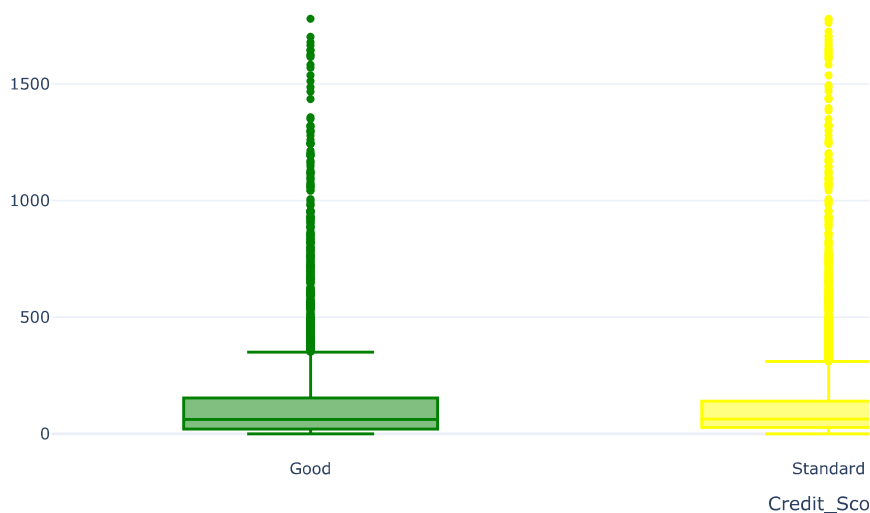


```

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



```

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',

```

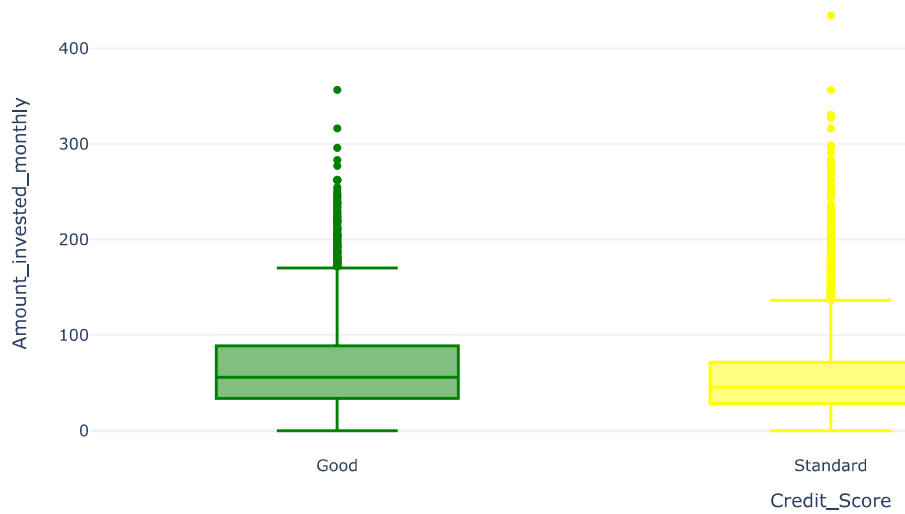


```

        'Standard': 'yellow',
        'Good': 'green'})
fig.update_traces(quartilemethod="exclusive")
fig.show()

```

Credit Scores Based on Amount Invested Monthly

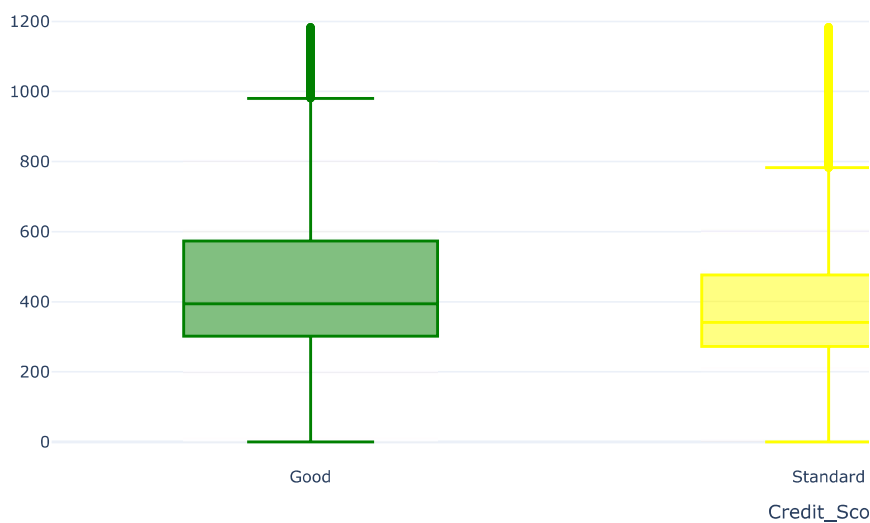


```

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



```

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

```

```

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']])

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)

<ipython-input-68-ea8253e80d41>:6: DataConversionWarning:
A column-vector y was passed when a 1d array was expected. Please change the shape of y to

  ▾ RandomForestClassifier
  RandomForestClassifier()

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]])

Credit Score Prediction :
Annual Income: 120000
Monthly Inhand Salary: 12000
Number of Bank Accounts: 2
Number of Credit cards: 5
Interest rate: 14
Number of Loans: 5
Average number of days delayed by the person: 5
Number of delayed payments: 4
Credit Mix (Bad: 0, Standard: 1, Good: 3) : 1
Outstanding Debt: 580000
Credit History Age: 5
Monthly Balance: 400

print("Predicted Credit Score = ", model.predict(features))

Predicted Credit Score =  ['Standard']

```

✓ 0s completed at 12:28 PM

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