

**Objective:** To explore various AutoEDA capabilities and perform analysis on a given dataset

This notebook will focus on DataPrep

## 2. AutoEDA - DataPrep

**Dataset Reference:** Loan Prediction dataset from Kaggle

**Features:**

- General Overview - Quick insights of all variables in the dataset using the plot dataframe.
- Details about each variables / features in the dataset by using create\_report - overview, variables, interactions, correlations, missing values
- Interactions - based on x-axis and y-axis scatter plots
- Correlations between variables - Pearson's Correlation Coefficient, Spearman's Rank Correlation Coefficient, Kendall's Rank Correlation Coefficient
- Missing Values - Bar chart, Spectrum, Heatmap, Dendogram representations
- We can pick one particular feature and analyze - Stats, Bar chart, Pie chart, Word Count, Word Frequency etc as per applicability

**When To Use?**

- Dataset size is fairly very large (this seems to be 10X faster than Pandas Profiling tools due to it's highly optimized Dask-based computing module)
- Need some quick insights about an unknown dataset
- Use this as a basis for your further EDA analysis on top of it

```
In [36]: import pandas as pd
import warnings

warnings.filterwarnings("ignore")
```

```
In [37]: !pip --disable-pip-version-check install dataprep # Please use it for the first time if it is not installed in your environment
Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js
```

```
In [38]: from dataprep.eda import create_report, plot, plot_correlation, plot_missing, plot_diff
```

```
In [39]: df_train = pd.read_csv("../input/loan-eligible-dataset/loan-train.csv")

df_train.head()
```

```
Out[39]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_His
0	LP001002	Male	No	0	Graduate	No	5849	0.0	NaN	360.0	
1	LP001003	Male	Yes	1	Graduate	No	4583	1508.0	128.0	360.0	
2	LP001005	Male	Yes	0	Graduate	Yes	3000	0.0	66.0	360.0	
3	LP001006	Male	Yes	0	Not Graduate	No	2583	2358.0	120.0	360.0	
4	LP001008	Male	No	0	Graduate	No	6000	0.0	141.0	360.0	

```
In [40]: df_test = pd.read_csv("../input/loan-eligible-dataset/loan-test.csv")

df_test.head()
```

```
Out[40]:
```

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_His
0	LP001015	Male	Yes	0	Graduate	No	5720	0	110.0	360.0	
1	LP001022	Male	Yes	1	Graduate	No	3076	1500	126.0	360.0	
2	LP001031	Male	Yes	2	Graduate	No	5000	1800	208.0	360.0	
3	LP001035	Male	Yes	2	Graduate	No	2340	2546	100.0	360.0	
4	LP001051	Male	No	0	Not Graduate	No	3276	0	78.0	360.0	

```
In [41]: df_train.shape
```

```
Out[41]: (614, 13)
```

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```
In [42]: df_test.shape
```

Out[42]: (367, 12)

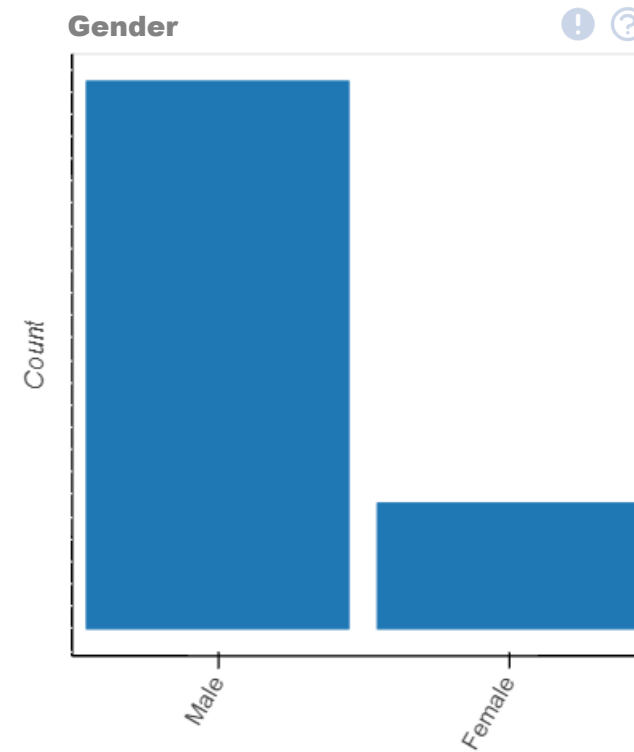
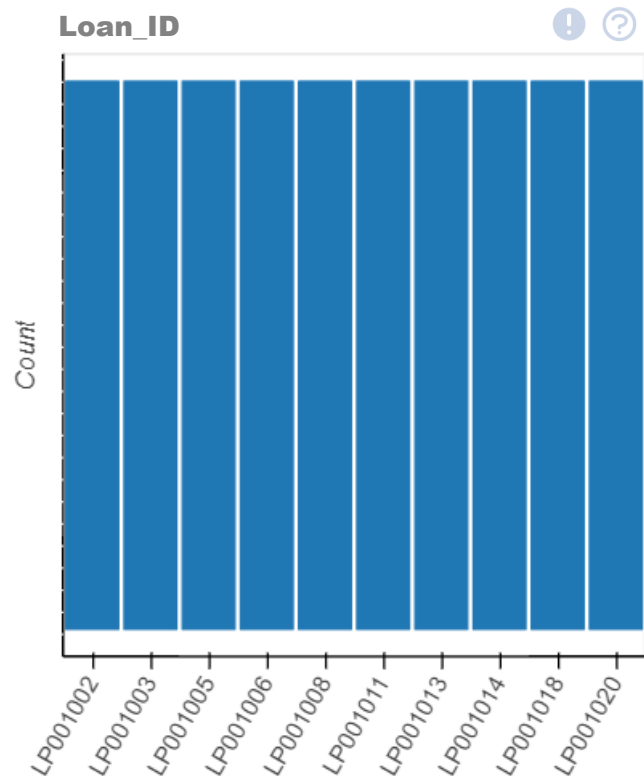
## 2.1 Analyze distributions

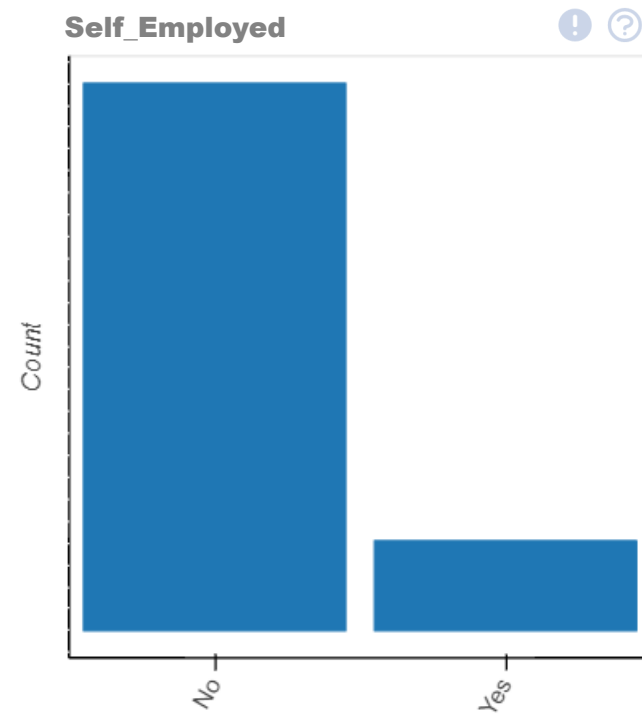
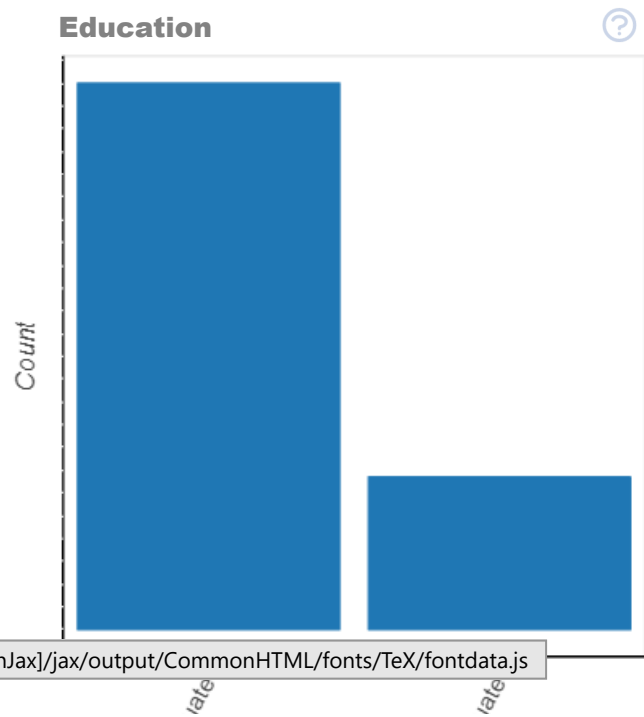
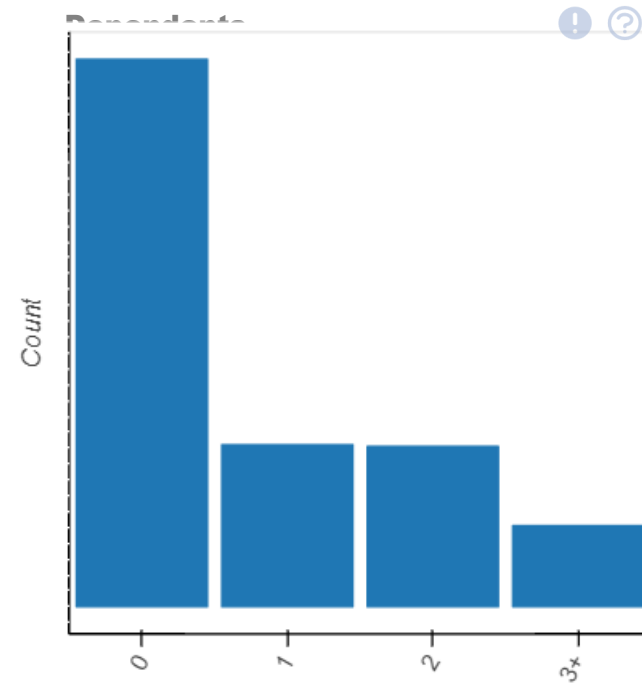
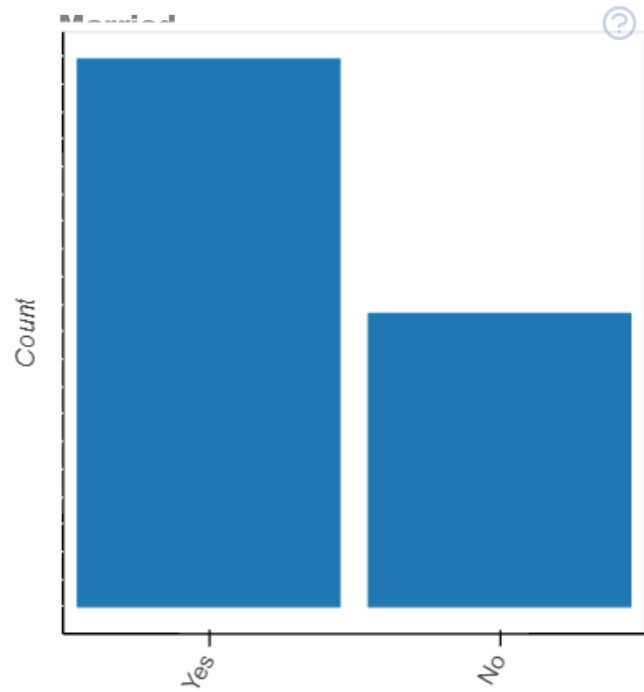
- `plot(df)`: plots the distribution of each column and computes dataset statistics
- `plot(df, col1)`: plots the distribution of column `col1` in various ways, and computes its statistics
- `plot(df, col1, col2)`: generates plots depicting the relationship between columns `col1` and `col2`

In [44]: `plot(df_train)`

Out[44]:

Show Stats and Insights

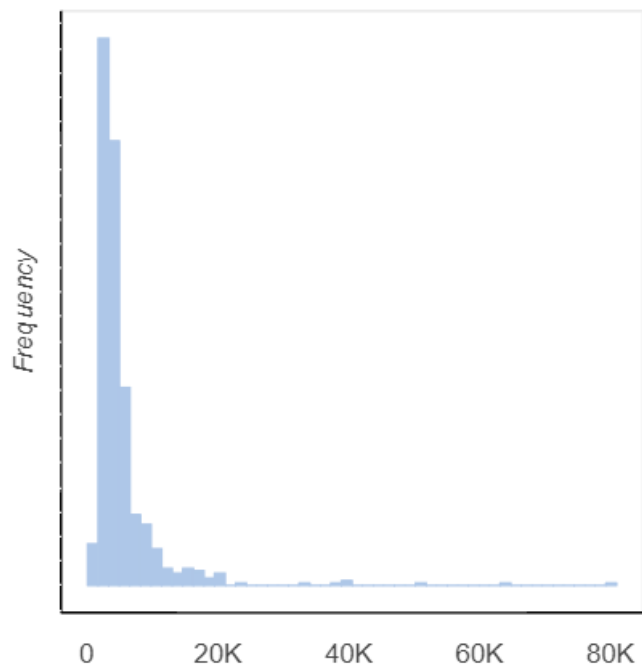




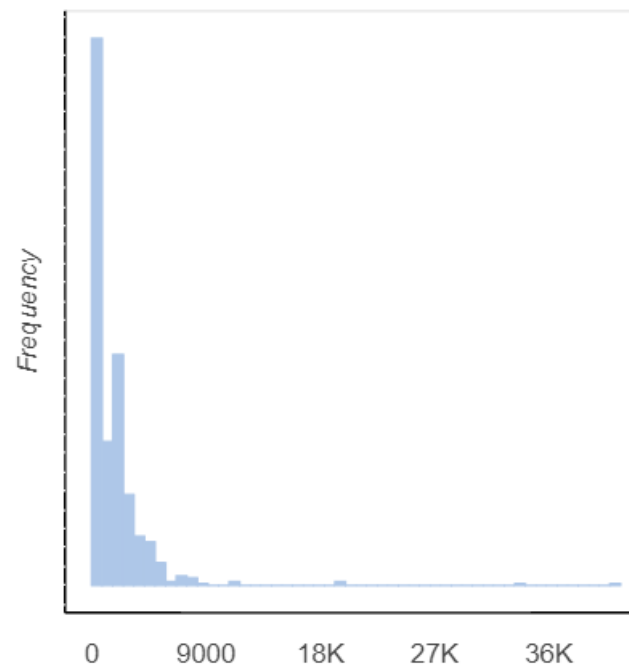
Gradu

Not Gradu

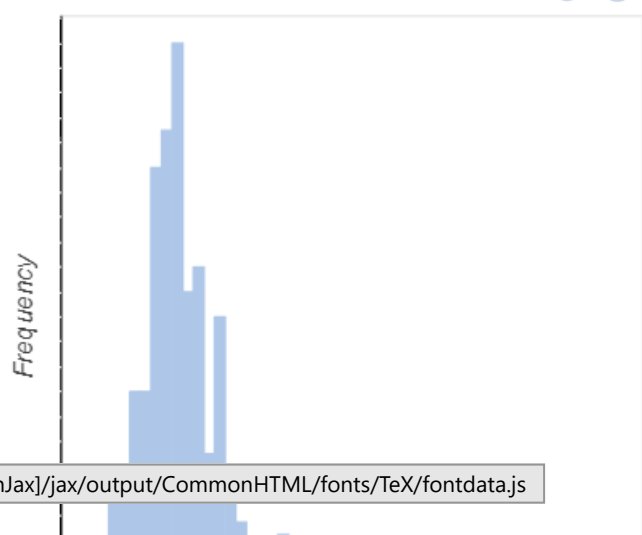
ApplicantIncome



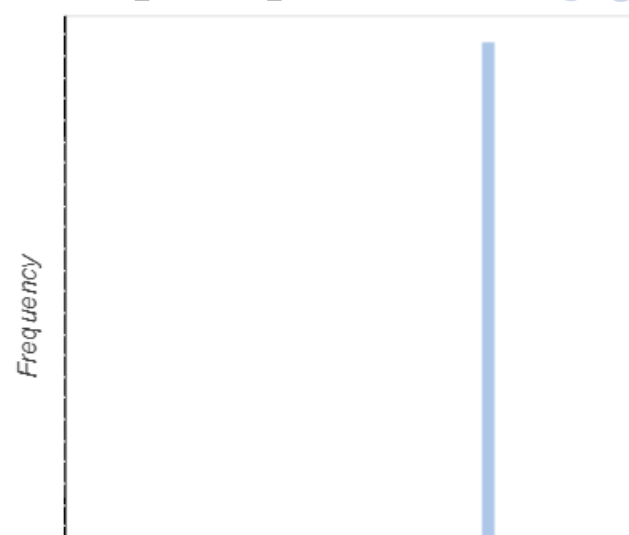
CoapplicantIncome

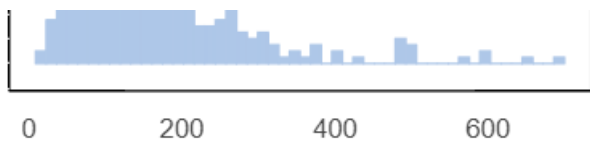


LoanAmount

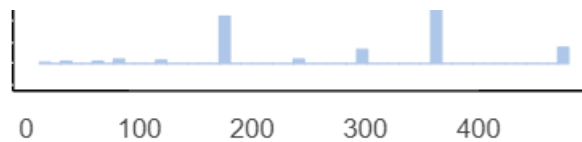
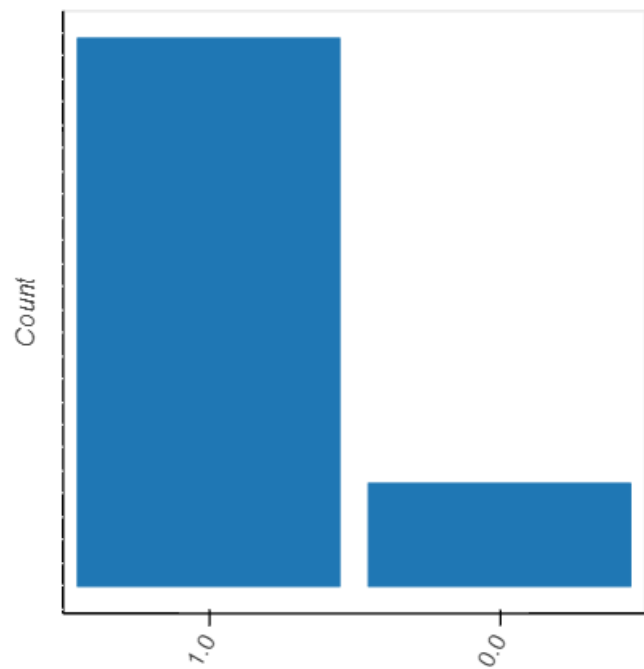


Loan\_Amount\_Term

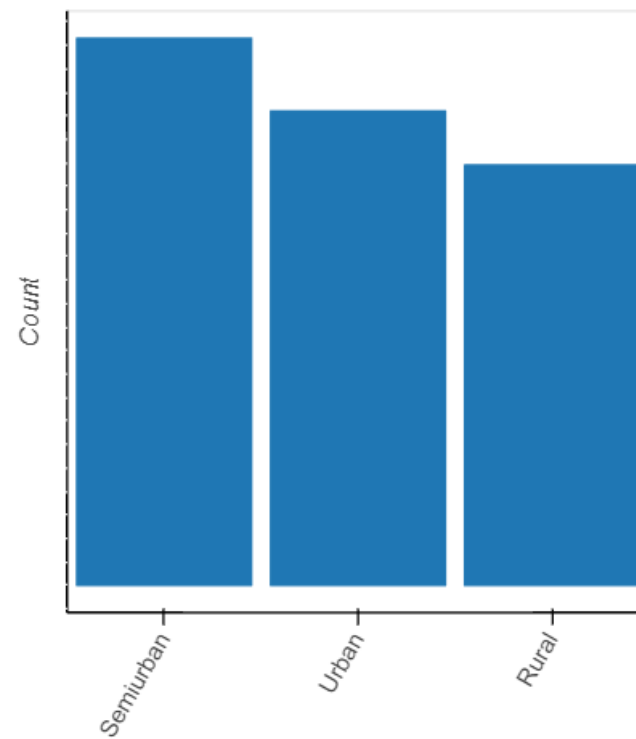




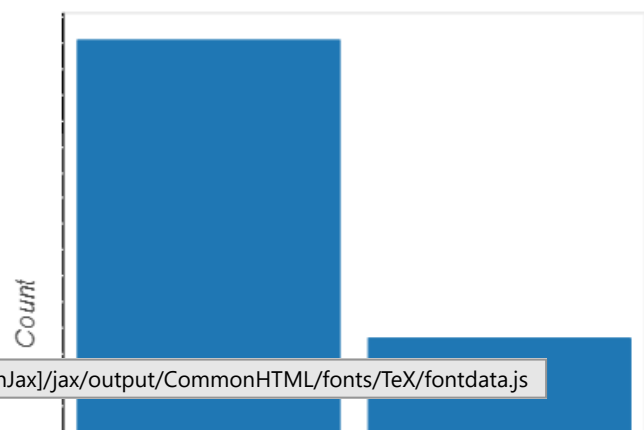
**Credit\_History**

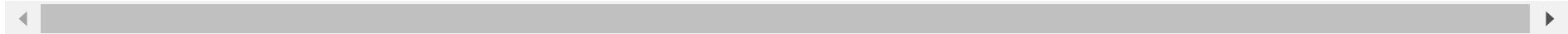


**Property\_Area**



**Loan\_Status**





```
In [45]: # plots the distribution of column x in various ways and calculates column statistics

plot(df_train, "Property_Area")
```

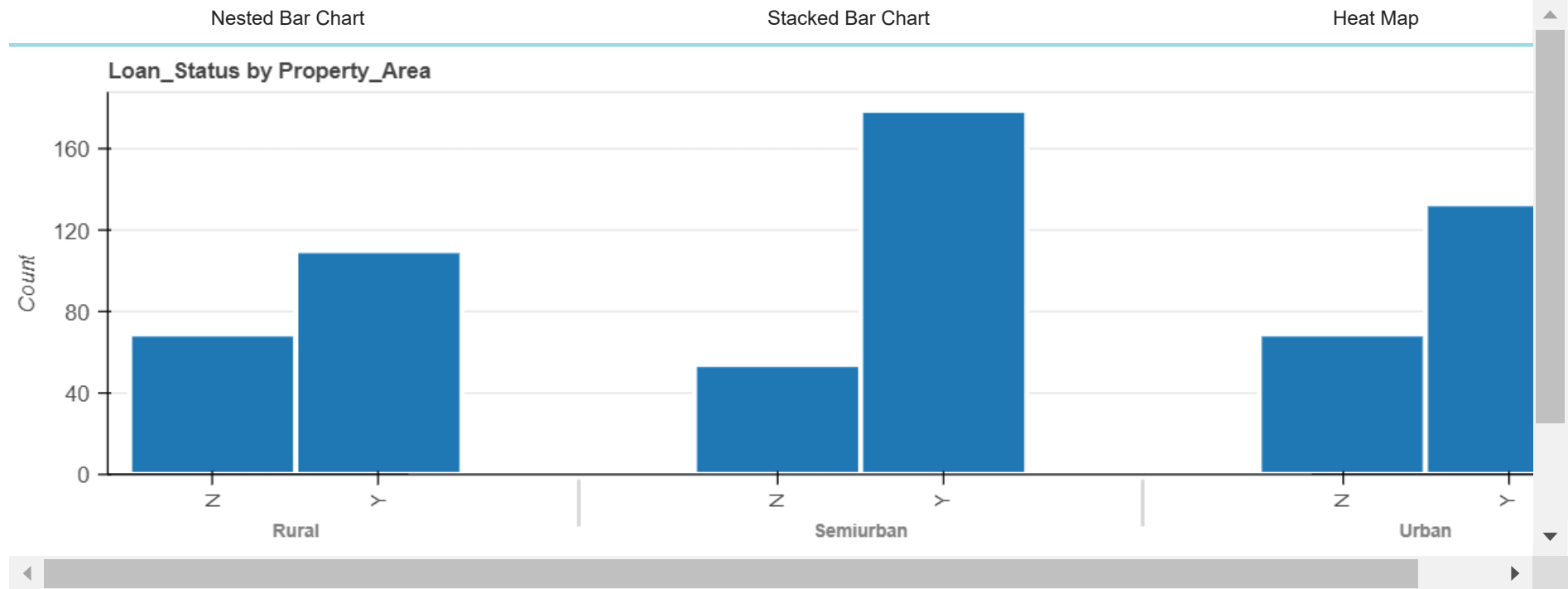
Out[45]: Stats Bar Chart Pie Chart Word Cloud Word Frequency Word Length Value Table

Overview			Sample	
Approximate Distinct Count	3		1st row	Urban
Approximate Unique (%)	0.5%		2nd row	Rural
Missing	0		3rd row	Urban
Missing (%)	0.0%		4th row	Urban
Memory Size	42.9 KB		5th row	Urban
Length			Letter	
Mean	6.5179		Count	4002
Standard Deviation	1.9426		Lowercase Letter	3388
Median	5		Space Separator	0
Minimum	5		Uppercase Letter	614
Maximum	9		Dash Punctuation	0
			Decimal Number	0

```
In [47]: # generates plots depicting the relationship between columns
```

```
plot(df_train, "Property_Area", "Loan_Status")
```

Out[47]:



## 2.2 Analyze correlations

- `plot_correlation(df)`: plots correlation matrices (correlations between all pairs of columns)
- `plot_correlation(df, col1)`: plots the most correlated columns to column `col1`
- `plot_correlation(df, col1, col2)`: plots the joint distribution of column `col1` and column `col2` and computes a regression line

In [48]: *# plots correlation matrices (correlations between all pairs of columns)*

```
plot_correlation(df_train)
```

Out[48]:

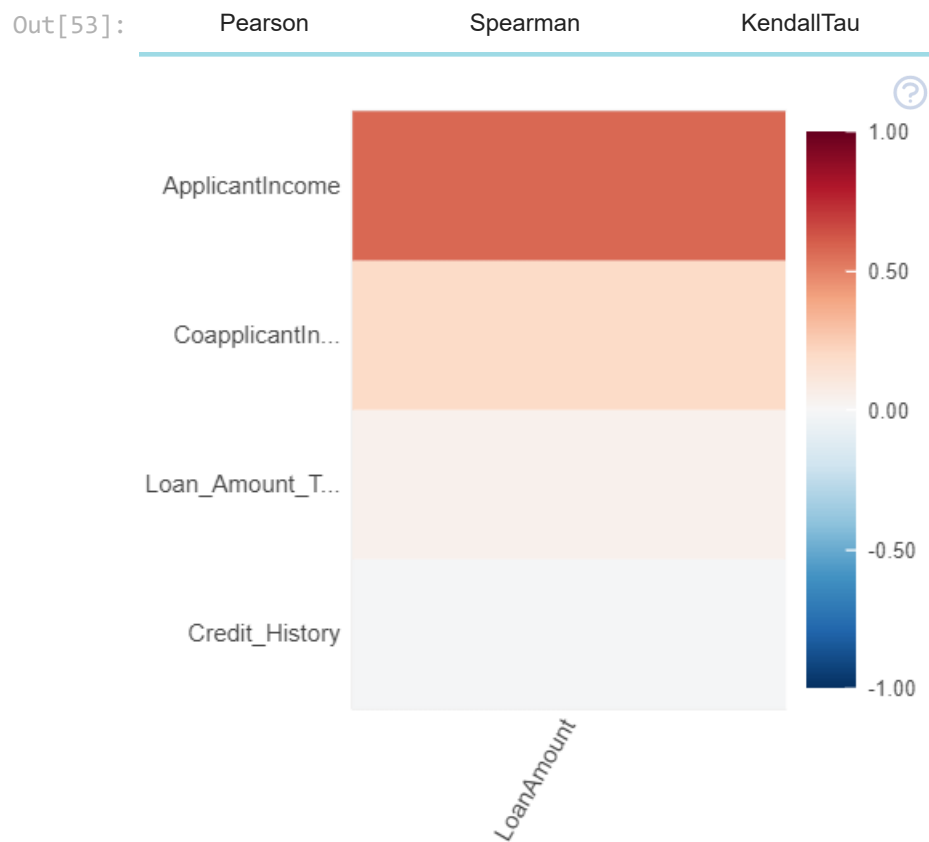
Stats	Pearson	Spearman	KendallTau
	<b>Pearson</b>	<b>Spearman</b>	<b>KendallTau</b>
Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js	0.571	0.515	0.372



	Pearson	Spearman	KendallTau
<b>Highest Negative Correlation</b>	-0.117	-0.32	-0.23
<b>Lowest Correlation</b>	0.001	0.002	0.002
<b>Mean Correlation</b>	0.044	0.038	0.029

```
In [53]: # plots the most correlated columns to column x
# Please ensure x are numerical columns to be analyzed for this

plot_correlation(df_train, "LoanAmount")
```



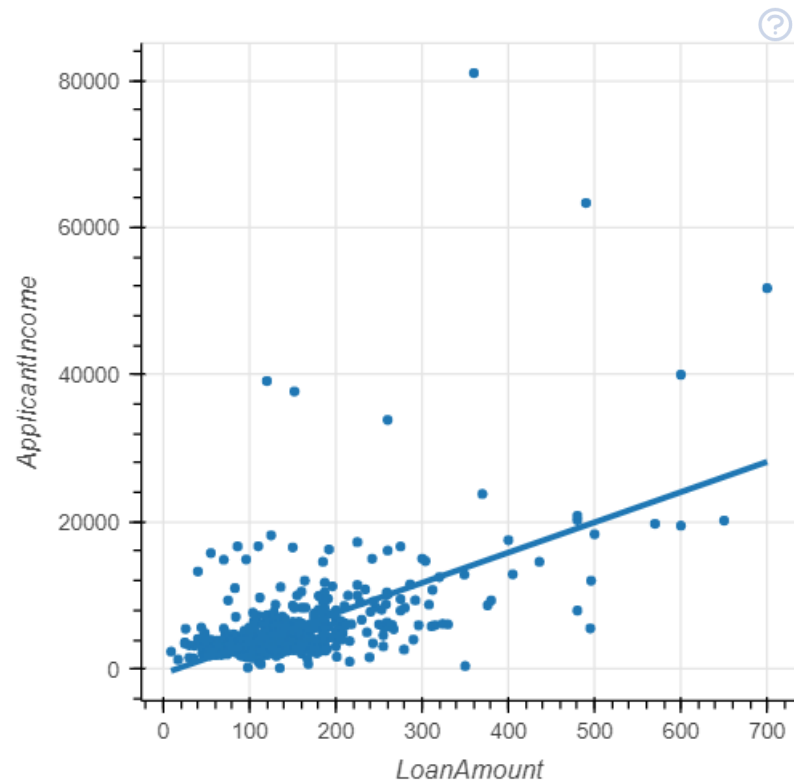
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```
In [54]: # plots the joint distribution of column col1 and column col2 and computes a regression line
```

```
plot_correlation(df_train, "LoanAmount", "ApplicantIncome")
```

Out[54]:

Scatter Plot & Regression Line



## 2.3 Analyze missing values

- `plot_missing(df)`: plots the amount and position of missing values, and their relationship between columns
- `plot_missing(df, col1)`: plots the impact of the missing values in column `col1` on all other columns
- `plot_missing(df, col1, col2)`: plots the impact of the missing values from column `col1` on column `col2` in various ways.

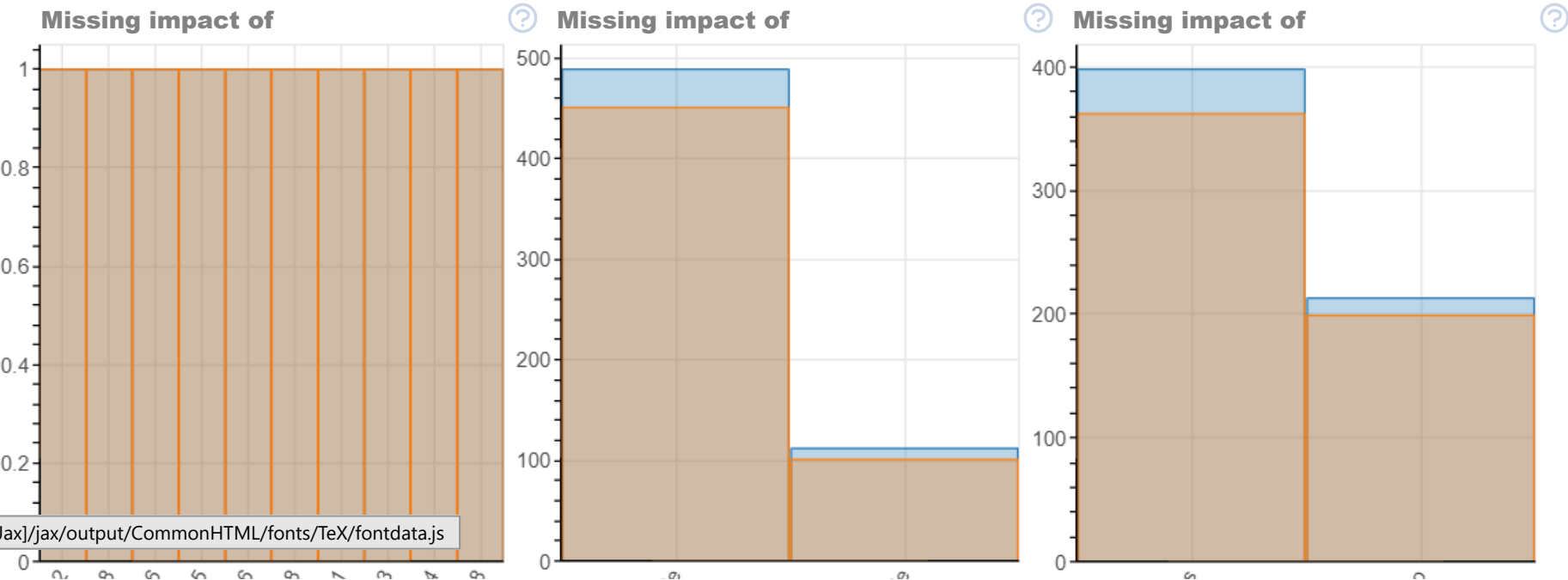
```
In [56]: # plots the amount and position of missing values, and their relationship between columns
```

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Missing Statistics

Missing Cells	149
Missing Cells (%)	1.9%
Missing Columns	7
Missing Rows	134
Avg Missing Cells per Column	11.46
Avg Missing Cells per Row	0.24

```
In [58]: # plots the impact of the missing values in column col1 on all other columns
plot_missing(df_train, "Credit_History")
```



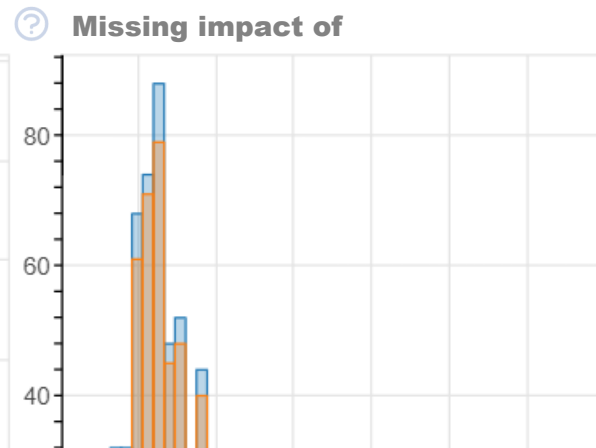
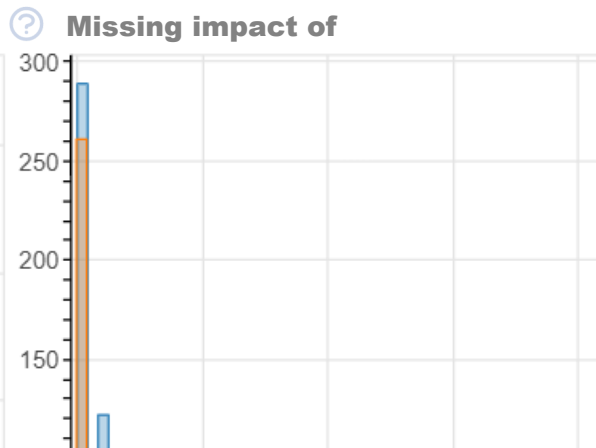
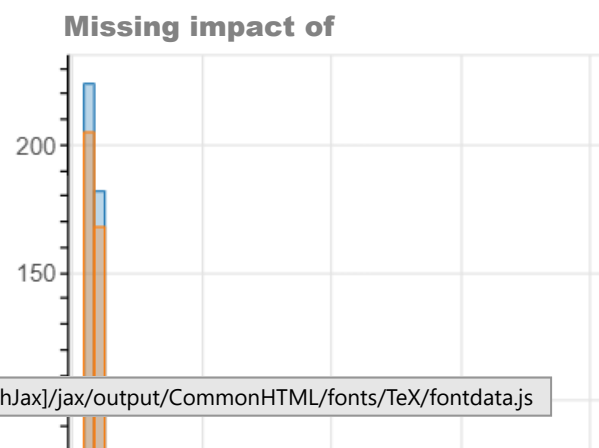
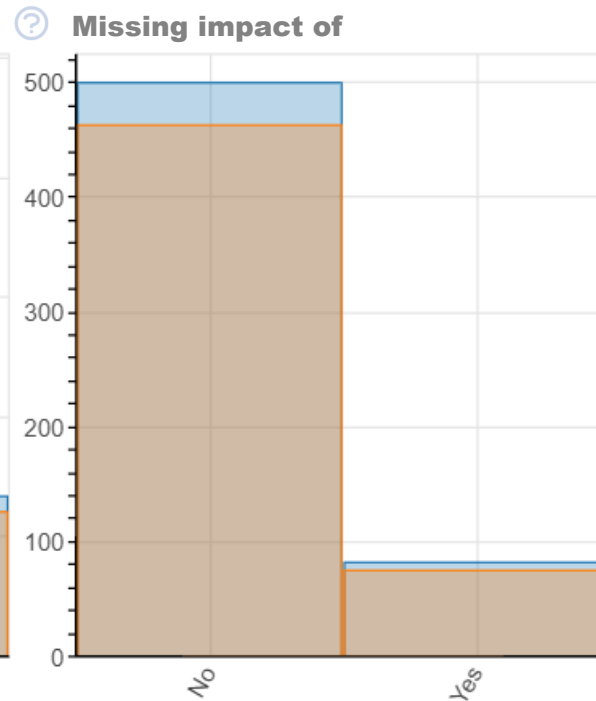
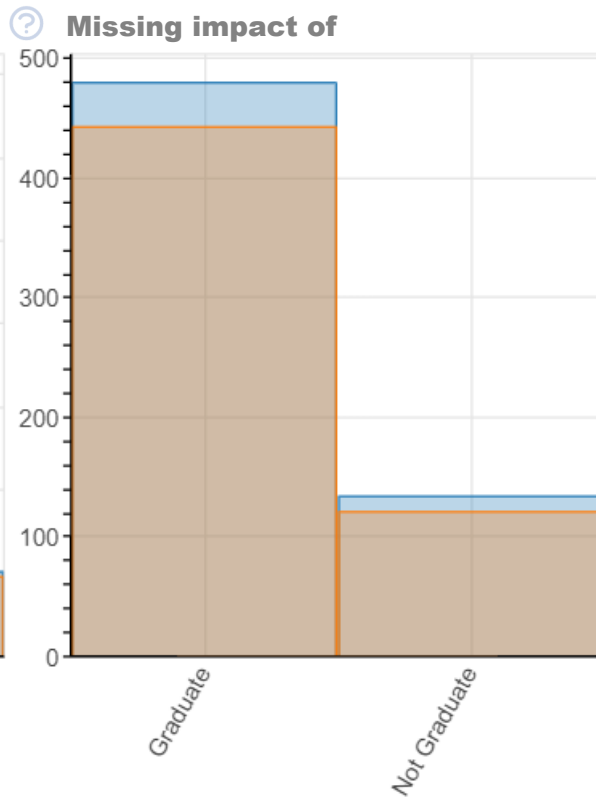
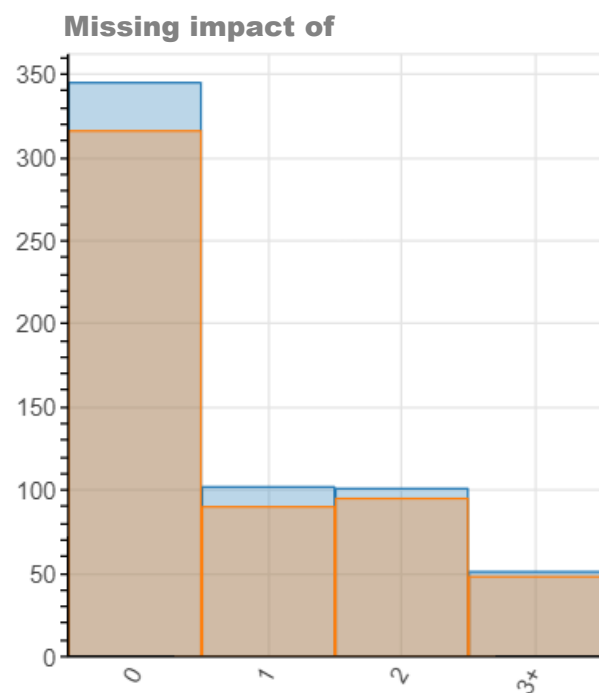
LP001004  
LP001032  
LP001032  
LP001002  
LP001002  
LP001002  
LP001011  
LP001011  
LP001014  
LP001014  
LP001014  
Top 10.0 out of 614.0

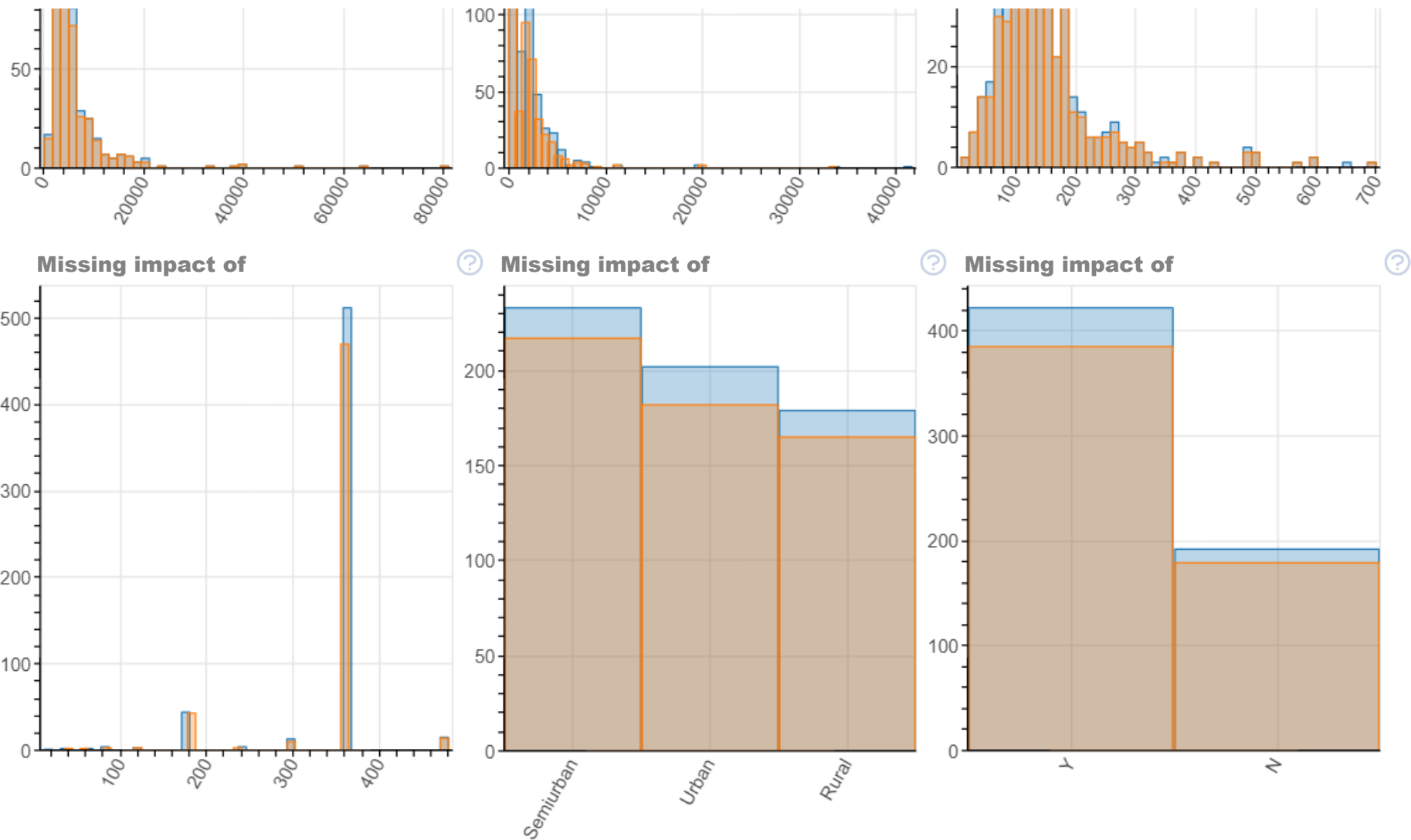
Male

Female

Yes

No

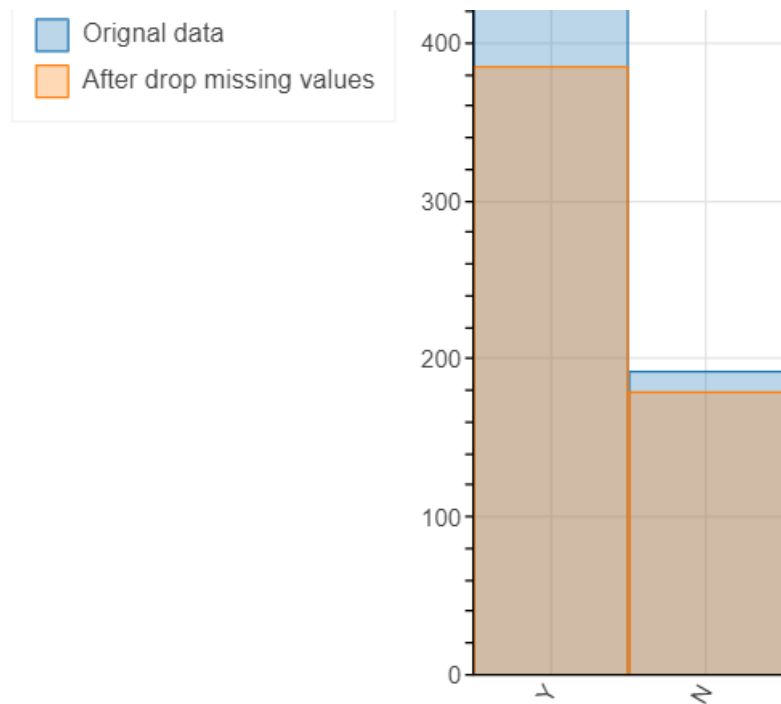




```
In [60]: # plots the impact of the missing values from column col1 on column col2 in various ways
plot_missing(df_train, "Credit_History", "Loan_Status")
```

Out[60]: Missing impact of Credit\_History by Loan\_Status

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## 2.4 Analyze difference between dataframes

- `plot_diff()`: explores the difference of column distributions and statistics across multiple datasets

```
In [61]: # We can analyze differences with plot_diff()
# This is a quick way to get some insights between Train and Test datasets

plot_diff([df_train,df_test])
```

Out[61]:

Show Stats

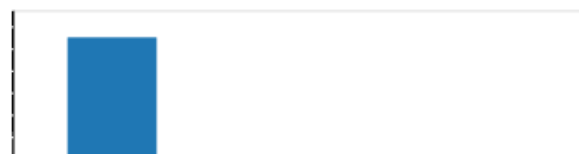
df1

df2

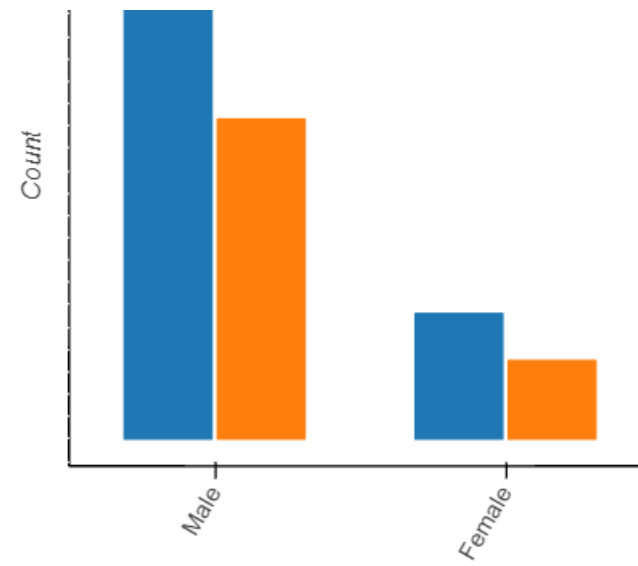
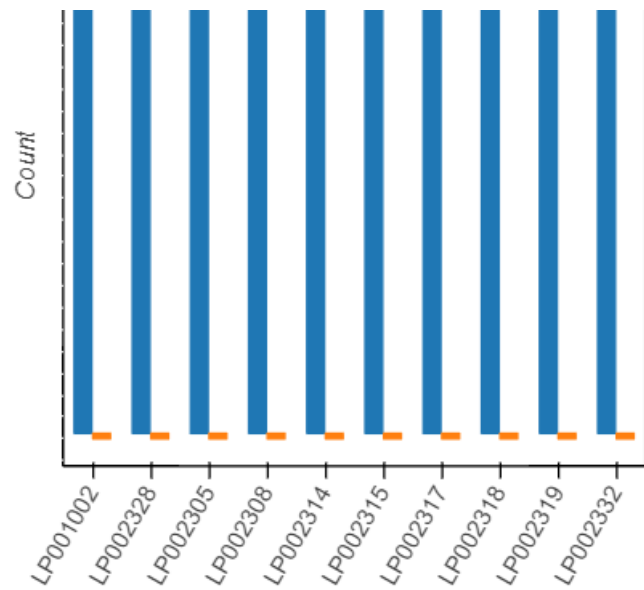
Loan\_ID



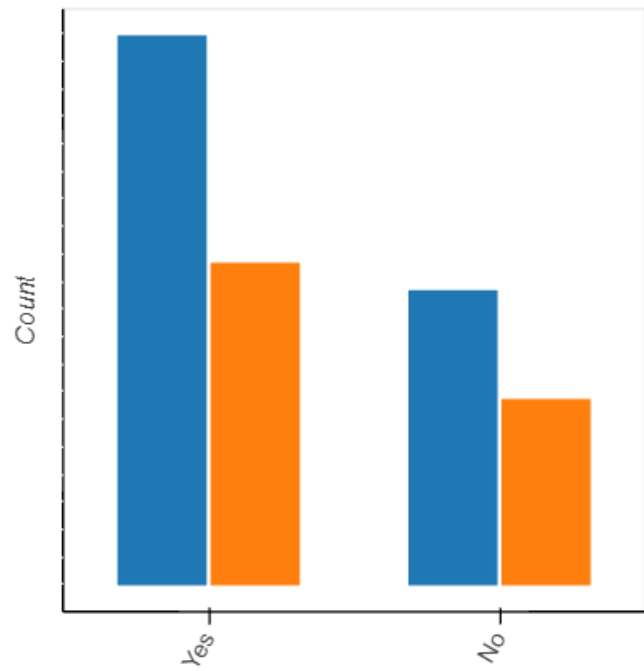
Gender



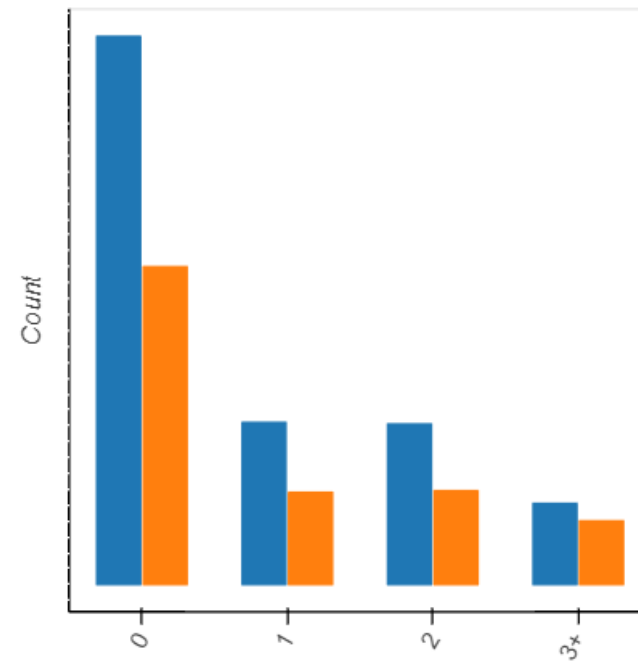
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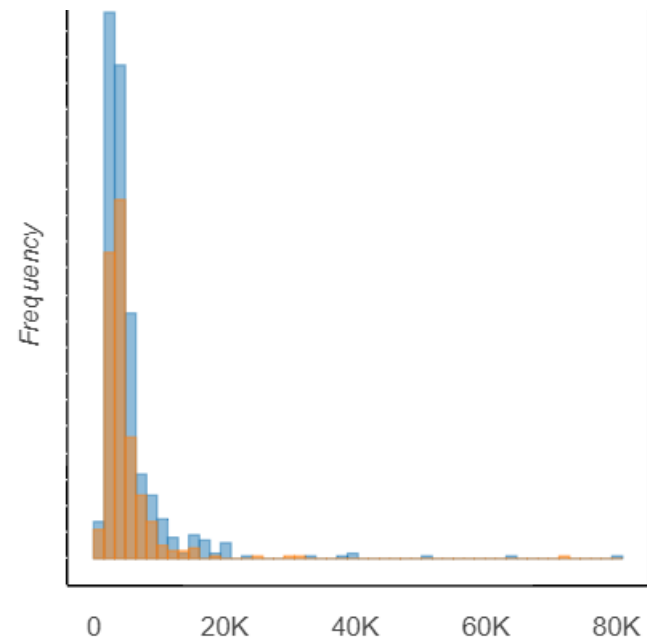
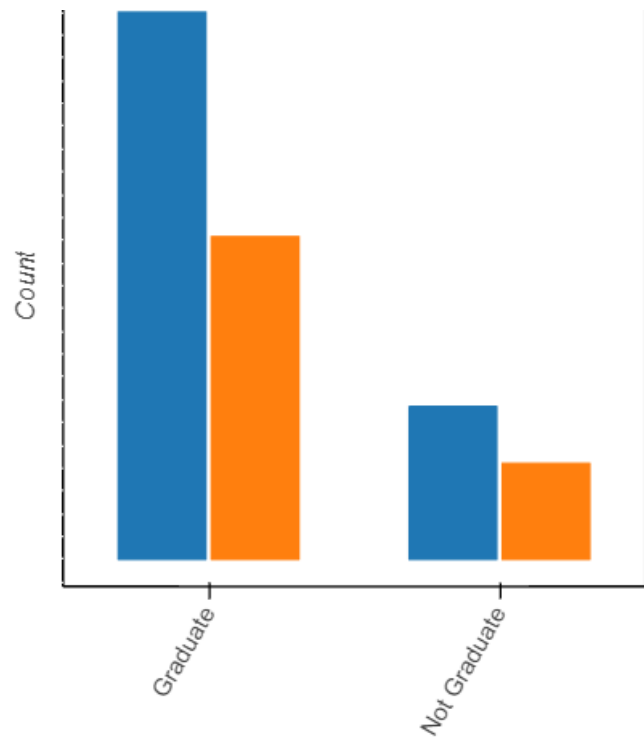
**Married**



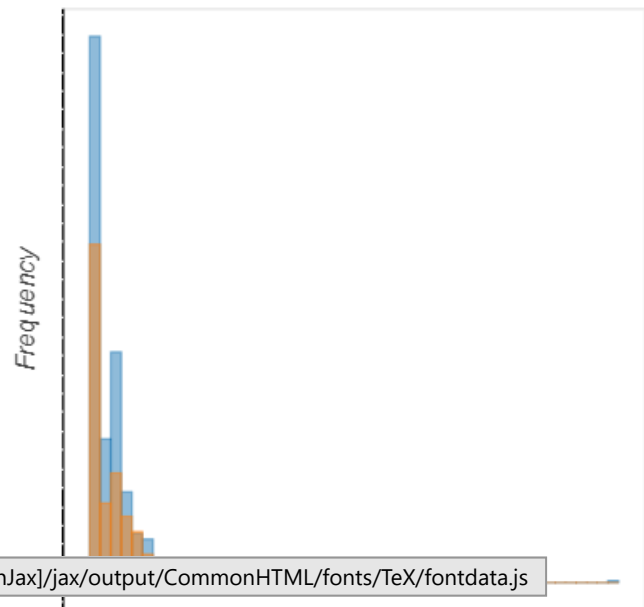
**Dependents**



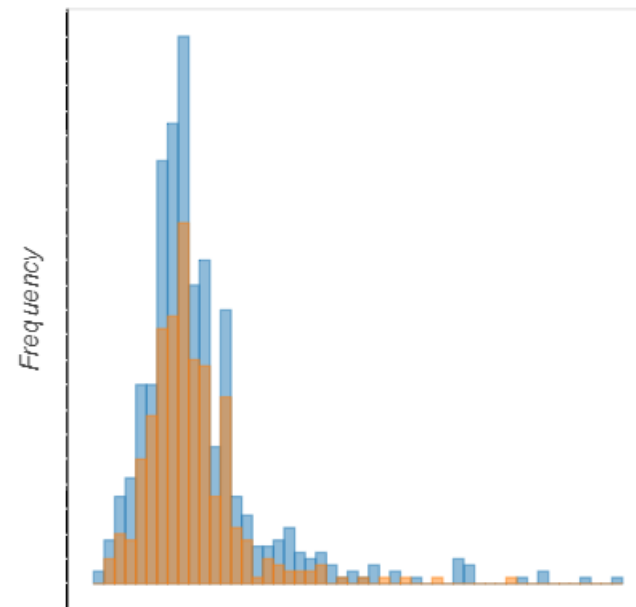
**ApplicantIncome**



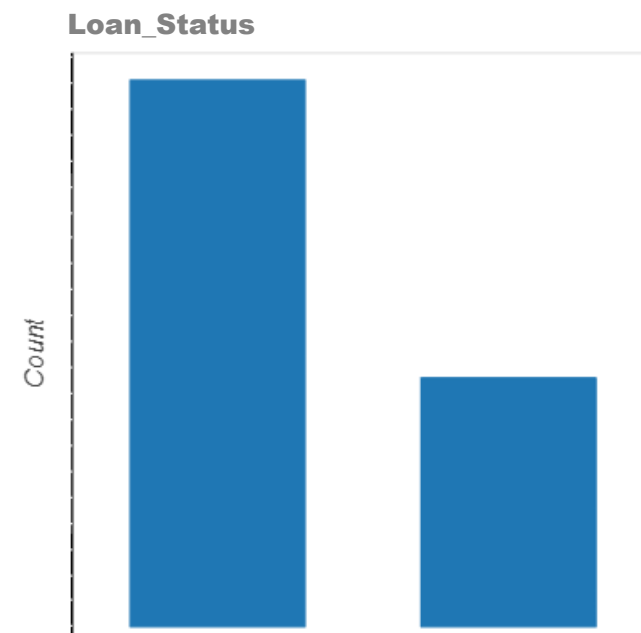
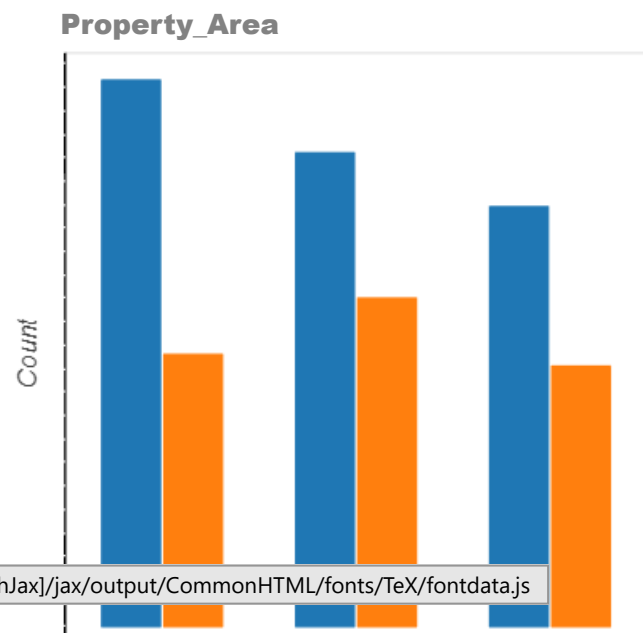
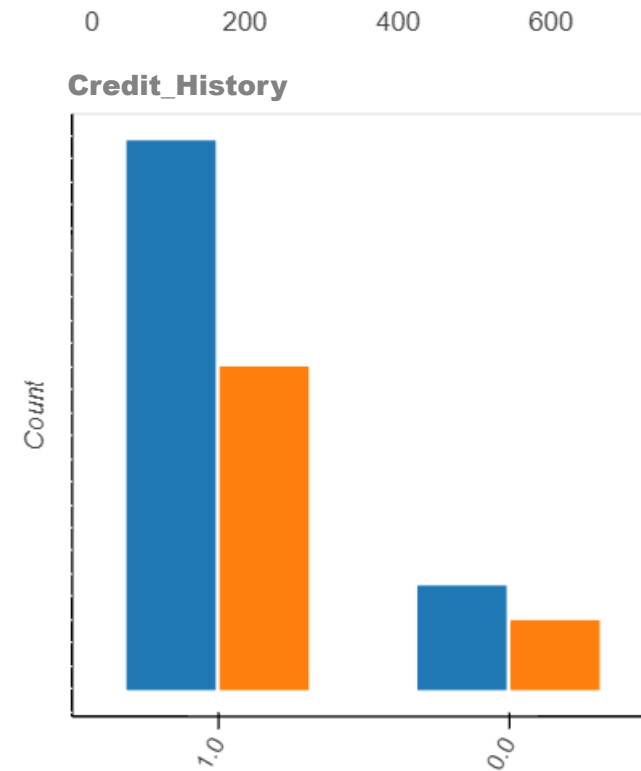
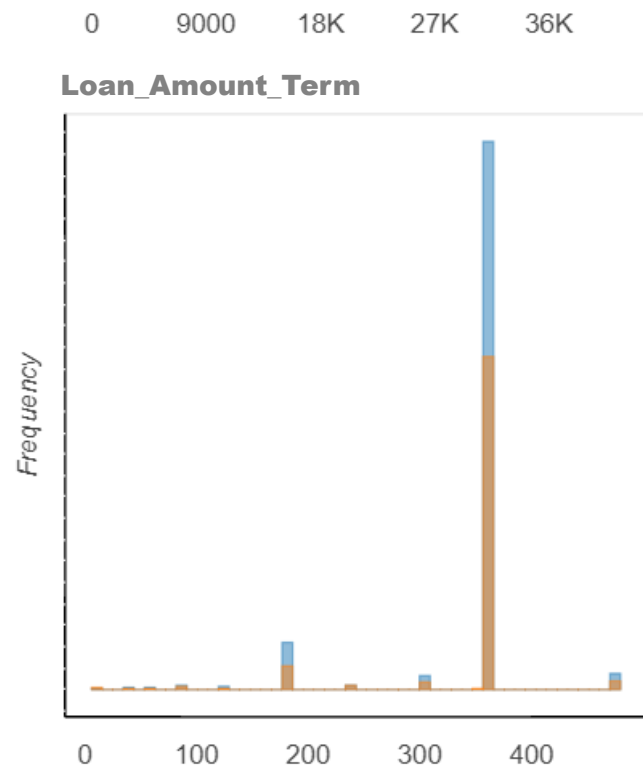
**CoapplicantIncome**

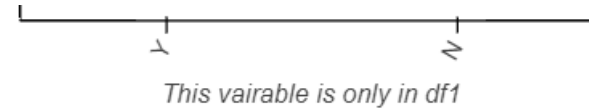


**LoanAmount**









## 2.5 Create Profile Report

- Captures a consolidated report with summary
  - Overview: detect the types of columns in a dataframe
  - Variables: variable type, unique values, distinct count, missing values
  - Quantile statistics like minimum value, Q1, median, Q3, maximum, range, interquartile range
  - Descriptive statistics like mean, mode, standard deviation, sum, median absolute deviation, coefficient of variation, kurtosis, skewness
  - Text analysis for length, sample and letter
  - Correlations: highlighting of highly correlated variables, Spearman, Pearson and Kendall matrices
  - Missing Values: bar chart, heatmap and spectrum of missing values

```
In [62]: create_report(df_train)
```

```
Out[62]:
```

[DataPrep Report](#)[Overview](#)[Variables](#)[Interactions](#)[Correlations](#)[Missing Values](#)

## Overview

### Dataset Statistics

Number of Variables	13
---------------------	----

Number of Rows	614
----------------	-----

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missing cells 149

### Dataset Insights

Gender	has 13 (2.12%) missing values	Missing
--------	-------------------------------	---------

Dependents	has 15 (2.44%) missing values	Missing
------------	-------------------------------	---------

Self_Employed	has 32 (5.21%) missing values	Missing
---------------	-------------------------------	---------

Missing Cells (%)	1.9%
Duplicate Rows	0
Duplicate Rows (%)	0.0%
Total Size in Memory	316.6 KB
Average Row Size in Memory	528.0 B
Variable Types	Categorical: 8 GeoGraphy: 1 Numerical: 4

LoanAmount has 22 (3.58%) missing values

Missing

Loan\_Amount\_Term has 14 (2.28%) missing values

Missing

Credit\_History has 50 (8.14%) missing values

Missing

ApplicantIncome is skewed

Skewed

CoapplicantIncome is skewed

Skewed

LoanAmount is skewed

Skewed

Loan\_Amount\_Term is skewed

Skewed

1

2

Variables

Loan\_ID  
categorical

Show Details

Approximate Distinct Count	614
Approximate Unique (%)	100.0%
Missing	0
Missing (%)	0.0%
Memory Size	43.8 KB

Loan\_ID

Top 10 of 614 Loan\_ID

Gender

Warning icon

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Approximate Distinct Count	2
Approximate Unique (%)	0.3%

Gender  
categorical

Show Details

Missing

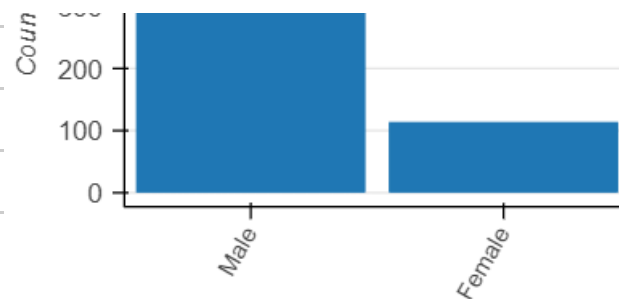
13

Missing (%)

2.1%

Memory Size

40.7 KB



Married  
categorical

Show Details

Approximate Distinct Count

2

Approximate Unique (%)

0.3%

Missing

3

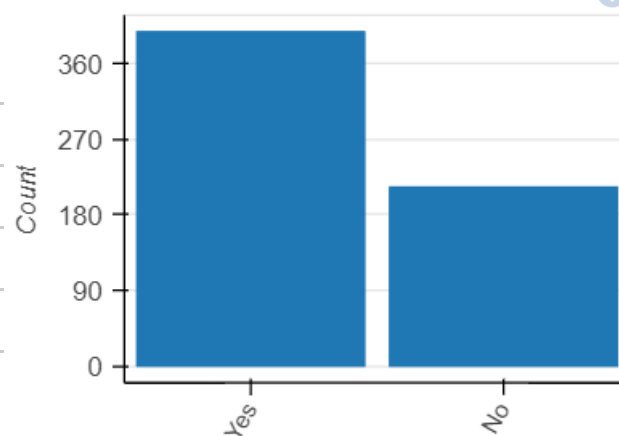
Missing (%)

0.5%

Memory Size

40.4 KB

Married



Dependents  
categorical

Show Details

Approximate Distinct Count

4

Approximate Unique (%)

0.7%

Missing

15

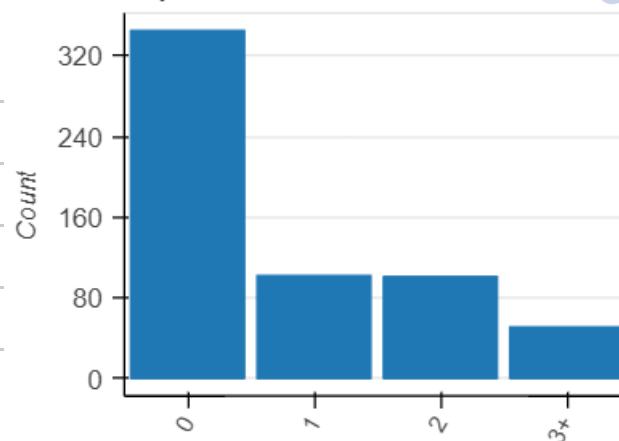
Missing (%)

2.4%

Memory Size

38.7 KB

Dependents



Education



Approximate Distinct Count

2

Education categorical	Approximate Unique (%)		0.3%		
	Missing		0		
	Missing (%)		0.0%		
	Memory Size		44.3 KB		
	<a href="#">Show Details</a>				

Self_Employed categorical	Approximate Distinct Count		2		
	Approximate Unique (%)		0.3%		
	Missing		32		
	Missing (%)		5.2%		
	Memory Size		38.2 KB		
<a href="#">Show Details</a>					

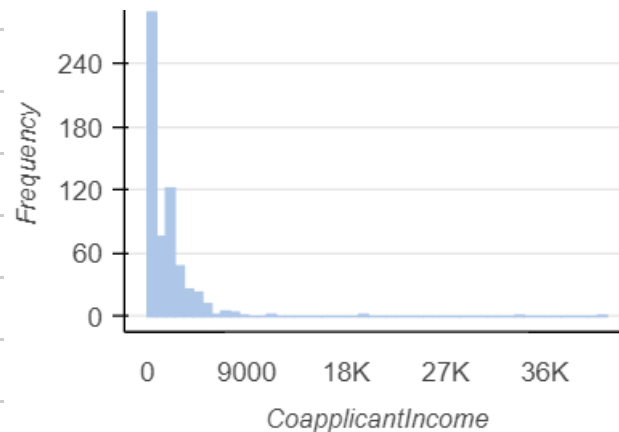
ApplicantIncome numerical	Approximate Distinct Count		505		Mean		5403.4593		
	Approximate Unique (%)		82.2%		Minimum		150		
	Missing		0		Maximum		81000		
	Missing (%)		0.0%		Zeros		0		
	Infinite		0		Zeros (%)		0.0%		
	Infinite (%)		0.0%		Negatives		0		
	Memory Size		9.6 KB		Negatives (%)		0.0%		
	<a href="#">Show Details</a>								

ax/output/CommonHTML/fonts/TeX/fontdata.js		287		Distinct Count		CoapplicantIncome		
Approximate								

CoapplicantIncome numerical	Approximate Unique (%)	46.7%
	Missing	0
	Missing (%)	0.0%
	Infinite	0
	Infinite (%)	0.0%
	Memory Size	9.6 KB
	Mean	1621.2458

Show Details

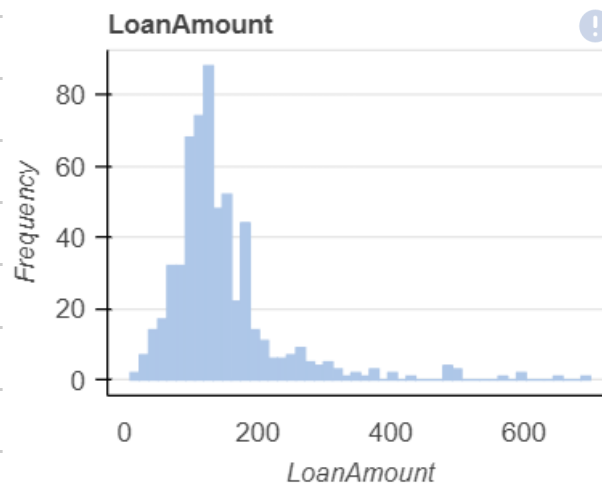
Minimum	0
Maximum	41667
Zeros	273
Zeros (%)	44.5%
Negatives	0
Negatives (%)	0.0%



LoanAmount numerical	Approximate Distinct Count	203
	Approximate Unique (%)	34.3%
	Missing	22
	Missing (%)	3.6%
	Infinite	0
	Infinite (%)	0.0%
	Memory Size	9.2 KB

Show Details

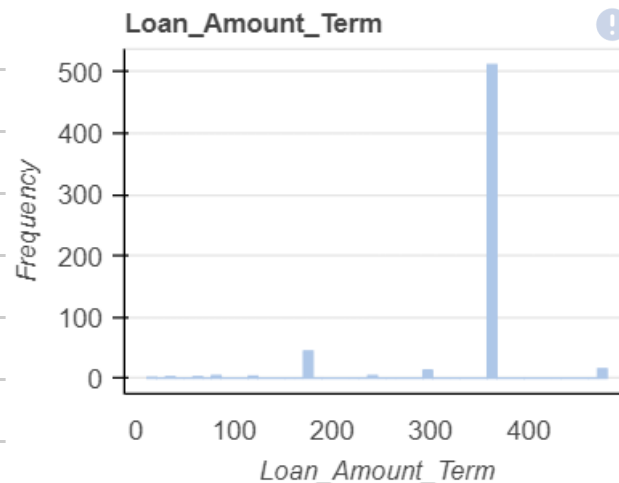
Mean	146.4122
Minimum	9
Maximum	700
Zeros	0
Zeros (%)	0.0%
Negatives	0
Negatives (%)	0.0%



Loan_Amount_Term numerical	Approximate Distinct Count	10
	Approximate Unique (%)	1.7%
	Missing	14
	Missing (%)	2.3%
	Infinite	0
	Infinite (%)	0.0%
	Memory Size	9.4 KB

Show Details

Mean	342
Minimum	12
Maximum	480
Zeros	0
Zeros (%)	0.0%
Negatives	0
Negatives (%)	0.0%



Credit\_History  
categorical

Show Details

Approximate Distinct Count 2

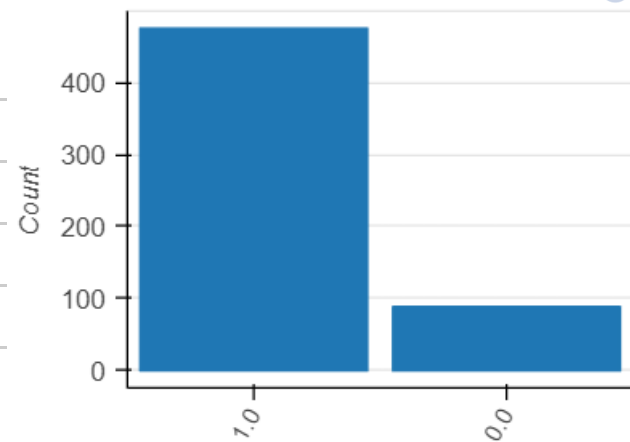
Approximate Unique (%) 0.4%

Missing 50

Missing (%) 8.1%

Memory Size 37.5 KB

Credit\_History



Property\_Area  
categorical

Show Details

Approximate Distinct Count 3

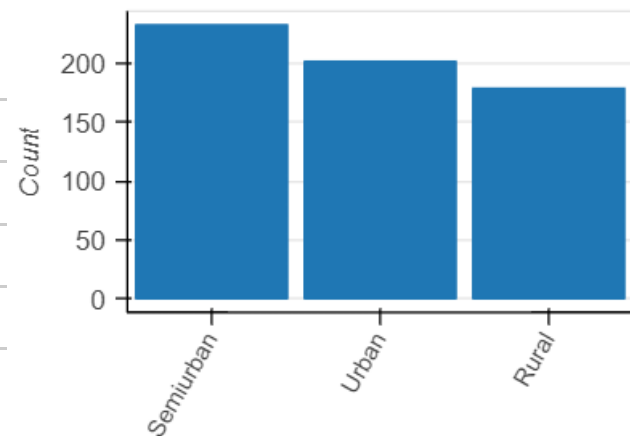
Approximate Unique (%) 0.5%

Missing 0

Missing (%) 0.0%

Memory Size 42.9 KB

Property\_Area



Loan\_Status  
categorical

Show Details

Approximate Distinct Count 2

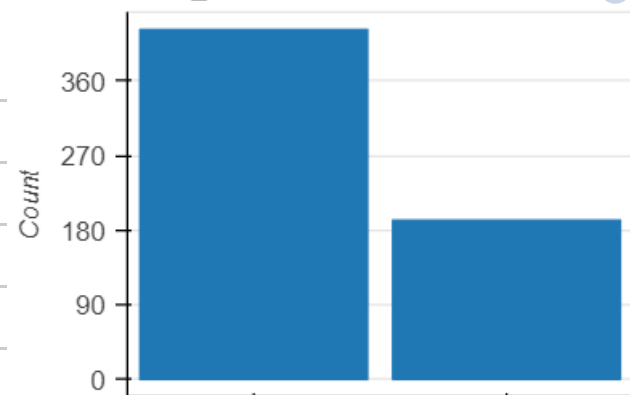
Approximate Unique (%) 0.3%

Missing 0

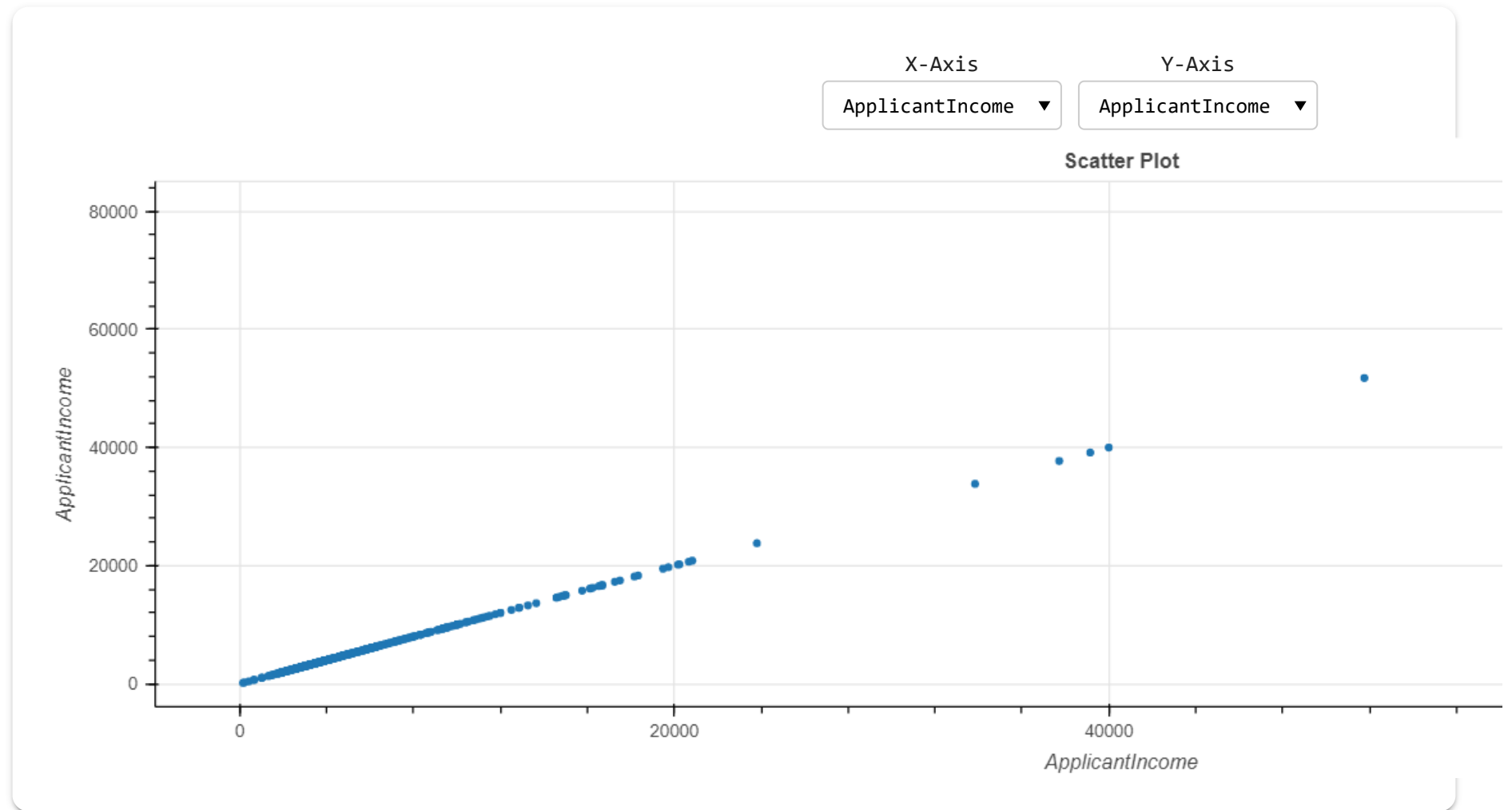
Missing (%) 0.0%

Memory Size 39.6 KB

Loan\_Status

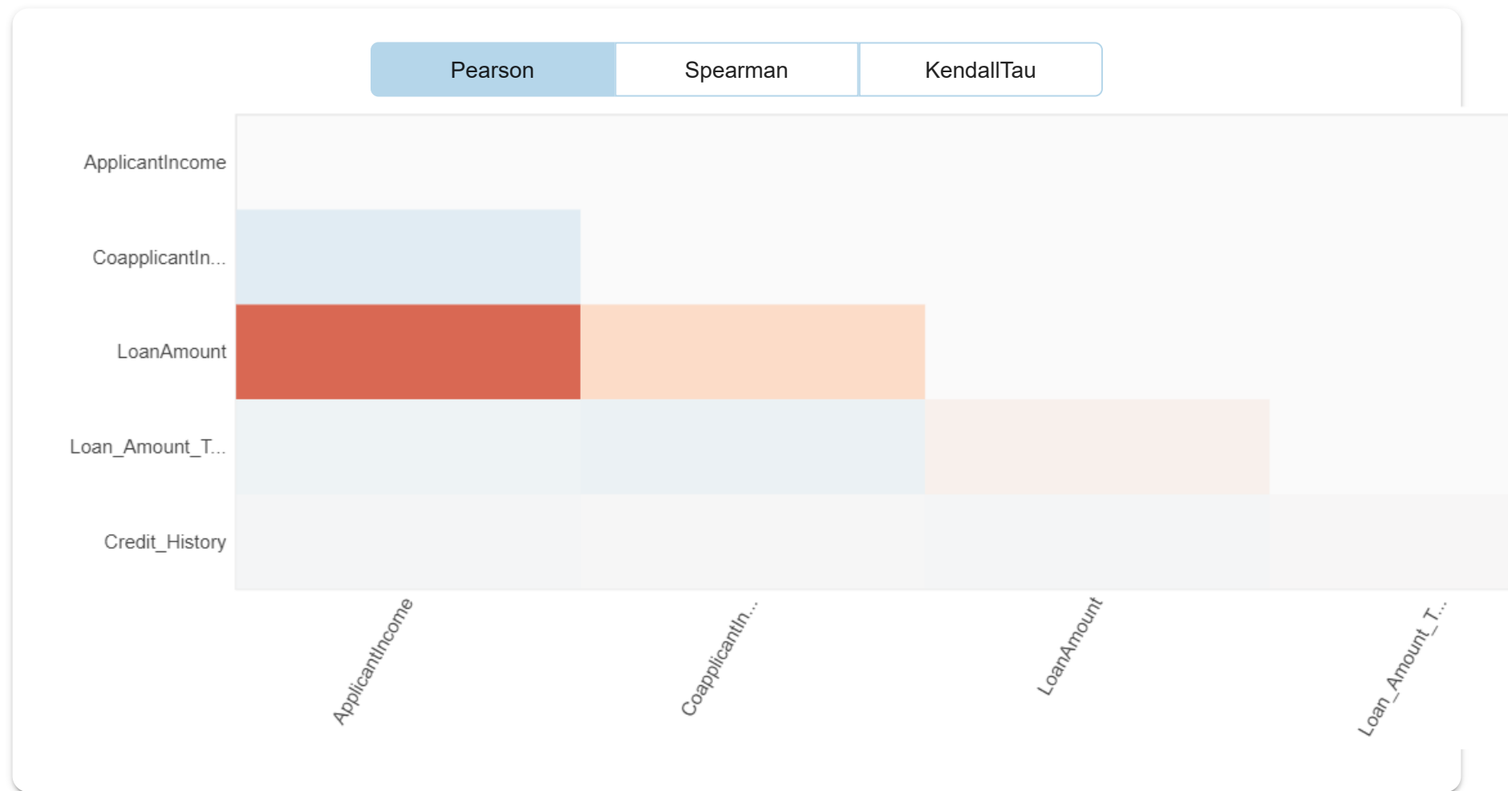


# Interactions

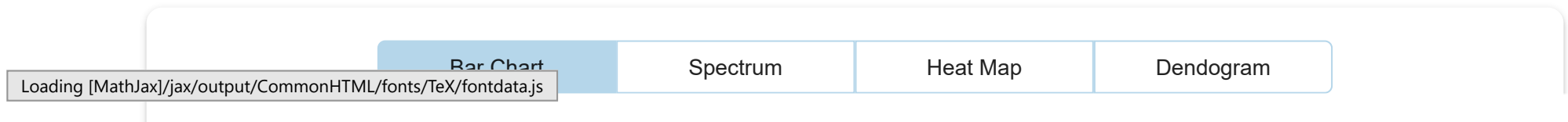


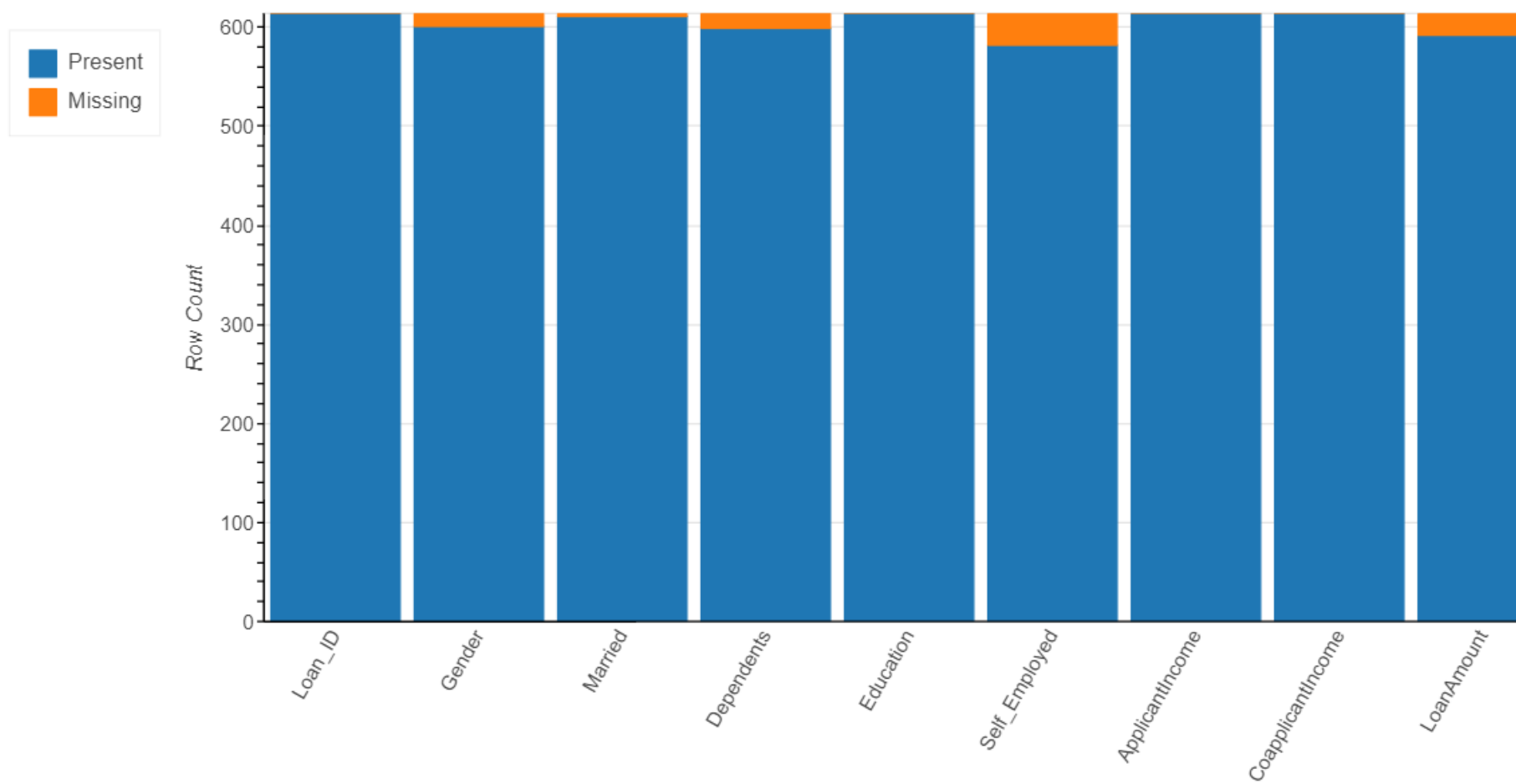


# Correlations



# Missing Values





Report generated with [DataPrep](#)



In [ ]: