

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

This notebook will focus on SweetViz

3. AutoEDA - SweetViz

Dataset Reference: Loan Prediction dataset from Kaggle

Features:

- General Overview - Quick insights of all variables in the dataset using the associations / correlation in the form of a heatmap (including how many duplicates, categorical/numerical/text variables etc.)
- Details about each variables / features in the dataset - missing values, distinct etc.
- Compares Train and Test datasets
- Provides visualization of target variable in context of train dataset

When To Use?

- Need some quick insights about an unknown dataset
- Use this as a basis for your further EDA analysis on top of it
- Need to compare some quick statistical insights between train and test datasets

```
In [1]: import pandas as pd

df_train = pd.read_csv("C:/input/loan-eligible-dataset/loan-train.csv")

df_train.head()
```

```
Out[1]:
```

	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	

	Loan_ID	Gender	Married	Dependents	Education	Self_Employed	ApplicantIncome	CoapplicantIncome	LoanAmount	Loan_Amount_Term	Credit_His
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 [2]: df_test = pd.read_csv("C:/input/loan-eligible-dataset/loan-test.csv")
df_test.head()
```

	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 [3]: df_train.shape
```

```
Out[3]: (614, 13)
```

```
In [4]: df_test.shape
```

```
Out[4]: (367, 12)
```

```
In [6]: !pip install sweetviz
```

Collecting sweetviz

Downloading sweetviz-2.1.3-py3-none-any.whl (15.1 MB)

Requirement already satisfied: scipy>=1.3.2 in c:\programdata\anaconda3\lib\site-packages (from sweetviz) (1.5.2)

Collecting importlib-resources>=1.2.0

```
Downloading importlib_resources-5.2.0-py3-none-any.whl (27 kB)
Requirement already satisfied: tqdm>=4.43.0 in c:\programdata\anaconda3\lib\site-packages (from sweetviz) (4.50.2)
Requirement already satisfied: matplotlib>=3.1.3 in c:\programdata\anaconda3\lib\site-packages (from sweetviz) (3.3.2)
Requirement already satisfied: jinja2>=2.11.1 in c:\programdata\anaconda3\lib\site-packages (from sweetviz) (2.11.2)
Requirement already satisfied: pandas!=1.0.0,!1.0.1,!1.0.2,>=0.25.3 in c:\programdata\anaconda3\lib\site-packages (from sweetviz) (1.1.3)
Requirement already satisfied: numpy>=1.16.0 in c:\users\mishr\appdata\roaming\python\python38\site-packages (from sweetviz) (1.20.1)
Requirement already satisfied: zipp>=3.1.0 in c:\programdata\anaconda3\lib\site-packages (from importlib-resources>=1.2.0->sweetviz) (3.4.0)
Requirement already satisfied: MarkupSafe>=0.23 in c:\programdata\anaconda3\lib\site-packages (from jinja2>=2.11.1->sweetviz) (1.1.1)
Requirement already satisfied: pyparsing!=2.0.4,!2.1.2,!2.1.6,>=2.0.3 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.1.3->sweetviz) (2.4.7)
Requirement already satisfied: pillow>=6.2.0 in c:\users\mishr\appdata\roaming\python\python38\site-packages (from matplotlib>=3.1.3->sweetviz) (7.2.0)
Requirement already satisfied: certifi>=2020.06.20 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.1.3->sweetviz) (2020.6.20)
Requirement already satisfied: cycler>=0.10 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.1.3->sweetviz) (0.10.0)
Requirement already satisfied: kiwisolver>=1.0.1 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.1.3->sweetviz) (1.3.0)
Requirement already satisfied: python-dateutil>=2.1 in c:\programdata\anaconda3\lib\site-packages (from matplotlib>=3.1.3->sweetviz) (2.8.1)
Requirement already satisfied: six in c:\programdata\anaconda3\lib\site-packages (from cycler>=0.10->matplotlib>=3.1.3->sweetviz) (1.15.0)
Requirement already satisfied: pytz>=2017.2 in c:\programdata\anaconda3\lib\site-packages (from pandas!=1.0.0,!1.0.1,!1.0.2,>=0.25.3->sweetviz) (2020.1)
Installing collected packages: importlib-resources, sweetviz
Successfully installed importlib-resources-5.2.0 sweetviz-2.1.3
```

```
WARNING: Ignoring invalid distribution -illow (c:\users\mishr\appdata\roaming\python\python38\site-packages)
WARNING: Ignoring invalid distribution -illow (c:\users\mishr\appdata\roaming\python\python38\site-packages)
WARNING: Ignoring invalid distribution -illow (c:\users\mishr\appdata\roaming\python\python38\site-packages)
WARNING: Ignoring invalid distribution -illow (c:\users\mishr\appdata\roaming\python\python38\site-packages)
WARNING: Ignoring invalid distribution -illow (c:\users\mishr\appdata\roaming\python\python38\site-packages)
```

```
In [7]: import sweetviz as sv
```

```
In [8]: analysis_report = sv.analyze(df_train)
```

```
In [9]: # analysis_report.show_html() # This will generate a separate report named SWEETVIZ_REPORT.html
        analysis_report.show_notebook(w="100%",h="full")
```



2.1.3

[Get updates, docs & report issues here](#)

Created & maintained by [Francois Bertrand](#)

Graphic design by [Jean-Francois Hains](#)

DataFrame

NO COMPARISON TARGET

614 ROWS
0 DUPLICATES
324.2 kb RAM
13 FEATURES
9 CATEGORICAL
3 NUMERICAL
1 TEXT

ASSOCIATIONS

DataFrame

1

Loan_ID

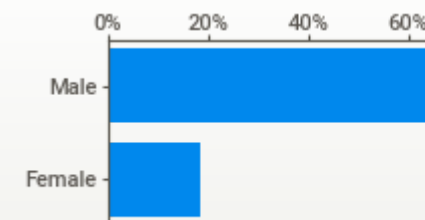
VALUES: 614 (100%)
MISSING: ---
DISTINCT: 614 (100%)

1	<1%	LP002832
1	<1%	LP002226
1	<1%	LP002472
1	<1%	LP001155
1	<1%	LP001790
1	<1%	LP002446
1	<1%	LP002723
607	99%	(Other)

2

Gender

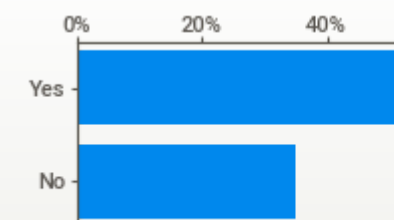
VALUES: 601 (98%)
MISSING: 13 (2%)
DISTINCT: 2 (<1%)



3

Married

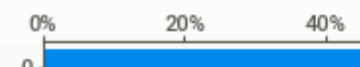
VALUES: 611 (>99%)
MISSING: 3 (<1%)
DISTINCT: 2 (<1%)



4

Dependents

VALUES: 599 (98%)
MISSING: 15 (2%)



DISTINCT: 4 (<1%)

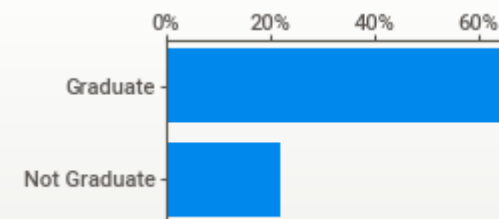


5 Education

VALUES: 614 (100%)

MISSING: ---

DISTINCT: 2 (<1%)

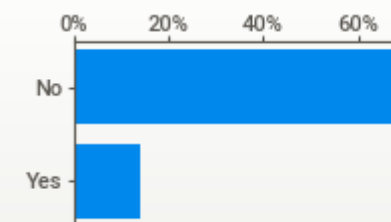


6 Self_Employed

VALUES: 582 (95%)

MISSING: 32 (5%)

DISTINCT: 2 (<1%)



7 ApplicantIncome

VALUES: 614 (100%)

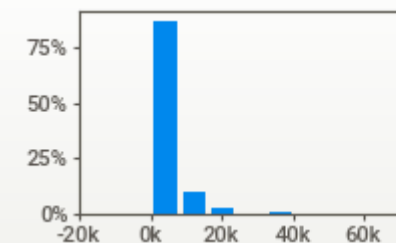
MISSING: ---

DISTINCT: 505 (82%)

ZEROES: ---

MAX	81,000
95%	14,583
Q3	5,795
AVG	5,403
MEDIAN	3,812
Q1	2,878
5%	1,898
MIN	150

RANGE	80,850
IQR	2,918
STD	6,109
VAR	37.3M
KURT.	60.5
SKEW	6.54
SUM	3.3M



8 CoapplicantIncome

VALUES: 614 (100%)

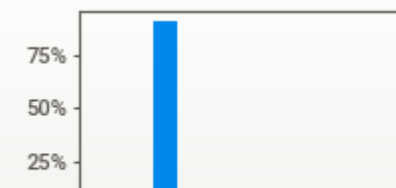
MISSING: ---

DISTINCT: 287 (47%)

ZEROES: 273 (44%)

MAX	41,667
95%	4,997
Q3	2,297
AVG	1,621
MEDIAN	1,188
Q1	0
5%	0

RANGE	41,667
IQR	2,297
STD	2,926
VAR	8.6M
KURT.	85.0
SKEW	7.49



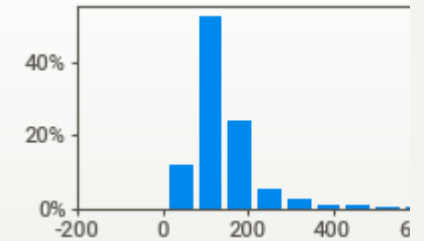
9

LoanAmount

VALUES: 592 (96%)
MISSING: 22 (4%)
DISTINCT: 203 (33%)
ZEROES: ---

MIN 0
MAX 700
95% 298
Q3 168
AVG 146
MEDIAN 128
Q1 100
5% 56
MIN 9

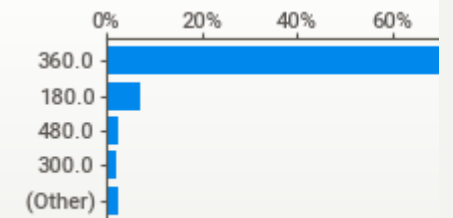
SUM 995k
RANGE 691
IQR 68.0
STD 85.6
VAR 7,325
KURT. 10.4
SKEW 2.68
SUM 86,676



10

Loan_Amount_Term

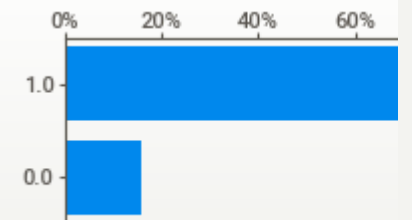
VALUES: 600 (98%)
MISSING: 14 (2%)
DISTINCT: 10 (2%)



11

Credit_History

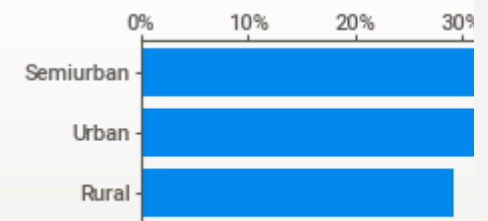
VALUES: 564 (92%)
MISSING: 50 (8%)
DISTINCT: 2 (<1%)



12

Property_Area

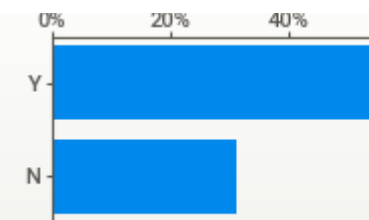
VALUES: 614 (100%)
MISSING: ---
DISTINCT: 3 (<1%)



13

Loan_Status

VALUES: 614 (100%)
MISSING: ---
DISTINCT: 2 (<1%)



```
In [10]: analysis_report2 = sv.analyze([df_train,'Train'], target_feat='Loan_Status')
```

```
In [11]: analysis_report2.show_notebook(w="100%",h="full")
```

Sweetviz

2.1.3

[Get updates, docs & report issues here](#)

Created & maintained by [Francois Bertrand](#)

Graphic design by [Jean-Francois Hains](#)

Train

NO COMPARISON TARGET

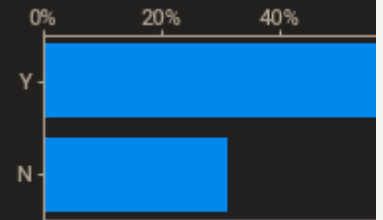
614	ROWS
0	DUPLICATES
324.2 kb	RAM
13	FEATURES
9	CATEGORICAL
3	NUMERICAL
1	TEXT

ASSOCIATIONS

Train 
% Loan_Status 

Loan_Status

VALUES: 614 (100%)
MISSING: ---
DISTINCT: 2 (<1%)



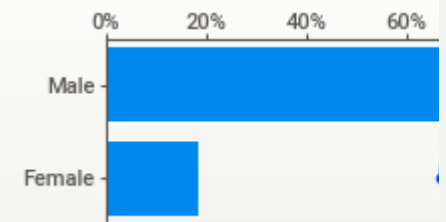
Loan_ID

VALUES: 614 (100%)
MISSING: ---
DISTINCT: 614 (100%)

1	<1%	LP002832
1	<1%	LP002226
1	<1%	LP002472
1	<1%	LP001155
1	<1%	LP001790
1	<1%	LP002446
1	<1%	LP002723
607	99%	(Other)

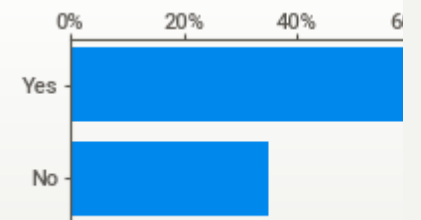
Gender

VALUES: 601 (98%)
MISSING: 13 (2%)
DISTINCT: 2 (<1%)



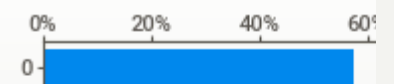
Married

VALUES: 611 (>99%)
MISSING: 3 (<1%)
DISTINCT: 2 (<1%)

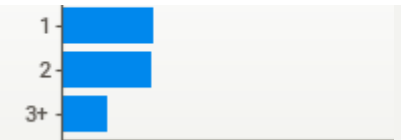


Dependents

VALUES: 599 (98%)
MISSING: 15 (2%)



DISTINCT: 4 (<1%)

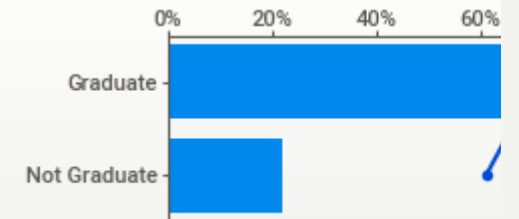


5 Education

VALUES: 614 (100%)

MISSING: ---

DISTINCT: 2 (<1%)

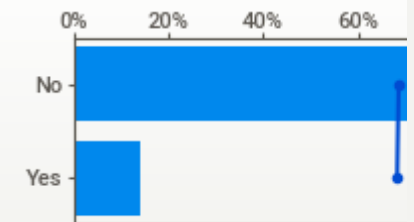


6 Self_Employed

VALUES: 582 (95%)

MISSING: 32 (5%)

DISTINCT: 2 (<1%)



7 ApplicantIncome

VALUES: 614 (100%)

MISSING: ---

DISTINCT: 505 (82%)

ZEROES: ---

MAX 81,000

95% 14,583

Q3 5,795

AVG 5,403

MEDIAN 3,812

Q1 2,878

5% 1,898

MIN 150

RANGE 80,850

IQR 2,918

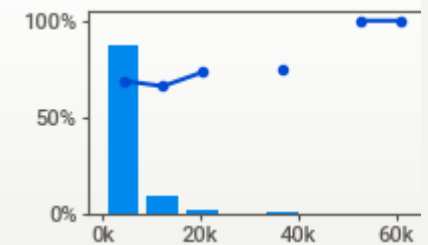
STD 6,109

VAR 37.3M

KURT. 60.5

SKEW 6.54

SUM 3.3M



8 CoapplicantIncome

VALUES: 614 (100%)

MISSING: ---

DISTINCT: 287 (47%)

ZEROES: 273 (44%)

MAX 41,667

95% 4,997

Q3 2,297

AVG 1,621

MEDIAN 1,188

Q1 0

5% 0

MIN 0

RANGE 41,667

IQR 2,297

STD 2,926

VAR 8.6M

KURT. 85.0

SKEW 7.49

SUM 995k



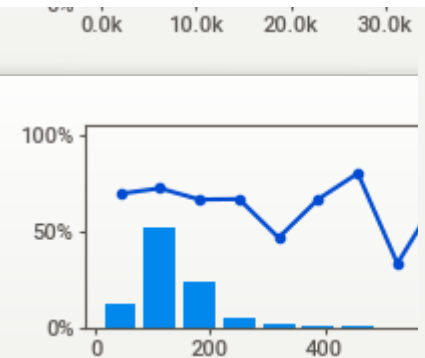
9

LoanAmount

VALUES: 592 (96%)
MISSING: 22 (4%)
DISTINCT: 203 (33%)
ZEROES: ---

MAX 700
95% 298
Q3 168
AVG 146
MEDIAN 128
Q1 100
5% 56
MIN 9

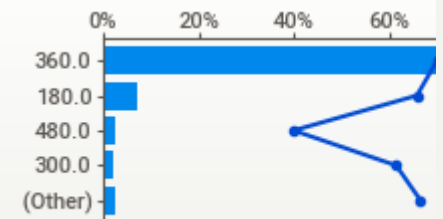
RANGE 691
IQR 68.0
STD 85.6
VAR 7,325
KURT. 10.4
SKEW 2.68
SUM 86,676



10

Loan_Amount_Term

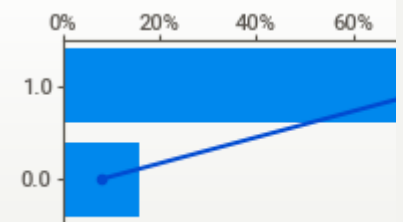
VALUES: 600 (98%)
MISSING: 14 (2%)
DISTINCT: 10 (2%)



11

Credit_History

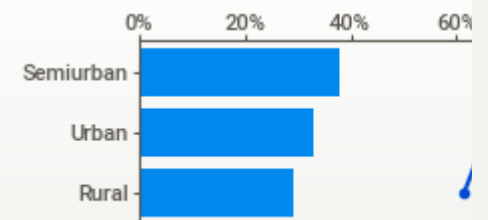
VALUES: 564 (92%)
MISSING: 50 (8%)
DISTINCT: 2 (<1%)



12

Property_Area

VALUES: 614 (100%)
MISSING: ---
DISTINCT: 3 (<1%)



```
In [12]: analysis_report3 = sv.compare([df_train, 'Train'], [df_test, 'Test'], target_feat='Loan_Status')
analysis_report3.show_notebook(w="100%", h="full")
```

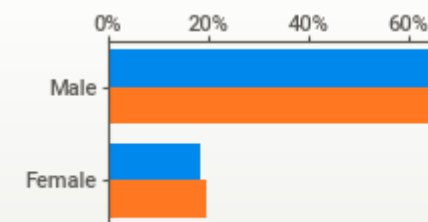


VALUES: 614 (100%) 367 (100%)
MISSING: --- ---
DISTINCT: 614 (100%) 367 (100%)

1	<1%	-	-	LP002832
1	<1%	-	-	LP002226
1	<1%	-	-	LP002472
1	<1%	-	-	LP001155
1	<1%	-	-	LP001790
1	<1%	-	-	LP002446
1	<1%	-	-	LP002723
607	99%	367	100%	(Other)

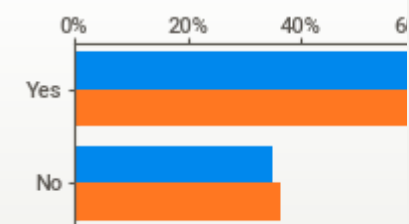
2 Gender

VALUES: 601 (98%) 356 (97%)
MISSING: 13 (2%) 11 (3%)
DISTINCT: 2 (<1%) 2 (<1%)



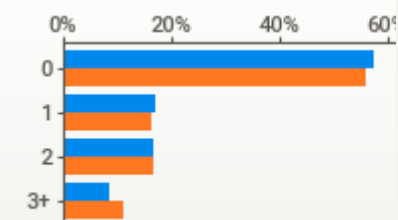
3 Married

VALUES: 611 (>99%) 367 (100%)
MISSING: 3 (<1%) ---
DISTINCT: 2 (<1%) 2 (<1%)



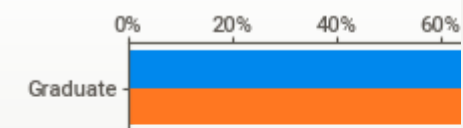
4 Dependents

VALUES: 599 (98%) 357 (97%)
MISSING: 15 (2%) 10 (3%)
DISTINCT: 4 (<1%) 4 (1%)



5 Education

VALUES: 614 (100%) 367 (100%)
MISSING: --- ---
DISTINCT: 2 (<1%) 2 (<1%)

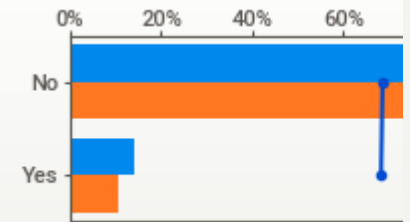


Not Graduate

6

Self_Employed

VALUES: 582 (95%) 344 (94%)
 MISSING: 32 (5%) 23 (6%)
 DISTINCT: 2 (<1%) 2 (<1%)



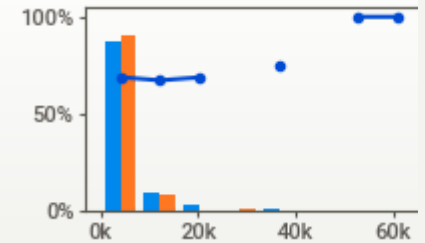
7

ApplicantIncome

VALUES: 614 (100%) 367 (100%)
 MISSING: --- ---
 DISTINCT: 505 (82%) 314 (86%)
 ZEROES: --- 2 (<1%)

MAX	81,000	72,529
95%	14,583	10,000
Q3	5,795	5,060
AVG	5,403	4,806
MEDIAN	3,812	3,786
Q1	2,878	2,864
5%	1,898	1,861
MIN	150	0

RANGE	80,850	72,529
IQR	2,918	2,196
STD	6,109	4,911
VAR	37.3M	24.1M
KURT.	60.5	103
SKEW	6.54	8.44
SUM	3.3M	1.8M



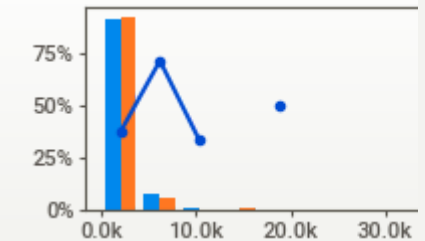
8

CoapplicantIncome

VALUES: 614 (100%) 367 (100%)
 MISSING: --- ---
 DISTINCT: 287 (47%) 194 (53%)
 ZEROES: 273 (44%) 156 (43%)

MAX	41,667	24,000
95%	4,997	4,336
Q3	2,297	2,430
AVG	1,621	1,570
MEDIAN	1,188	1,025
Q1	0	0
5%	0	0
MIN	0	0

RANGE	41,667	24,000
IQR	2,297	2,430
STD	2,926	2,334
VAR	8.6M	5.4M
KURT.	85.0	30.2
SKEW	7.49	4.26
SUM	995k	576k



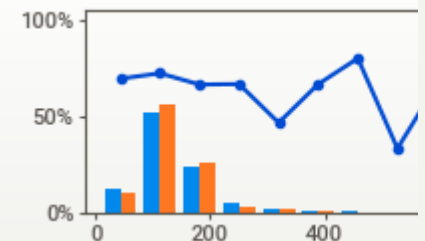
9

LoanAmount

VALUES: 592 (96%) 362 (99%)
 MISSING: 22 (4%) 5 (1%)
 DISTINCT: 203 (33%) 144 (39%)
 ZEROES: --- ---

MAX	700	550
95%	298	240
Q3	168	158
AVG	146	136
MEDIAN	128	125
Q1	100	100
5%	56	64
MIN	9	28

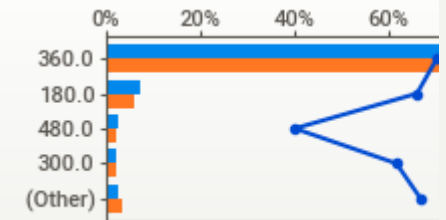
RANGE	691	522
IQR	68.0	57.8
STD	85.6	61.4
VAR	7,325	3,766
KURT.	10.4	9.41
SKEW	2.68	2.22
SUM	86,676	49,280



10

Loan_Amount_Term

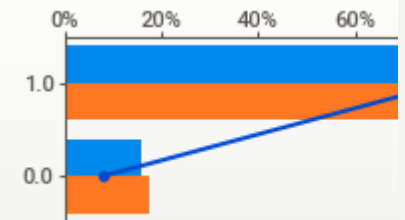
VALUES: 600 (98%) 361 (98%)
MISSING: 14 (2%) 6 (2%)
DISTINCT: 10 (2%) 12 (3%)



11

Credit_History

VALUES: 564 (92%) 338 (92%)
MISSING: 50 (8%) 29 (8%)
DISTINCT: 2 (<1%) 2 (<1%)



12

Property_Area

VALUES: 614 (100%) 367 (100%)
MISSING: --- ---
DISTINCT: 3 (<1%) 3 (<1%)

