

PyAudit: Python Data Audit Library API

Release 1.00

Wenqiang Feng and Ming Chen

# **CONTENTS**

1	Prefa	ace	3
	1.1	About	3
		1.1.1 About this API	3
		1.1.2 About the author	3
	1.2	Feedback and suggestions	4
2	How	to Install	5
	2.1	Install with pip	5
	2.2	Install from Repo	5
		2.2.1 Clone the Repository	5
		2.2.2 Install	5
		2.2.3 Uninstall	5
		2.2.4 Test	6
2	D (I		4.4
3			11
	3.1		11
		71 —	11
		<i>C</i> =	11
		_	12
			12
			13
			14
		3.1.7 correlation matrix	14
	3.2	Summary Functions	15
		3.2.1 numeric_summary	15
		3.2.2 category_summary	16
		3.2.3 auditing	16
4	Andi	ting Demos	19
-	4.1		19
	4.2		21
5	Main	n Reference	29
	14466111	4 410101 01100	_/

Bibliography	31
Python Module Index	33
Index	35



Welcome to our **PyAudit: Python Data Audit Library API!** The PDF version can be downloaded from HERE.

You can install the PyAudit from [PyPI](https://pypi.org/project/PyAudit):

pip install PyAudit

CONTENTS 1

2 CONTENTS

**CHAPTER** 

**ONE** 

## **PREFACE**

#### **Chinese proverb**

Good tools are prerequisite to the successful execution of a job. – old Chinese proverb

## 1.1 About

#### 1.1.1 About this API

This document is the API book for our PyAudit: Python Data Audit Library [PyAudit] API. The PDF version can be downloaded from HERE. You may download and distribute it. Please be aware, however, that the note contains typos as well as inaccurate or incorrect description.

The API assumes that the reader has a preliminary knowledge of python programing and Linux. And this document is generated automatically by using sphinx.

#### 1.1.2 About the author

#### Wengiang Feng

- Sr. Data Scientist and PhD in Mathematics
- University of Tennessee at Knoxville
- Webpage: http://web.utk.edu/~wfeng1/
- Email: von198@gmail.com

#### • Ming Chen

- Data Scientist and PhD in Genome Science and Technology
- University of Tennessee at Knoxville

- Email: ming.chen0919@gmail.com

#### • Biography

Wenqiang Feng is Data Scientist within DST's Applied Analytics Group. Dr. Feng's responsibilities include providing DST clients with access to cutting-edge skills and technologies, including Big Data analytic solutions, advanced analytic and data enhancement techniques and modeling.

Dr. Feng has deep analytic expertise in data mining, analytic systems, machine learning algorithms, business intelligence, and applying Big Data tools to strategically solve industry problems in a cross-functional business. Before joining DST, Dr. Feng was an IMA Data Science Fellow at The Institute for Mathematics and its Applications (IMA) at the University of Minnesota. While there, he helped startup companies make marketing decisions based on deep predictive analytics.

Dr. Feng graduated from University of Tennessee, Knoxville, with Ph.D. in Computational Mathematics and Master's degree in Statistics. He also holds Master's degree in Computational Mathematics from Missouri University of Science and Technology (MST) and Master's degree in Applied Mathematics from the University of Science and Technology of China (USTC).

#### Declaration

The work of Wenqiang Feng was supported by the IMA, while working at IMA. However, any opinion, finding, and conclusions or recommendations expressed in this material are those of the author and do not necessarily reflect the views of the IMA, UTK and DST.

## 1.2 Feedback and suggestions

Your comments and suggestions are highly appreciated. I am more than happy to receive corrections, suggestions or feedbacks through email (Wenqiang Feng: von198@gmail.com and Ming Chen: ming.chen0919@gmail.com) for improvements.

**CHAPTER** 

**TWO** 

## **HOW TO INSTALL**

# 2.1 Install with pip

You can install the PyAudit from [PyPI](https://pypi.org/project/PyAudit):

pip install PyAudit

# 2.2 Install from Repo

# 2.2.1 Clone the Repository

git clone https://github.com/runawayhorse001/PyAudit.git

## 2.2.2 Install

cd PyAudit
pip install -r requirements.txt
python setup.py install

## 2.2.3 Uninstall

pip uninstall statspy

#### 2.2.4 Test

```
cd PyAudit/test
python test.py
```

test.py

```
from PyAudit.basics import missing_rate, zero_rate, dtypes_class
from PyAudit.basics import feature_variance, freq_items_df, feature_len
from PyAudit.basics import corr_matrix, numeric_summary, category_
⇔summary
import pandas as pd
import os, sys
output = os.path.abspath(os.path.join(sys.path[0])) + '/output'
print (output)
d = \{'A': [1, 0, None, 3],
     'B': [1, 0, 0, 0],
     'C': ['a', None, 'c', 'd']}
# create DataFrame
df = pd.DataFrame(d)
print (missing_rate(df))
print (zero_rate(df))
print (feature_variance(df))
print(df)
print (feature_len(df))
print (numeric_summary(df, output))
print(category_summary(df, output))
print(corr_matrix(df, output))
d = {
    'num': list('1223334444'),
    'cat': list('wxxyyyzzzz')
df = pd.DataFrame(d)
df = df.astype({"num": int, "cat": object})
print(freq_items_df(df, top_n=4))
# read df
df = pd.read_csv('Heart.csv', dtype={'Sex': bool})
print (df.head(5))
(num_fields, cat_fields, bool_fields, data_types, type_class) = dtypes_
⇔class(df)
```

```
print (num_fields)
print (cat_fields)
print (bool_fields)
print (data_types)
print (type_class)
print (missing_rate(df))
print (zero_rate(df))

print (freq_items_df(df, top_n=4))
print (feature_len(df))
print (numeric_summary(df, output))
print (category_summary(df, output))
print (corr_matrix(df, output))
```

#### Results:

```
feature missing_rate
                 0.25
       Α
1
       В
                 0.00
                 0.25
       С
 feature zero_rate
       A 0.333333
1
       В
         0.750000
       C 0.000000
 feature feature variance
       Α
1
       В
                      0.5
                      1.0
2
       С
        Sex
               ChestPain RestBP Chol ... Oldpeak Slope
  Age
                                                            Са
     Thal AHD
                             145
                                   233 ...
                                                2.3
 63
      True
                 typical
                                                           0.0
0
   fixed
           No
   67
        True asymptomatic
                             160
                                   286 ...
                                                1.5
                                                           3.0
→ normal Yes
                                                           2.0 _
        True asymptomatic
                             120
                                   229 ...
                                                2.6
⇔reversable Yes
3
   37
        True
               nonanginal
                             130
                                   250
                                                3.5
                                                           0.0
→ normal No
 41 False
               nontypical
                             130
                                   204 ...
                                                1.4
                                                        1 0.0
→ normal No
[5 rows x 14 columns]
['Age', 'RestBP', 'Chol', 'Fbs', 'RestECG', 'MaxHR', 'ExAng', 'Oldpeak
→', 'Slope', 'Ca']
['ChestPain', 'Thal', 'AHD']
```

['S	Sex']			
	feature	dtypes		
0	Age	int64		
1	Sex	bool		
2	ChestPain	object		
3	RestBP	int64		
4	Chol	int64		
5	Fbs	int64		
6	RestECG	int64		
7	MaxHR	int64		
8	ExAng	int64		
9	Oldpeak	float64		
10	Slope	int64		
11	Ca	float64		
12	Thal	object		
13	AHD	object		
10	feature	dtypes	class	
0	Age	int64	numeric	
1	Sex	bool	bool	
2	ChestPain	object	category	
3	RestBP	int64	numeric	
4	Chol	int64	numeric	
	Fbs	int64		
5			numeric	
6	RestECG	int64	numeric	
7	MaxHR	int64	numeric	
8	ExAng	int64	numeric	
9	Oldpeak	float64	numeric	
10	Slope	int64	numeric	
11	Ca	float64	numeric	
12	Thal	object	category	
13	AHD	object	category	
	feature	missing_	rate	
0	Age	0.00		
1	Sex	0.00		
2	ChestPain	0.00		
3	RestBP	0.00	0000	
4	Chol	0.00	0000	
5	Fbs	0.00	0000	
6	RestECG	0.00	0000	
7	MaxHR	0.00	0000	
8	ExAng	0.00	0000	
9	Oldpeak	0.00		
10	Slope	0.00		
11	Ca	0.013201		
12	Thal	0.00		

## PyAudit: Python Data Audit Library API, Release 1.00

(continued from previous page)

13 AHD 0.000000

Process finished with exit code 0

**CHAPTER** 

#### THREE

## **PYTHON DATA AUDIT FUNCTIONS**

## 3.1 Basic Functions

## 3.1.1 dtypes\_class

```
PyAudit.basics.dtypes_class (df_in) numerical, categorical and bool name list in the DataFrame
```

Parameters df\_in - input pandas DataFrame

Returns numerical, categorical and bool name list

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

## 3.1.2 missing\_rate

```
PyAudit.basics.missing_rate (df\_in) calculate missing rate for each feature in the DataFrame
```

**Parameters** df\_in – input pandas DataFrame

**Returns** missing rate

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

```
>>> import pandas as pd
>>> d = {'A': [1, 0, None, 3],
        'B': [1, 0, 0, 0],
        'C': ['a', None, 'c', 'd']}
>>> # create DataFrame
>>> df = pd.DataFrame(d)
>>> from PyAudit.basics import missing_rate
>>> missing_rate(df)
        feature missing_rate
      0 A
                   0.25
      1
             В
                       0.00
      2
             С
                       0.25
```

#### 3.1.3 zero\_rate

```
PyAudit.basics.zero_rate(df_in) calculate the percentage of 0 value for each feature in the DataFrame
```

**Parameters** df\_in – input pandas DataFrame

Returns zero rate

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

## 3.1.4 feature\_variance

```
PyAudit.basics.feature_variance (df_in) calculate the variance for each feature
```

**Parameters** df\_in – input pandas DataFrame

**Returns** feature variance

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

```
>>> import pandas as pd
>>> d = {'A': [1, 0, None, 3],
         'B': [1, 0, 0, 0],
         'C': ['a', None, 'c', 'd']}
>>> # create DataFrame
>>> df = pd.DataFrame(d)
>>> from PyAudit.basics import zero_rate
>>> zero_rate(df)
         feature feature_variance
       0
              Α
                                1.0
                                0.5
       1
               В
               С
       2
                                1.0
```

### 3.1.5 freq\_items\_df

```
PyAudit.basics.freq_items_df(df_in, top_n=3)
```

find out the top n values and the corresponding frequency for each feature

#### **Parameters**

- **df\_in** input pandas DataFrame
- top\_n the number of the top values

**Returns** top n values and the corresponding frequency for each feature

**Author** Wenqiang Feng and Ming Chen

Email von198@gmail.com

## 3.1.6 feature\_len

PyAudit.basics.feature\_len(df\_in)

find out the min and max length of values for each feature

**Parameters** df\_in – input pandas DataFrame

Returns min and max length DataFrame

**Author** Wenqiang Feng and Ming Chen

Email von198@gmail.com

```
>>> d = {'A': [1, 0, None, 3],
         'B': [1, 0, 0, 0],
>>>
         'C': ['a', None, 'c', 'd']}
>>>
>>> # create DataFrame
>>> df = pd.DataFrame(d)
>>> print(df)
         A B
                  C
   0 1.0 1
                  а
   1 0.0 0 None
    2 NaN 0
    3 3.0 0
                  d
>>> print(feature_len(df))
    feature min_length max_length
  0
         Α
                      3
                      1
                                  1
 1
          В
  2
          C
                      1
                                  4
```

#### 3.1.7 correlation matrix

PyAudit.basics.corr\_matrix (*df\_in*, *output\_dir*) generate correlation matrix for numerical dataframe

#### **Parameters**

- **df\_in** input pandas DataFrame
- output\_dir output path

**Returns** correlation matrix

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

# 3.2 Summary Functions

### 3.2.1 numeric summary

PyAudit.basics.numeric\_summary (*df\_in*, *output\_dir*, *top\_n=4*, *deciles=False*) generate statistical summary for numerical DateFrame

#### **Parameters**

- **df\_in** input pandas DataFrame
- output\_dir output files directory
- top\_n the number of the top item to show
- deciles flag for percentiles style

**Returns** statistical summary for numerical data

**Author** Wengiang Feng and Ming Chen

Email von198@gmail.com

```
>>> d = {'A': [1, 0, None, 3],
         'B': [1, 0, 0, 0],
>>>
         'C': ['a', None, 'c', 'd']}
>>> # create DataFrame
>>> df = pd.DataFrame(d)
>>> print (numeric_summary (df))
      feature data_type min_digits ... zero_rate pos_rate
→rate
             float64
                                 3 ... 0.333333 0.666667
   Α
           Α
→ 0.0
           В
                 int64
                                         0.750000 0.250000
   В
→ 0.0
```

#### 3.2.2 category\_summary

PyAudit.basics.category\_summary (*df\_in*, *output\_dir*, *top\_n=4*) generate statistical summary for numerical DateFrame

#### **Parameters**

- **df\_in** input pandas DataFrame
- output\_dir output files directory
- top\_n the number of the top item to show

**Returns** statistical summary for numerical data

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

```
>>> d = {'A': [1, 0, None, 3],
>>> 'B': [1, 0, 0, 0],
>>> 'C': ['a', None, 'c', 'd']}
>>> # create DataFrame
>>> df = pd.DataFrame(d)
>>> print(numeric_summary(df))
feature data_type min_digits ... top_values top_freqs_

→missing_rate
C C object 1 ... [a, d, c] [1, 1, 1]

→ 0.25
```

## 3.2.3 auditing

PyAudit.basics.auditing (df\_in, output\_dir, top\_n=4, deciles=False) generate audited results

#### **Parameters**

- **df\_in** input pandas DataFrame
- output\_dir output files directory
- top\_n the number of the top item to show
- deciles flag for percentiles style

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

```
>>> d = {'A': [1, 0, None, 3],
       'B': [1, 0, 0, 0],
>>>
>>>
      'C': ['a', None, 'c', 'd']}
>>> # create DataFrame
>>> df = pd.DataFrame(d)
>>> print(auditing(df,path))
     feature data_type min_digits ... zero_rate pos_rate neg_
⊶rate
  A
         A float64
                             3 ... 0.333333 0.666667
→ 0.0
                         3 ... 0.750000 0.250000
   В
         B int64
→ 0.0
```



**CHAPTER** 

**FOUR** 

## **AUDITING DEMOS**

The following demos are designed to show how to use PyAudit to aduit pd. DataFrame.

# 4.1 Auditing in one function

For example:

```
# import python libraries
import os
import sys
import pandas as pd
# import PyAudit module
from PyAudit.basics import auditing
# Audit output path
output = os.path.abspath(os.path.join(sys.path[0])) + '/output'
# load DataFrame
df = pd.read_csv('Heart.csv', dtype={'Sex': bool})
print (df.head(5))
# generate the audit results (.csv files in output folder)
num_summary, cat_summary, corr = auditing(df, output)
# the following is optional, since the .csv files are in the output...
-folder
print (num_summary)
print (cat_summary)
print(corr)
```

Result:

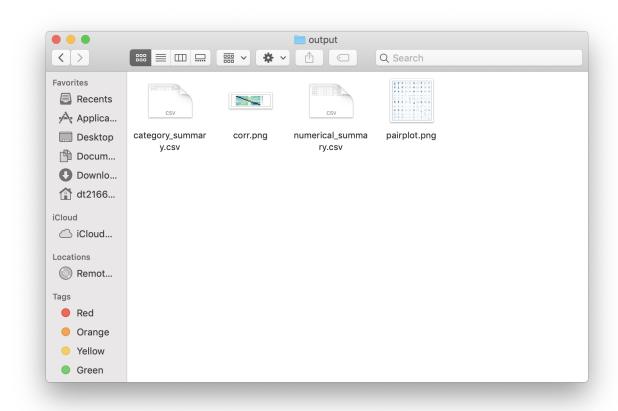
Age	Sex	ChestPain	RestBP	Chol		Oldpeak	Slope	Ca	
→ That	al AHD								
0 63 → fixe	True ed No	typical	145	233	• • •	2.3	3	0.0	ш
→ fixe 1 67		ymptomatic	160	286		1.5	2	3.0	
→ norma									
2 67 →reversa		ymptomatic	120	229	• • •	2.6	2	2.0	_
3 37		nonanginal	130	250		3.5	3	0.0	ت ۔
→ norma									
4 41 I → norma		nontypical	130	204	• • •	1.4	1	0.0	ш
[5 rows z	x 14 colu	ımns]							
	feature	data_type	min_digi	lts	. ze	ero_rate	pos_rate	e ne	g_
→rate Age	Age	int64		4	0	.000000	1.000000	1	
→0.0	Age	111004		4 ••		.000000	1.000000	J	
RestBP	RestBP	int64		4	. 0	.000000	1.000000	)	ш
→0.0 Chol	Chol	int64		5	0	.000000	1.000000	<b>)</b>	
<b>→</b> 0.0	CHOI	111004		J	•	.000000	1.000000	,	
Fbs	Fbs	int64		3	. 0	.851485	0.148515	5	ш
→0.0 RestECG	RestECG	int64		3	. 0	.498350	0.501650	)	
<b>→</b> 0.0									
MaxHR	MaxHR	int64		4	. 0	.000000	1.000000	)	ш
ExAng	ExAng	int64		3	. 0	.673267	0.326733	3	
<b>→0.0</b>					_			_	
Oldpeak →0.0	Oldpeak	float64		3	. 0	.326733	0.67326	7	ш
Slope	Slope	int64		3	. 0	.000000	1.000000	)	
<b>→0.0</b>	Q -	61 + 64		2	0	F00600	0 41107	1	
Ca	Ca	float64		3		.588629	0.411371	L	
[10 rows	x 21 col	umns]							
	feat	ure data_ty	-			p_freqs	missing_		
Sex			ool			.06, 97]		00000	
ChestPair		_				50, 23]		00000	
Thal		hal obje		[1		17, 18]		06601	
AHD		AHD obje	ect		[16	54, 139]	0.00	00000	
[4 rows	k 10 colu	ımns]							
_	Age	e RestBP	Cho	ol	01	.dpeak	Slope		ш
<del>→Ca</del>							(continues	on next	page)

```
Age
       1.000000 0.284946 0.208950 ... 0.203805 0.161770
→362605
RestBP 0.284946 1.000000 0.130120 ... 0.189171 0.117382
→098773
Chol
    0.208950 0.130120 1.000000 ... 0.046564 -0.004062
→119000
Fbs
   0.118530 0.175340 0.009841 ... 0.005747 0.059894
                                                         0.
→145478
RestECG 0.148868 0.146560 0.171043 ... 0.114133 0.133946 0.
→128343
MaxHR -0.393806 -0.045351 -0.003432 ... -0.343085 -0.385601 -0.
→264246
ExAng 0.091661 0.064762 0.061310 ... 0.288223 0.257748 0.
→145570
Oldpeak 0.203805 0.189171 0.046564 ... 1.000000 0.577537 0.
→295832
Slope 0.161770 0.117382 -0.004062 ... 0.577537 1.000000 0.
→110119
Ca 0.362605 0.098773 0.119000 ... 0.295832 0.110119 1.
→000000
[10 rows x 10 columns]
Process finished with exit code 0
```

and

## 4.2 Auditing function by function

#### For example:



```
# load DataFrame
df = pd.read_csv('Heart.csv', dtype={'Sex': bool})
print(df.head(5))

# generate the audit results (.csv files in output folder)
print(numeric_summary(df, output))
print(category_summary(df, output))
print(corr_matrix(df, output))
```

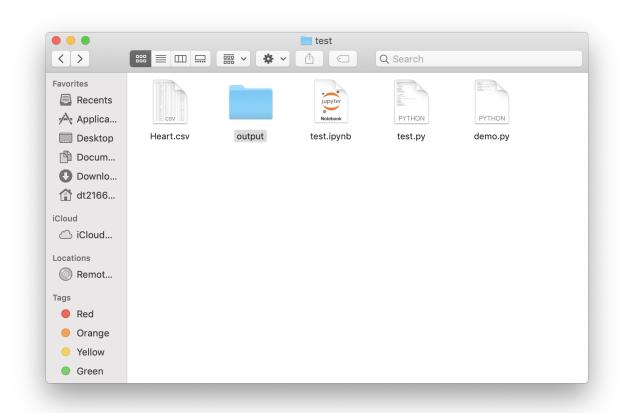
#### Result:

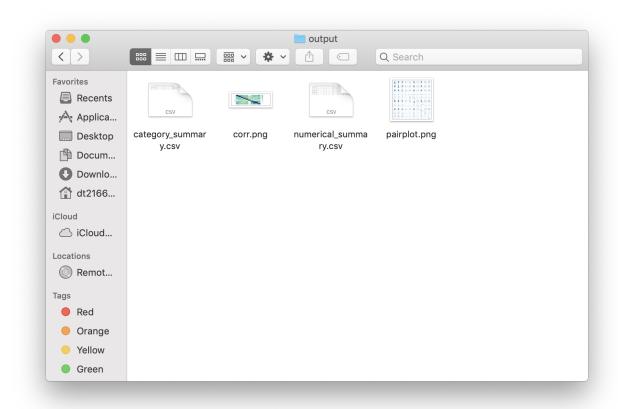
Ag	re Sex	ChestPain	RestBP	Chol		. Oldpeak	Slope	Ca			
$\hookrightarrow$	Thal AHD					-	_		]		
	3 True	typical	145	233		2.3	3	0.0			
$\hookrightarrow$	fixed No										
		symptomatic	160	286		1.5	2	3.0	1		
	ormal Yes										
		symptomatic	120	229	• •	2.6	2	2.0	1		
	rersable Yes						_				
	7 True	nonanginal	130	250	• •	. 3.5	3	0.0	ı		
	ormal No		1 2 0	0.0.4		1 4	1	0 0			
	1 False	nontypical	130	204	• •	1.4	1	0.0	1		
→ n	ormal No										
[5 rows x 14 columns]											
[5 10		data_type	min diai	ts	. 5	zero_rate	pos_rate	e neo	r		
ن ⊶rat		aaca <u>c</u> eppe			•		pob_rac		<b>3</b> —		
Age	Age	int64		4		0.000000	1.00000	0			
<b>→</b> 0.0	-										
RestB	RestBP	int64		4		0.000000	1.00000	0	]		
<b>→</b> 0.0											
Chol	Chol	int64		5		0.000000	1.00000	0	1		
<b>→</b> 0.0											
Fbs	Fbs	int64		3	•	0.851485	0.14851	5	ı		
<b>→</b> 0.0						0 400050	0 50165	^			
RestE		int64		3	•	0.498350	0.50165	U	u		
⊶0.0 MaxHR		int64		4		0.000000	1.00000	$\cap$			
MaxhR		111C04		4	•	0.00000	1.00000	U	]		
ExAng		int64		3		0.673267	0.32673	3			
→0.0	_	T11C 0 4		J	•	0.075207	0.52075.	<i>-</i>			
Oldpe		float64		3		0.326733	0.67326	7			
→0.0	-				-				]		
Slope		int64		3		0.000000	1.00000	0			
<b>→</b> 0.0							(continues				

```
Ca
             Ca
                 float64
                                          0.588629 0.411371
\rightarrow 0.0
[10 rows x 21 columns]
            feature data_type ...
                                         top_freqs missing_rate
               Sex
                      bool ...
                                         [206, 97]
                                                       0.000000
                      object ... [144, 86, 50, 23]
ChestPain ChestPain
                                                       0.000000
                      object ... [166, 117, 18]
Thal
              Thal
                                                       0.006601
                                        [164, 139]
AHD
               AHD
                   object
                                                       0.000000
                             . . .
[4 rows x 10 columns]
            Age RestBP Chol ... Oldpeak
                                                     Slope
→Ca
Age
        1.000000 0.284946 0.208950
                                    ... 0.203805 0.161770 0.
<del>→</del>362605
RestBP 0.284946 1.000000 0.130120
                                    ... 0.189171 0.117382
→098773
                                    ... 0.046564 -0.004062
    0.208950 0.130120 1.000000
Chol
                                                           0.
→119000
Fbs 0.118530 0.175340 0.009841 ... 0.005747 0.059894 0.
→145478
RestECG 0.148868 0.146560 0.171043
                                   ... 0.114133 0.133946 0.
→128343
MaxHR -0.393806 -0.045351 -0.003432
                                   ... -0.343085 -0.385601 -0.
→264246
ExAng 0.091661 0.064762 0.061310 ... 0.288223 0.257748 0.
→145570
Oldpeak 0.203805 0.189171 0.046564 ... 1.000000 0.577537 0.
→295832
Slope 0.161770 0.117382 -0.004062 ... 0.577537 1.000000 0.
→110119
        0.362605 0.098773 0.119000 ... 0.295832 0.110119 1.
\rightarrow 000000
[10 rows x 10 columns]
Process finished with exit code 0
```

and

```
·//·
//i /- /*;;
```





```
\;;
                      .;;;,
/ 0
         \;
                   ,;*;;;*;,
         \_.-') __) --.;;;;**;;;,
`""`;;\
           /-')<u></u>) <u>`</u>';;;;;
 ; * ; ; ;
; ; ; ; |
           0 | ;;*;;
  *;*;\|
                  0 / ;;;;;*
 ) \ | ;;;;;;
/ /` | ';;;*;
 |.
 ,;*;;;\/
              ;;;;;/
  '*wf*/
```

# CHAPTER FIVE

# **MAIN REFERENCE**

# **BIBLIOGRAPHY**

[PyAudit] Wenqiang Feng and Ming Chen. Python Data Audit Library API, 2019.

32 Bibliography

# **PYTHON MODULE INDEX**

# p

PyAudit.basics, 11

# **INDEX**