HomeloanDefaultRisk

April 27, 2022

[12]:	0	SK_ID_CURR 100002	TARGET 1	-	ACT_TYPE	-	ER FLAG_OWN	_CAR \ N	
	1	100002	0		sh loans		F	N	
	2	100004	0	Revolvi			M	Y	
	3	100006	0		sh loans		F	N	
	4	100007	0		sh loans		M	N	
		FLAG_OWN_REA	LTY CNT	_CHILDREN	AMT_INC	OME_TOTAL	AMT_CREDIT	AMT_ANNUITY	_ \
(0		Y	0		202500.0	406597.5	24700.5	,
	1		N	0		270000.0	1293502.5	35698.5	
	2		Y	0		67500.0	135000.0	6750.0	
	3		Y	0		135000.0	312682.5	29686.5	
4	4		Y	0		121500.0	513000.0	21865.5	,
		FLAG_DOC	UMENT_18	FLAG_DOCUI	MENT_19	FLAG_DOCUM	ENT_20 FLAG	_DOCUMENT_21	\
(0	•••	0		0		0	0	
	1		0		0		0	0	
	2		0		0		0	0	
;	3		0		0		0	0	
4	4	•••	0		0		0	0	
		AMT_REQ_CRED	IT BUREAU	J HOUR AMT	REQ CRE	DIT BUREAU	DAY \		
(0	_ `-	_	0.0	-		0.0		
	1			0.0			0.0		
4	2			0.0			0.0		
;	3			NaN			NaN		
4	4			0.0			0.0		
		AMT_REQ_CRE	DIT BURE	AU WEEK A1	MT REQ C	REDIT_BURE	AU MON \		
(0	_ _	_	0.0	- '-	_	0.0		
	1			0.0			0.0		
:	2			0.0			0.0		
;	3			NaN			NaN		
4	4			0.0			0.0		
		AMT_REQ_CRE	DIT_BURE	AU_QRT AM	Γ_REQ CR	EDIT_BUREA	J_YEAR		
(0		_	0.0	_ · _	_	1.0		

```
2
                                0.0
                                                             0.0
      3
                                NaN
                                                             NaN
      4
                                0.0
                                                             0.0
      [5 rows x 122 columns]
[13]: (50000, 122)
[14]: float64
                 65
      int64
                 41
      object
                 16
      dtype: int64
[15]: NAME_CONTRACT_TYPE
                                      2
      CODE_GENDER
                                      3
                                      2
      FLAG_OWN_CAR
                                      2
      FLAG_OWN_REALTY
      NAME_TYPE_SUITE
                                      7
      NAME INCOME TYPE
                                      8
      NAME_EDUCATION_TYPE
                                      5
      NAME_FAMILY_STATUS
                                      6
      NAME HOUSING TYPE
                                      6
      OCCUPATION_TYPE
                                     18
      WEEKDAY_APPR_PROCESS_START
                                      7
      ORGANIZATION_TYPE
                                     58
      FONDKAPREMONT_MODE
                                      4
      HOUSETYPE_MODE
                                      3
                                      7
      WALLSMATERIAL_MODE
                                      2
      EMERGENCYSTATE_MODE
      dtype: int64
                                      0.69922
[16]: COMMONAREA_MEDI
      COMMONAREA_AVG
                                      0.69922
      COMMONAREA_MODE
                                      0.69922
      NONLIVINGAPARTMENTS_MODE
                                      0.69430
      NONLIVINGAPARTMENTS_MEDI
                                      0.69430
      REG_CITY_NOT_LIVE_CITY
                                      0.00000
     LIVE_REGION_NOT_WORK_REGION
                                      0.00000
      REG_REGION_NOT_WORK_REGION
                                      0.00000
      HOUR_APPR_PROCESS_START
                                      0.00000
      SK_ID_CURR
                                      0.00000
      Length: 122, dtype: float64
[18]:
             SK_ID_CURR TARGET NAME_CONTRACT_TYPE CODE_GENDER FLAG_OWN_CAR \
                 100002
                                         Cash loans
      0
                               1
                                                               М
                                                                            N
```

0.0

1

0.0

1	100003		0	Cash	loans		F	N		
2	100004		0	Revolving	loans		M	Y		
3	100006		0	Cash	loans		F	N		
4	100007		0	Cash	loans		M	N		
•••						•••	•••			
49995	157872		0	Cash	loans		M	N		
49996	157873		0	Cash	loans		M	N		
49997	157874		0	Cash	loans		F	N		
49998	157875		0	Cash	loans		F	N		
49999	157876		0	Cash	loans		F	N		
	FLAG_OWN_REAL	.TY	CNT	CHTLDREN A	MT INCO	ME_TOTAL	AMT_CREDI	T\		
0	1 2114_0 1111_112112	Y	0	0		202500.0	406597.			
1		N		0		270000.0	1293502.			
2		Y		0		67500.0	1350002.			
3		Y		0		135000.0	312682.			
4		Y		0		121500.0	513000.			
		1						O		
49995		N		0		126000.0	1125000.	0		
49996		N		1		112500.0	900000.			
49997		Y		0		270000.0	820638.			
49998		Y		0		117000.0	254700.			
49999		Y		0		67500.0	343800.			
10000		•		v		01000.0	010000.	•		
	AMT_ANNUITY		FLAG	_DOCUMENT_1	8 FLAG_	DOCUMENT_	19 FLAG_DO	CUMENT	_20	\
0	AMT_ANNUITY 24700.5		FLAG		8 FLAG_ 0	DOCUMENT_	19 FLAG_DO 0	CUMENT	_20	\
0 1			FLAG			DOCUMENT_		CUMENT		\
	24700.5	•••	FLAG		0	DOCUMENT_	0	CUMENT	0	\
1	24700.5 35698.5		FLAG		0 0	DOCUMENT_	0	CUMENT	0	\
1 2	24700.5 35698.5 6750.0		FLAG		0 0 0	_DOCUMENT_	0 0 0	CUMENT	0 0 0	\
1 2 3	24700.5 35698.5 6750.0 29686.5		FLAG		0 0 0 0	DOCUMENT_	0 0 0	CUMENT	0 0 0	\
1 2 3	24700.5 35698.5 6750.0 29686.5		FLAG		0 0 0 0	_DOCUMENT_	0 0 0	CUMENT	0 0 0	\
1 2 3 4 	24700.5 35698.5 6750.0 29686.5 21865.5 		FLAG		0 0 0 0 0	_DOCUMENT_	0 0 0 0 0	CUMENT	0 0 0 0	\
1 2 3 4 49995	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5		FLAG		0 0 0 0 0	_DOCUMENT_ 	0 0 0 0 0 0	CUMENT	0 0 0 0 0	\
1 2 3 4 49995 49996	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0		FLAG	•••	0 0 0 0 0 0	DOCUMENT_	0 0 0 0 0 	CUMENT	0 0 0 0 0	\
1 2 3 4 49995 49996 49997	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5		FLAG		0 0 0 0 0 0	DOCUMENT_	0 0 0 0 0 0 	CUMENT	0 0 0 0 0	\
1 2 3 4 49995 49996 49997 49998 49999	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0 16155.0				0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0		0 0 0 0 0	\
1 2 3 4 49995 49996 49997 49998 49999	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0	 			0 0 0 0 0 0 0 0	 _HOUR AMT_	0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0	
1 2 3 4 49995 49996 49997 49998 49999	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0 16155.0	 			0 0 0 0 0 0 0 0	 HOUR AMT_ O.O	0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0	
1 2 3 4 49995 49996 49997 49998 49999	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0 16155.0	 0			0 0 0 0 0 0 0 0	 HOUR AMT_ 0.0 0.0	0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0 0 0	
1 2 3 4 49995 49996 49997 49998 49999	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0 16155.0	 0			0 0 0 0 0 0 0 0	 HOUR AMT_ 0.0 0.0 0.0	0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0 0 0	
1 2 3 4 49995 49996 49997 49998 49999 0 1 2 3	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0 16155.0	 0 0			0 0 0 0 0 0 0 0	 HOUR AMT_ 0.0 0.0 0.0 0.0 NaN	0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	
1 2 3 4 49995 49996 49997 49998 49999	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0 16155.0	 0			0 0 0 0 0 0 0 0	 HOUR AMT_ 0.0 0.0 0.0	0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0 0 0	
1 2 3 4 49995 49996 49997 49998 49999 0 1 2 3 4 	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0 16155.0	 0 0 0 0			0 0 0 0 0 0 0 0	HOUR AMT_ 0.0 0.0 0.0 NaN 0.0	0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	
1 2 3 4 49995 49996 49997 49998 49999 0 1 2 3 4 49995	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0 16155.0				0 0 0 0 0 0 0 0	HOUR AMT_ 0.0 0.0 0.0 NaN 0.0	0 0 0 0 0 0 0 0	_BUREA	0 0 0 0 0 0 0 0 0 0 0 0.0 0.0 0.0 0.0	
1 2 3 4 49995 49996 49997 49998 49999 0 1 2 3 4 	24700.5 35698.5 6750.0 29686.5 21865.5 47794.5 26316.0 34897.5 14751.0 16155.0	 0 0 0 0			0 0 0 0 0 0 0 0	HOUR AMT_ 0.0 0.0 0.0 NaN 0.0	0 0 0 0 0 0 0 0	_BUREA	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	

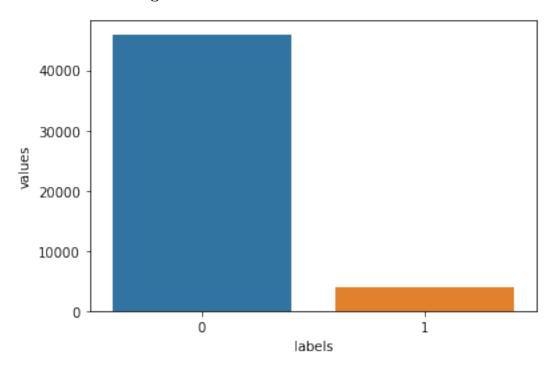
49998 49999	0 0	0.0	0.0
0 1 2 3 4	AMT_REQ_CREDIT_BUREAU_WEEK 0.0 0.0 0.0 NaN 0.0	AMT_REQ_CREDIT_BUREAU_MON 0.0 0.0 0.0 NaN 0.0	\
49995 49996 49997 49998 49999	 0.0 0.0 0.0 0.0	 0.0 0.0 0.0 0.0	
0 1 2 3 4	AMT_REQ_CREDIT_BUREAU_QRT 0.0 0.0 0.0 NaN 0.0	AMT_REQ_CREDIT_BUREAU_YEAR 1.0 0.0 0.0 NaN 0.0	
 49995 49996 49997 49998 49999	 0.0 0.0 2.0 0.0 0.0	 0.0 2.0 4.0 0.0 1.0	

[50000 rows x 81 columns]

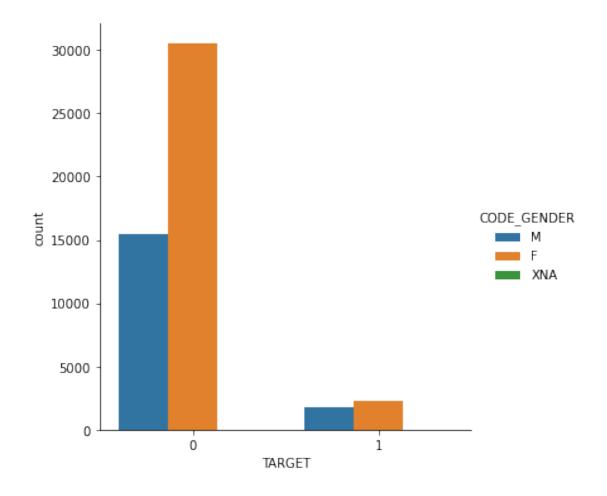
<ipython-input-20-608f89c312d9>:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

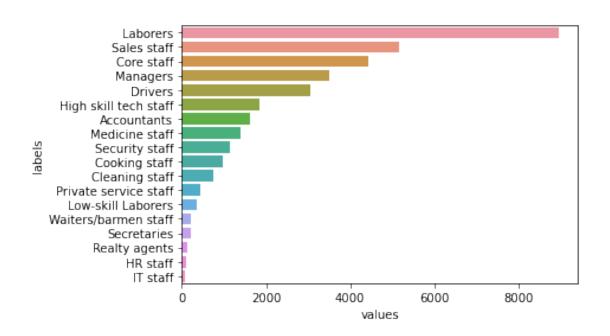
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy application_train['EXT_SOURCE_1']=temp_final['EXT_SOURCE_1']

0.0.1 Distribution of target



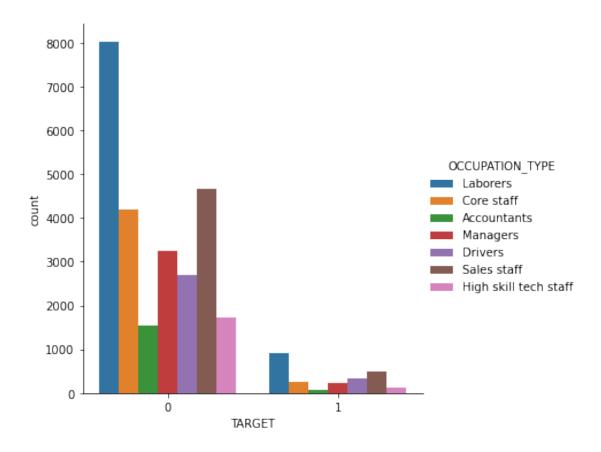
[22]: <seaborn.axisgrid.FacetGrid at 0x21cdc249f10>



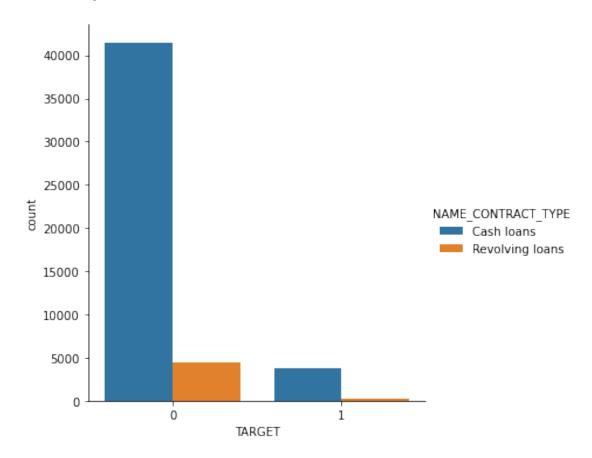


[24]: Laborers True Sales staff True Core staff True Managers True Drivers True High skill tech staff True Accountants True Medicine staff False Security staff False False Cooking staff Cleaning staff False Private service staff False Low-skill Laborers False Waiters/barmen staff False Secretaries False False Realty agents HR staff False IT staff False Name: OCCUPATION_TYPE, dtype: bool

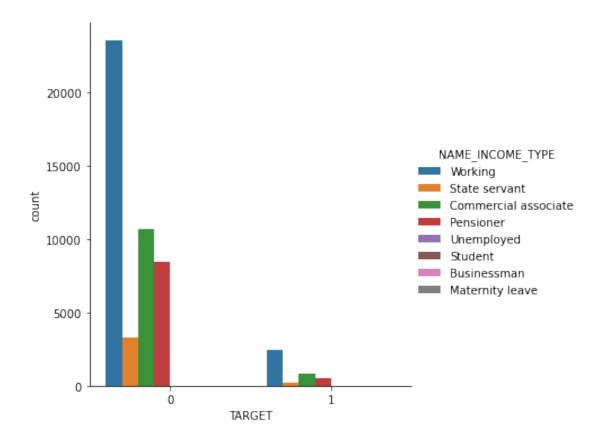
[25]: <seaborn.axisgrid.FacetGrid at 0x21c9ea99520>



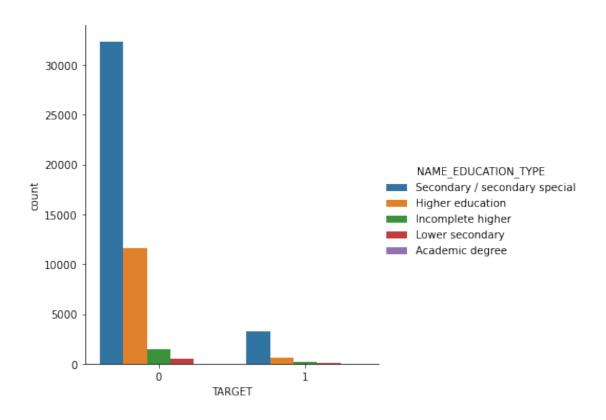
[26]: <seaborn.axisgrid.FacetGrid at 0x21c82062280>

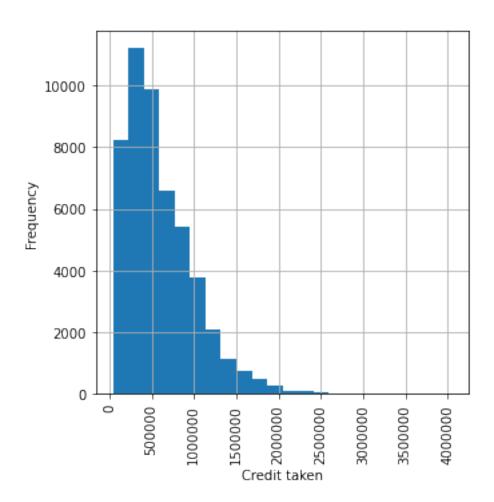


[27]: <seaborn.axisgrid.FacetGrid at 0x21c81709b80>

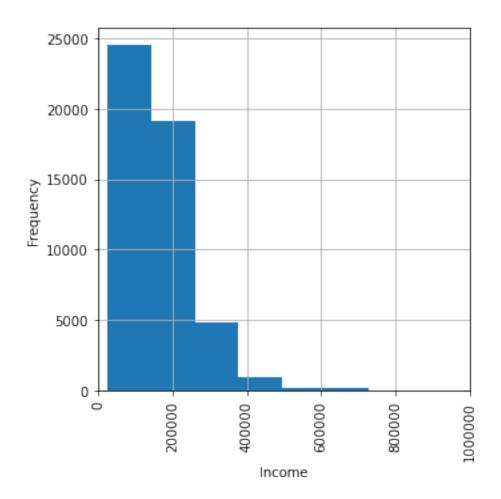


[28]: <seaborn.axisgrid.FacetGrid at 0x21c820b4e20>



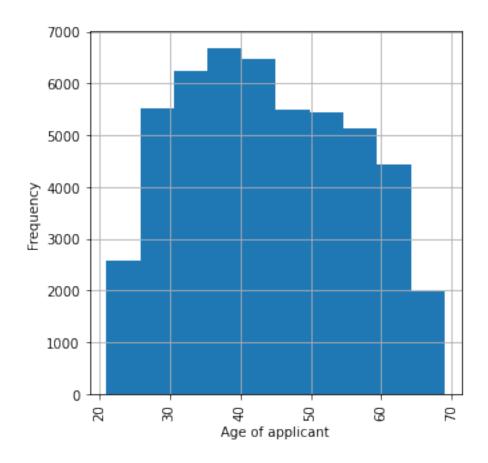


[30]: (0.0, 1000000.0)



<ipython-input-31-d8c0895208ad>:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy temp['Age'] = application_train['DAYS_BIRTH']*-1/365



3 columns were label encoded.

<ipython-input-32-00924bf856fb>:13: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy application_train[col] = le.transform(application_train[col])

Training Features shape: (50000, 193)
Training Features shape: (50000, 193)

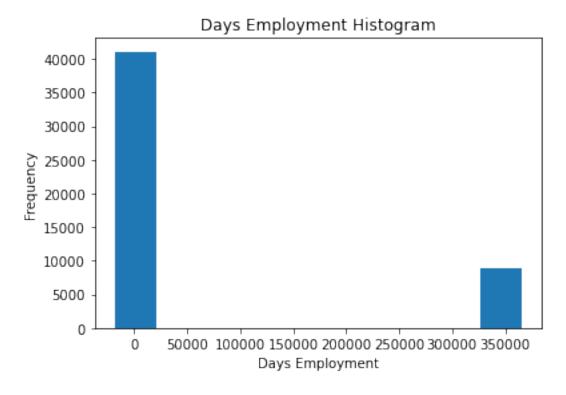
[35]:	count	50000.000000
	mean	43.896192
	std	11.948995
	min	21.041096
	25%	33.914384
	50%	43.098630
	75%	53.819178
	max	68.997260

Name: DAYS_BIRTH, dtype: float64

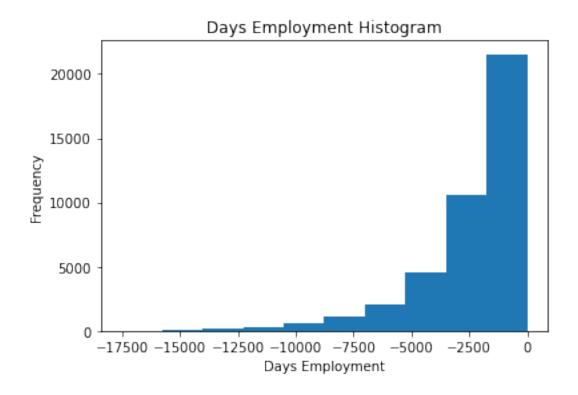
[26].	count	50000.000000
[30].	Count	50000.000000
	mean	63218.143580
	std	140793.489022
	min	-17531.000000
	25%	-2786.000000
	50%	-1221.000000
	75%	-292.000000
	max	365243.000000

Name: DAYS_EMPLOYED, dtype: float64

The maximum value (besides being positive) is about 1000 years! That doesn't look right!



The non-anomalies default on 8.58% of loans The anomalies default on 5.64% of loans There are 8924 anomalous days of employment



Most Positive Correlations:	
LIVE_CITY_NOT_WORK_CITY	0.032264
OCCUPATION_TYPE_Laborers	0.038188
REG_CITY_NOT_LIVE_CITY	0.038749
FLAG_EMP_PHONE	0.041406
DEF_30_CNT_SOCIAL_CIRCLE	0.041560
DAYS_REGISTRATION	0.042334
DEF_60_CNT_SOCIAL_CIRCLE	0.044210
FLAG_DOCUMENT_3	0.045046
DAYS_ID_PUBLISH	0.046925
REG_CITY_NOT_WORK_CITY	0.048438
NAME_INCOME_TYPE_Working	0.053948
NAME_EDUCATION_TYPE_Secondary / secondary special	0.055914
DAYS_LAST_PHONE_CHANGE	0.056131
CODE_GENDER_M	0.058687
REGION_RATING_CLIENT	0.066131
REGION_RATING_CLIENT_W_CITY	0.067080
DAYS_EMPLOYED	0.076535
DAYS_BIRTH	0.076792
TARGET	1.000000
FLAG_DOCUMENT_12	NaN
Name: TARGET, dtype: float64	

Most Negative Correlations:

```
EXT_SOURCE_3
                                               -0.159550
     EXT_SOURCE_2
                                              -0.158265
     EXT_SOURCE_1
                                              -0.156806
                                              -0.076792
     Age
     NAME EDUCATION TYPE Higher education
                                              -0.064011
     CODE GENDER F
                                              -0.058661
     EMERGENCYSTATE MODE No
                                             -0.048843
     NAME_INCOME_TYPE_Pensioner
                                             -0.041708
     DAYS EMPLOYED ANOM
                                             -0.041378
     ORGANIZATION_TYPE_XNA
                                             -0.041378
     AMT_GOODS_PRICE
                                             -0.041290
     REGION_POPULATION_RELATIVE
                                             -0.040797
     NAME_CONTRACT_TYPE
                                              -0.036767
     FLAG_PHONE
                                             -0.032688
     AMT_CREDIT
                                             -0.032424
     HOUR_APPR_PROCESS_START
                                             -0.032040
     NAME_FAMILY_STATUS_Married
                                             -0.029020
     OCCUPATION_TYPE_Core staff
                                             -0.027672
     FLOORSMAX_AVG
                                             -0.026643
     FLOORSMAX MEDI
                                             -0.026141
     Name: TARGET, dtype: float64
     0.0.2 Feature Engineering
     Polynomial Features shape:
                                  (50000, 21)
[43]: ['1',
       'EXT_SOURCE_1',
       'EXT_SOURCE_2',
       'EXT_SOURCE_3',
       'DAYS_EMPLOYED',
       'DAYS_BIRTH',
       'EXT_SOURCE_1^2',
       'EXT_SOURCE_1 EXT_SOURCE_2',
       'EXT_SOURCE_1 EXT_SOURCE_3',
       'EXT_SOURCE_1 DAYS_EMPLOYED',
       'EXT_SOURCE_1 DAYS_BIRTH',
       'EXT_SOURCE_2^2',
       'EXT_SOURCE_2 EXT_SOURCE_3',
       'EXT_SOURCE_2 DAYS_EMPLOYED',
       'EXT_SOURCE_2 DAYS_BIRTH']
     EXT_SOURCE_2 EXT_SOURCE_3
                                  -0.194199
     EXT_SOURCE_1 EXT_SOURCE_3
                                  -0.167388
     EXT_SOURCE_1 EXT_SOURCE_2
                                  -0.165564
     EXT_SOURCE_3
                                  -0.159550
     EXT_SOURCE_2
                                  -0.158265
     EXT_SOURCE_2^2
                                  -0.146820
```

EXT_SOURCE_3^2 -0.144592 EXT_SOURCE_1 -0.099736 EXT_SOURCE_1^2 -0.090331 DAYS_EMPLOYED^2 -0.075400

Name: TARGET, dtype: float64

EXT_SOURCE_1 DAYS_EMPLOYED 0.103650 EXT_SOURCE_3 DAYS_EMPLOYED 0.151375 EXT_SOURCE_2 DAYS_EMPLOYED 0.154554 TARGET 1.000000 1 NaN

Name: TARGET, dtype: float64

Training data with polynomial features shape: (50000, 30)

Training data shape: (50000, 29)

0.0.3 Logistic Regression

Confusion Matrix :

0 1

0 13788 0 1 1212 0

Test accuracy = 91.92 %

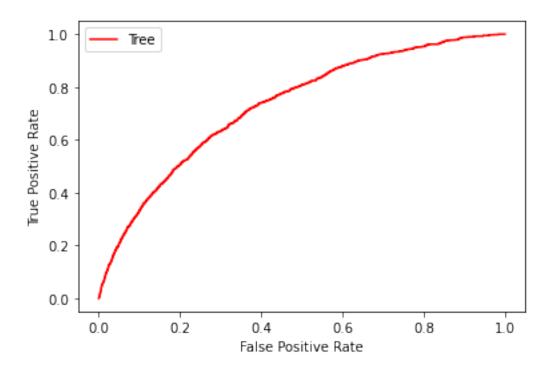
	precision	recall	f1-score	support
	0.00	4 00		40700
0	0.92	1.00	0.96	13788
1	0.00	0.00	0.00	1212
accuracy			0.92	15000
macro avg	0.46	0.50	0.48	15000
weighted avg	0.84	0.92	0.88	15000

AUC Tree for Logistic: 0.7304789715263955

C:\ProgramData\Anaconda3\lib\site-

packages\sklearn\metrics_classification.py:1221: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))



0.0.4 Random Forest Classifier

Confusion Matrix :

0 1

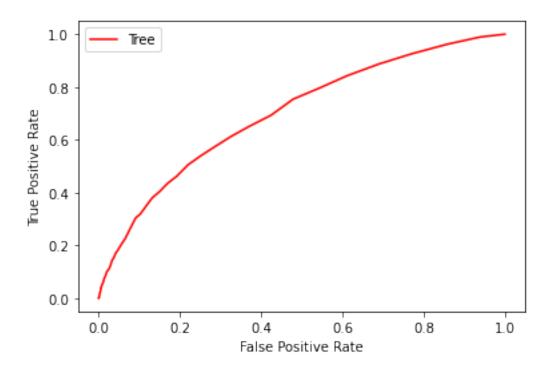
0 13745 43

1 1185 27

Test accuracy = 91.81333333333333333 %

v	precision	recall	f1-score	support
0	0.92	1.00	0.96	13788
1	0.39	0.02	0.04	1212
accuracy			0.92	15000
macro avg	0.65	0.51	0.50	15000
weighted avg	0.88	0.92	0.88	15000

AUC Tree for Random Forest: 0.7005428023220076



0.0.5 Decision tree

Confusion Matrix :

1

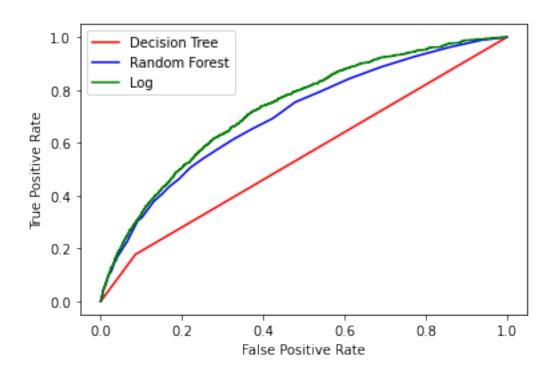
0 1 12610 1178

997 215

Test accuracy = 85.5 %

lest accuracy	precision	recall	f1-score	support
0	0.93	0.91	0.92	13788
1	0.15	0.18	0.17	1212
accuracy			0.85	15000
macro avg	0.54	0.55	0.54	15000
weighted avg	0.86	0.85	0.86	15000

AUC Tree for Decision Tree: 0.5459780638638276



0.0.6 SVM

C:\ProgramData\Anaconda3\lib\site-packages\sklearn\utils\validation.py:67: FutureWarning: Pass C=1.0 as keyword args. From version 0.25 passing these as positional arguments will result in an error

warnings.warn("Pass $\{\}$ as keyword args. From version 0.25 "

Confusion Matrix :

0 1 0 0 1212 1 0 13788 accuracy: 91.92

AUC Tree for SVM: 0.5812690113658886

0.0.7 Neural nets

---- Test data ----Accuracy: 91.86

Classification Report:

	precision	recall	f1-score	support
0 1	0.92 0.29	1.00	0.96 0.01	13788 1212
accuracy			0.92	15000

macro avg 0.60 0.50 0.48 15000 weighted avg 0.87 0.92 0.88 15000

Confusion Matrix: [[13773 15]

[1206 6]]

AUC Tree: 0.7544781131725009