

# Project-Pdfoutput-Team4

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## 1 Machine Learning I Final Project

Team 4

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```
[ ]: # import packages
import pandas as pd
from datetime import datetime
import numpy as np
from keras.models import Model, Sequential
from keras.layers import Dense, Activation, Dropout, BatchNormalization, Embedding, Flatten, Input, concatenate
from keras_tuner import BayesianOptimization, RandomSearch
from sklearn.preprocessing import MinMaxScaler, StandardScaler, LabelEncoder
import matplotlib.pyplot as plt
from sklearn.metrics import mean_absolute_error
import keras.backend as K
import sklearn
import category_encoders as ce
from keras.optimizers import SGD, Adam, RMSprop
from keras.callbacks import EarlyStopping
from sklearn.metrics import log_loss
from keras import regularizers
from sklearn.model_selection import train_test_split
from keras import initializers
from datatable import dt, f, by, g, join, sort, update, ifelse
from sklearn.model_selection import GridSearchCV
from lightgbm import LGBMClassifier
import seaborn as sns
```

## 2 Data Exploration

### 2.1 Data Import

```
[ ]: # Import the training data
train = pd.read_csv('Project Data/ProjectTrainingData.csv')
```

```
[ ]: # Import the test data
test = pd.read_csv('Project Data/ProjectTestData.csv')
```

```
[ ]: train.shape
# there are 31991090 rows and 24 columns
```

(31991090, 24)

```
[ ]: # shuffle the train dataset into 10 parts
shuffled = train.sample(frac=1,random_state=42)
result = np.array_split(shuffled, 10)
```

```
[ ]: # name those parts with number
for i in range(len(result)):
    exec(f'train_{i} = result[i]')
```

```
[ ]: print(train_3.shape)
```

(3199109, 24)

### 2.2 Data Overview

```
[ ]: # Explore the training data
train_3.head()
```

```
[ ]:
```

	id	click	hour	C1	banner_pos	site_id	\
5616589	1.211469e+19	0	14102210	1005	1	5114c672	
31825749	1.028758e+19	0	14102922	1005	0	85f751fd	
13548281	1.757268e+19	1	14102413	1005	0	2b1ddb24	
11682204	7.256027e+18	0	14102322	1005	0	85f751fd	
29186795	1.731978e+18	0	14102904	1005	0	8d7e1373	

	site_domain	site_category	app_id	app_domain	...	device_type	\
5616589	3f2f3819	3e814130	ecad2386	7801e8d9	...	1	
31825749	c4e18dd6	50e219e0	e2fcccd2	5c5a694b	...	1	
13548281	98acf46c	3e814130	ecad2386	7801e8d9	...	1	
11682204	c4e18dd6	50e219e0	8bcb1385	2347f47a	...	1	
29186795	3b953bf0	75fa27f6	ecad2386	7801e8d9	...	1	

	device_conn_type	C14	C15	C16	C17	C18	C19	C20	C21
5616589	0	21764	216	36	2506	0	35	100076	157

31825749	0	4687	320	50	423	2	39	100148	32
13548281	0	20128	320	50	2303	2	39	100148	23
11682204	0	15708	320	50	1722	0	35	-1	79
29186795	0	22951	320	50	2654	3	38	-1	23

[5 rows x 24 columns]

```
[ ]: # set test set naive prediction value as 0.5
test['click'] = 0.5
# set test set column names
test = test[['id', 'click', 'hour', 'C1', 'banner_pos', 'site_id',
↪ 'site_domain', 'site_category', 'app_id', 'app_domain', 'app_category',
↪ 'device_id', 'device_ip', 'device_model',
↪ 'device_type', 'device_conn_type', 'C14', 'C15', 'C16',
↪ 'C17', 'C18', 'C19', 'C20', 'C21']]
```

```
[ ]: test.shape
```

```
[ ]: (13015341, 24)
```

```
[ ]: train_3.describe(include='all')
```

```
[ ]:
count      id      click      hour      C1      banner_pos  \
count  3.199109e+06  3.199109e+06  3.199109e+06  3.199109e+06  3.199109e+06
unique      NaN      NaN      NaN      NaN      NaN
top      NaN      NaN      NaN      NaN      NaN
freq      NaN      NaN      NaN      NaN      NaN
mean  9.221241e+18  1.697601e-01  1.410251e+07  1.004971e+03  2.939415e-01
std   5.325019e+18  3.754219e-01  2.675345e+02  1.116570e+00  5.107964e-01
min   6.644102e+12  0.000000e+00  1.410210e+07  1.001000e+03  0.000000e+00
25%   4.609702e+18  0.000000e+00  1.410222e+07  1.005000e+03  0.000000e+00
50%   9.218715e+18  0.000000e+00  1.410251e+07  1.005000e+03  0.000000e+00
75%   1.383416e+19  0.000000e+00  1.410280e+07  1.005000e+03  1.000000e+00
max   1.844674e+19  1.000000e+00  1.410292e+07  1.012000e+03  7.000000e+00
```

```
count      site_id site_domain site_category      app_id app_domain  ...  \
count   3199109     3199109         3199109   3199109     3199109  ...
unique      3278         4010             23       4586         294  ...
top    85f751fd     c4e18dd6        50e219e0  ecad2386     7801e8d9  ...
freq    1117161     1160312        1273551   2081948     2199153  ...
mean      NaN         NaN             NaN         NaN         NaN  ...
std      NaN         NaN             NaN         NaN         NaN  ...
min      NaN         NaN             NaN         NaN         NaN  ...
25%      NaN         NaN             NaN         NaN         NaN  ...
50%      NaN         NaN             NaN         NaN         NaN  ...
75%      NaN         NaN             NaN         NaN         NaN  ...
max      NaN         NaN             NaN         NaN         NaN  ...
```

	device_type	device_conn_type	C14	C15 \
count	3.199109e+06	3.199109e+06	3.199109e+06	3.199109e+06
unique	NaN	NaN	NaN	NaN
top	NaN	NaN	NaN	NaN
freq	NaN	NaN	NaN	NaN
mean	1.018046e+00	3.294061e-01	1.858709e+04	3.188324e+02
std	5.407373e-01	8.574898e-01	4.972518e+03	2.145895e+01
min	0.000000e+00	0.000000e+00	3.750000e+02	1.200000e+02
25%	1.000000e+00	0.000000e+00	1.661500e+04	3.200000e+02
50%	1.000000e+00	0.000000e+00	2.015300e+04	3.200000e+02
75%	1.000000e+00	0.000000e+00	2.177000e+04	3.200000e+02
max	5.000000e+00	5.000000e+00	2.383600e+04	1.024000e+03

	C16	C17	C18	C19	C20 \
count	3.199109e+06	3.199109e+06	3.199109e+06	3.199109e+06	3.199109e+06
unique	NaN	NaN	NaN	NaN	NaN
top	NaN	NaN	NaN	NaN	NaN
freq	NaN	NaN	NaN	NaN	NaN
mean	6.017037e+01	2.081027e+03	1.432369e+00	2.217995e+02	5.364687e+04
std	4.760673e+01	6.093760e+02	1.323083e+00	3.495668e+02	4.992757e+04
min	2.000000e+01	1.120000e+02	0.000000e+00	3.300000e+01	-1.000000e+00
25%	5.000000e+01	1.800000e+03	0.000000e+00	3.500000e+01	-1.000000e+00
50%	5.000000e+01	2.299000e+03	2.000000e+00	3.900000e+01	1.000550e+05
75%	5.000000e+01	2.506000e+03	3.000000e+00	1.690000e+02	1.000940e+05
max	1.024000e+03	2.729000e+03	3.000000e+00	1.959000e+03	1.002480e+05

	C21
count	3.199109e+06
unique	NaN
top	NaN
freq	NaN
mean	8.052746e+01
std	6.698060e+01
min	1.000000e+00
25%	2.300000e+01
50%	6.100000e+01
75%	9.100000e+01
max	2.530000e+02

[11 rows x 24 columns]

```
[ ]: ## concat all data to explore
      # df = pd.concat([train, test])
      ## transform into a datatable just in case of speed limit
      # all_dt = dt.Frame(df)
```

## 2.3 Categorical Data Encoding

Category with names: \* id: drop \* site categories: \* site\_id: base 5 encoding \* site\_domain: base 5 encoding \* site\_category: base 10 encoding \* app cats: \* app\_id: base 5 encoding \* app\_domain: one-hot encoding \* app\_category: base 10 encoding \* device cats: \* device\_id: one-hot encoding \* device\_ip: drop \* device\_model: base 10 encoding

Category without names: \* C1: stay same \* C14: base 5 encoding \* C15: stay same \* C16: stay same \* C17: base 5 encoding \* C18: stay same \* C19: base 5 encoding \* C20: base 5 encoding \* C21: base 5 encoding

```
[ ]: # drop id in column
del train_3['id']
del test['id']
```

```
[ ]: # define a function to transform skewed categorical value based on frequency
def categorical_replace(train_data, test_data, column, pct = 0.01):
    """
    train_data: train dataset to input
    test_data: test to input
    column: column name string to input
    pct: transform frequency treshold, default 0.01
    """
    cond = train_data[column].value_counts(normalize = True) > pct
    non_others = cond[cond].index # define a list to save main category

    train_data['temp'] = 'other'
    train_data.loc[train_data[column].isin(non_others), 'temp'] = ''
    train_data[column]
    train_data[column] = train_data['temp'].values
    del train_data['temp']
    print("Train Test Replace Finished!")

    test_data['temp'] = 'other'
    test_data.loc[test_data[column].isin(non_others), 'temp'] = test_data[column]
    test_data[column] = test_data['temp'].values
    del test_data['temp']
    print("Test data replace finished!")
```

```
[ ]: # define function to draw category distribution
def hist_bar_cat(data, column):
    categories = data[column].value_counts().index.astype('str')
    counts = data[column].value_counts().values
    plt.bar(categories, counts, width=0.5)
    plt.title('Distribution for column {}'.format(column))
```

### 2.3.1 Site Categories

site\_id

```
[ ]: # check the number of unique values for each column
train_3['site_id'].nunique()
```

```
[ ]: 3278
```

```
[ ]: train_3['site_id'].value_counts(normalize = True).skew()
# as shown below, the skewness it is quite high
```

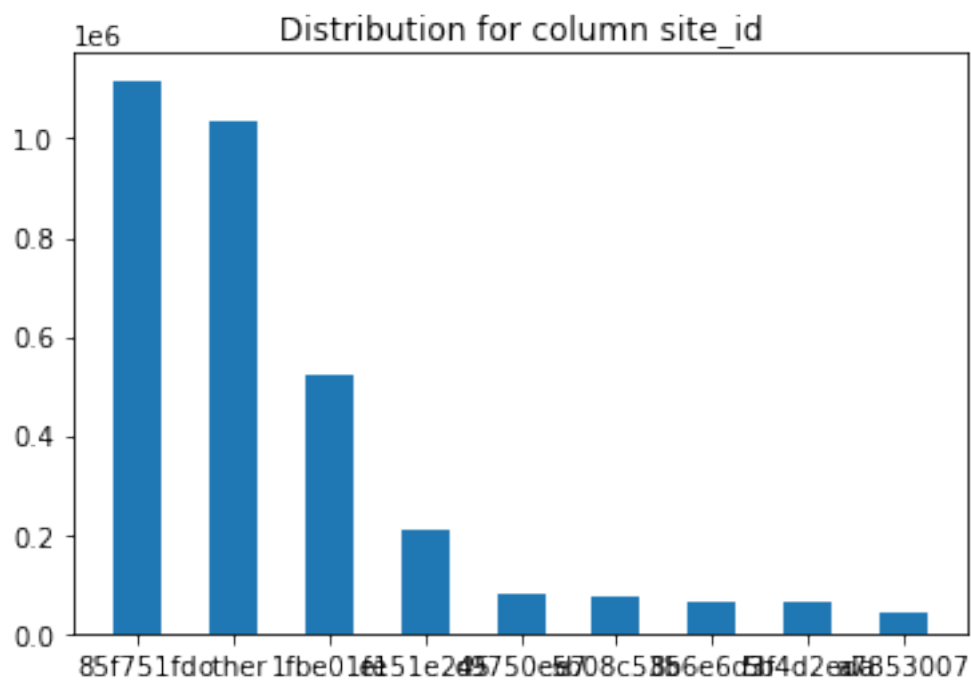
```
[ ]: 43.94039373056989
```

```
[ ]: categorical_replace(train_3, test, 'site_id')
```

Train Test Replace Finished!

Test data replace finished!

```
[ ]: hist_bar_cat(train_3, "site_id")
# much better
```



site\_domain

```
[ ]: train_3['site_domain'].nunique()
# check non-dup value
```

```
[ ]: 4010
```

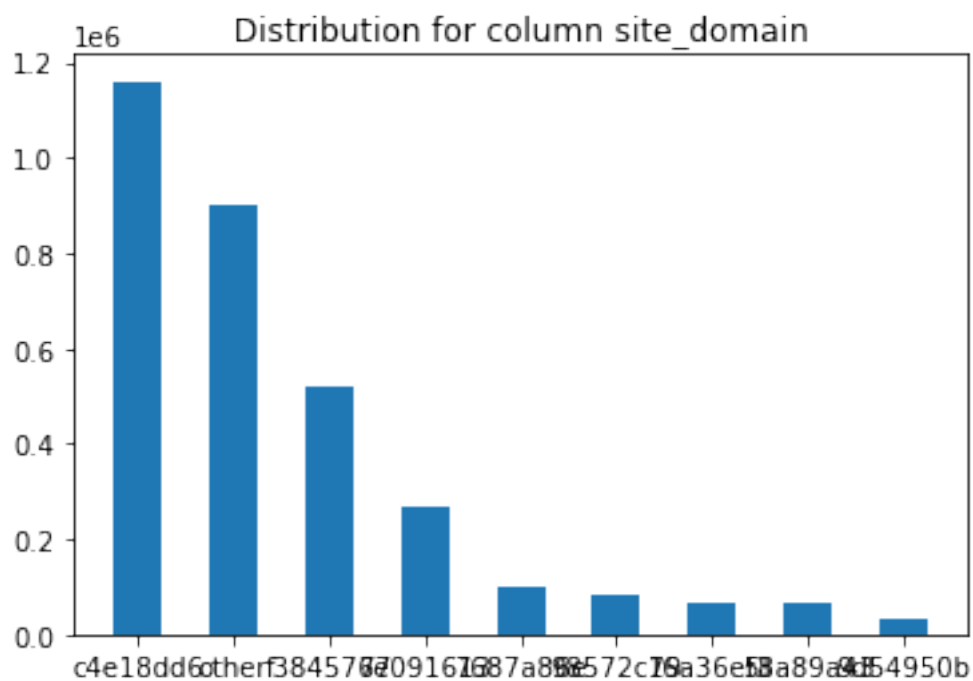
```
[ ]: train_3['site_domain'].value_counts(normalize = True).skew()
# as shown below, the skewness it is quite high
```

```
[ ]: 48.12296789950277
```

```
[ ]: categorical_replace(train_3, test, 'site_domain')
```

Train Test Replace Finished!  
Test data replace finished!

```
[ ]: hist_bar_cat(train_3, "site_domain")
# much better
```



site\_category Use Label Encoder Later

```
[ ]: train_3['site_category'].nunique()
```

```
[ ]: 23
```

### 2.3.2 App Categories

app\_id

```
[ ]: train_3['app_id'].nunique()
```

```
[ ]: 4586
```

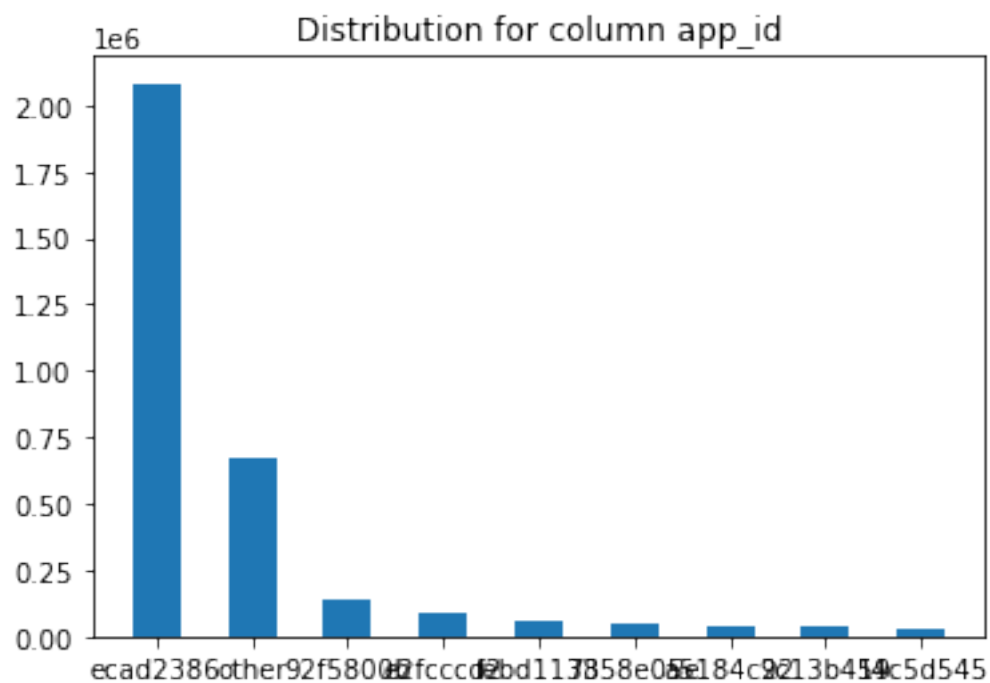
```
[ ]: train_3['app_id'].value_counts(normalize = True).skew()  
# as shown below, the skewness it is quite high
```

```
[ ]: 66.7564684913619
```

```
[ ]: categorical_replace(train_3, test, 'app_id')  
hist_bar_cat(train_3, "app_id")
```

Train Test Replace Finished!

Test data replace finished!



app\_domain

```
[ ]: train_3['app_domain'].nunique()
```

```
[ ]: 294
```

```
[ ]: train_3['app_domain'].value_counts(normalize = True).skew()  
# as shown below, the skewness it is quite high
```

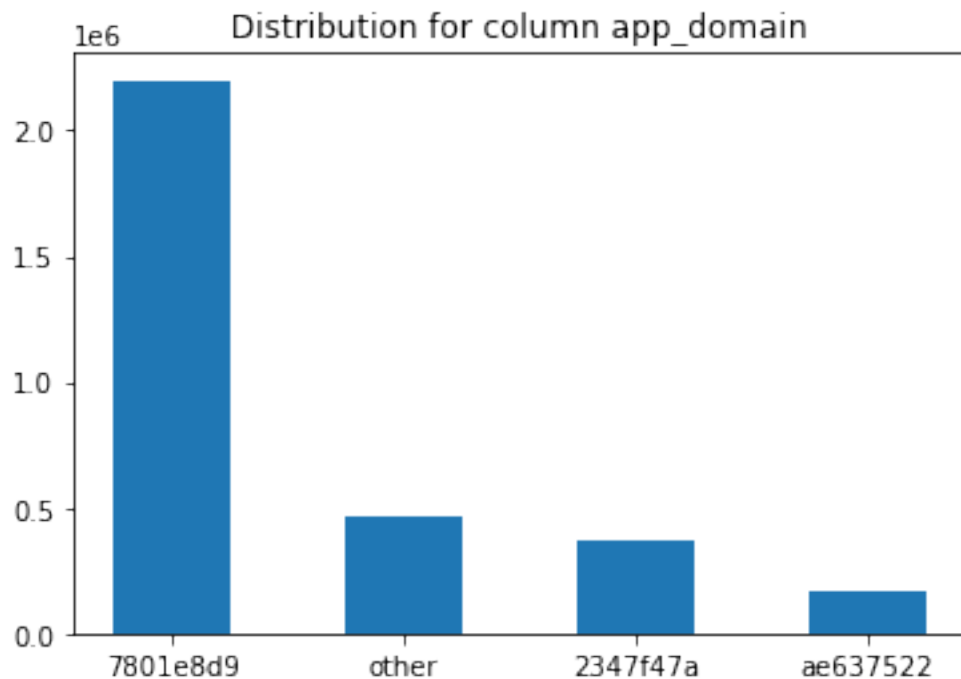
```
[ ]: 16.285788826071837
```



```
[ ]: categorical_replace(train_3, test, 'app_domain', pct = 0.05)
hist_bar_cat(train_3, "app_domain")
```

Train Test Replace Finished!

Test data replace finished!



**app\_category** Label Encode Later

```
[ ]: train_3['app_category'].nunique()
```

```
[ ]: 27
```

### 2.3.3 Device Categoris

**device\_id**

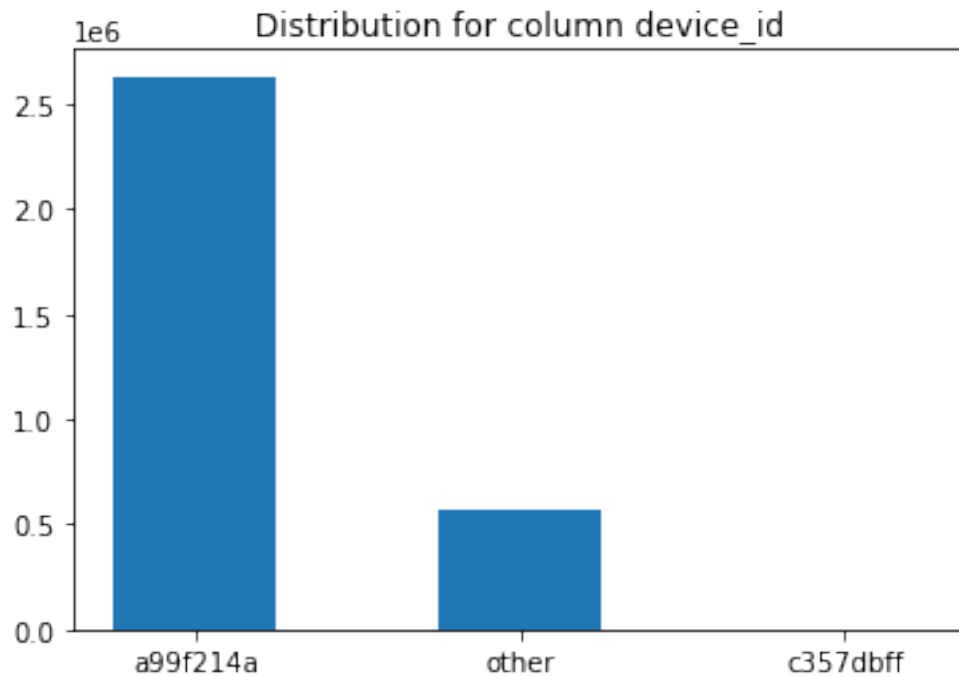
```
[ ]: train_3['device_id'].value_counts(normalize = True).skew()
# as shown below, the skewness it is quite high
```

```
[ ]: 646.6309681657631
```

```
[ ]: categorical_replace(train_3, test, 'device_id', pct = 0.0005)
hist_bar_cat(train_3, "device_id")
```

Train Test Replace Finished!

Test data replace finished!



device\_ip

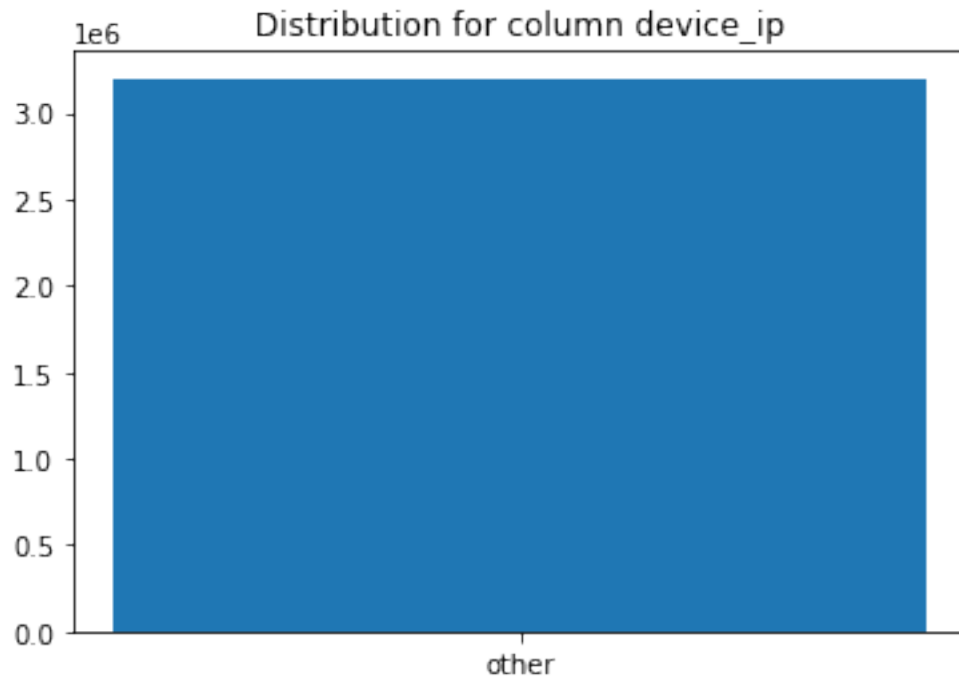
```
[ ]: train_3['device_ip'].value_counts(normalize = True).skew()  
# as shown below, the skewness it is quite high
```

```
[ ]: 252.18416764934165
```

```
[ ]: categorical_replace(train_3, test, 'device_ip')  
hist_bar_cat(train_3, "device_ip")
```

Train Test Replace Finished!

Test data replace finished!



This variable is too skew that we decide to drop it.

```
[ ]: del train_3['device_ip']  
     del test['device_ip']
```

device\_model

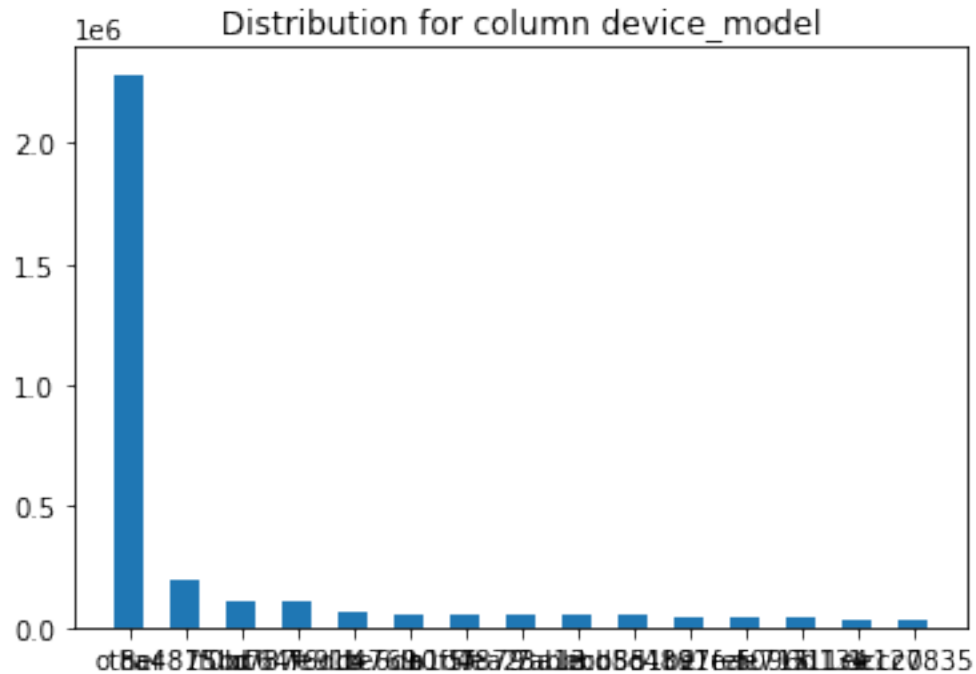
```
[ ]: train_3['device_model'].value_counts(normalize = True).skew()  
     # as shown below, the skewness it is quite high
```

```
[ ]: 25.976421186589164
```

```
[ ]: categorical_replace(train_3, test, 'device_model')  
     hist_bar_cat(train_3, "device_model")
```

Train Test Replace Finished!

Test data replace finished!



### 2.3.4 Anonymized Categorical Variables

C14

```
[ ]: train_3['C14'].nunique()
```

```
[ ]: 2284
```

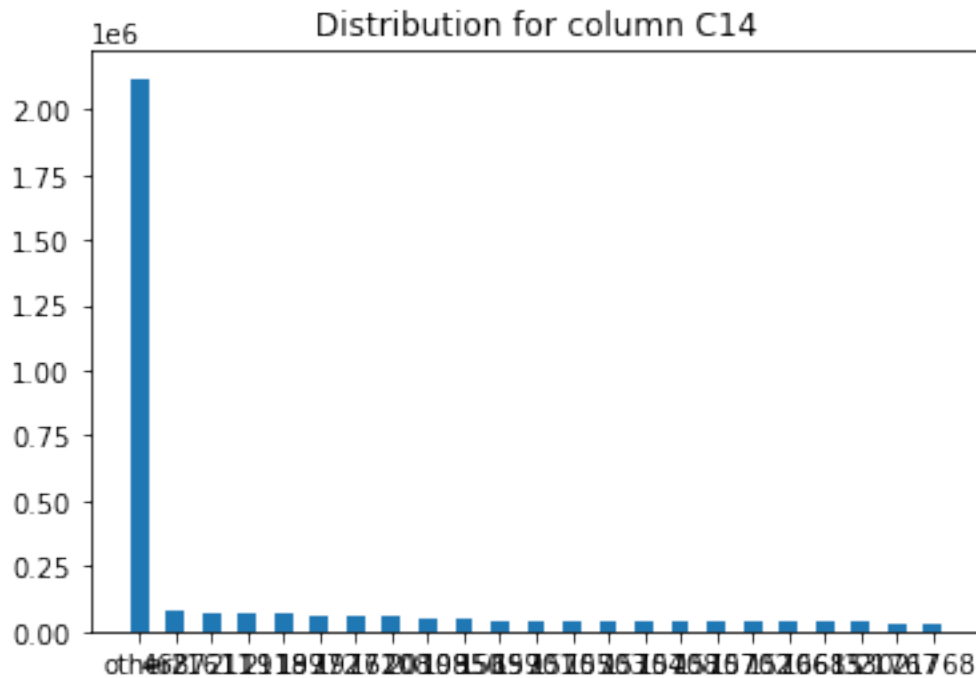
```
[ ]: train_3['C14'].value_counts(normalize = True).skew()
# as shown below, the skewness it is quite high
```

```
[ ]: 7.99323307497523
```

```
[ ]: categorical_replace(train_3, test, 'C14', pct = 0.01)
hist_bar_cat(train_3, "C14")
```

Train Test Replace Finished!

Test data replace finished!



C17

```
[ ]: train_3['C17'].nunique()
```

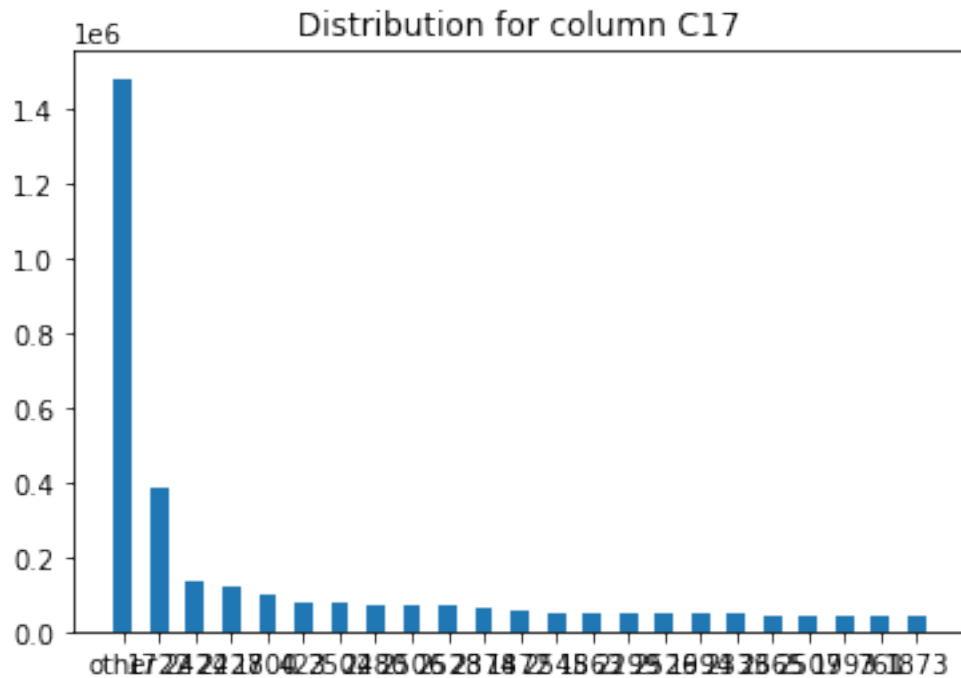
```
[ ]: 405
```

```
[ ]: train_3['C17'].value_counts(normalize = True).skew()
# as shown below, the skewness it is quite high
```

```
[ ]: 10.171913349163981
```

```
[ ]: categorical_replace(train_3, test, 'C17')
hist_bar_cat(train_3, "C17")
```

Train Test Replace Finished!  
Test data replace finished!



C19

```
[ ]: train_3['C19'].nunique()
```

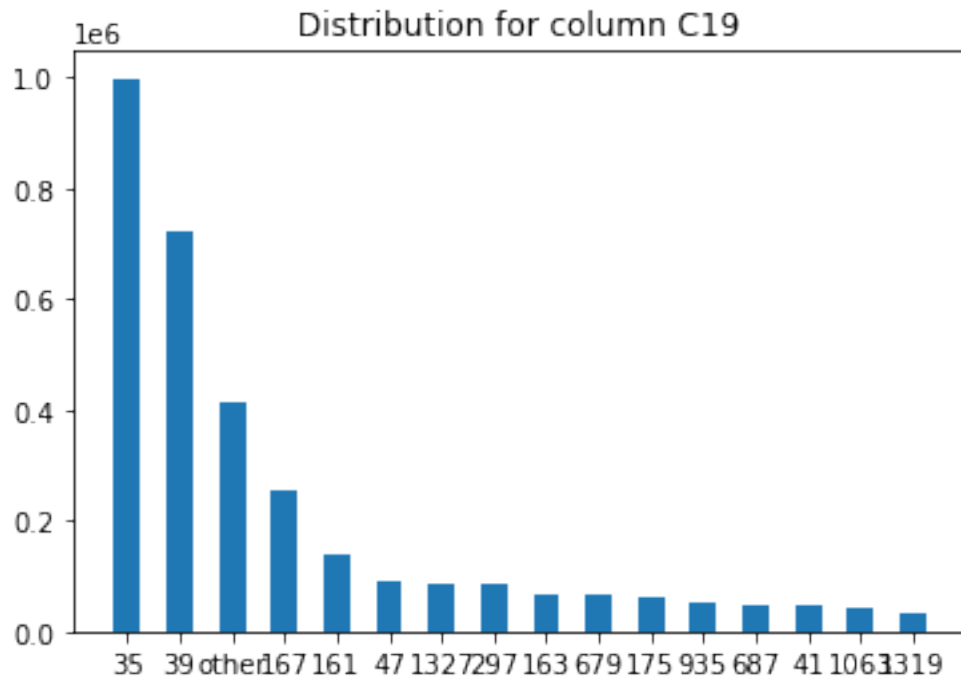
```
[ ]: 65
```

```
[ ]: train_3['C19'].value_counts(normalize = True).skew()
# as shown below, the skewness it is quite high
```

```
[ ]: 5.252812769397987
```

```
[ ]: categorical_replace(train_3, test, 'C19')
hist_bar_cat(train_3, "C19")
```

Train Test Replace Finished!  
Test data replace finished!



## C20

```
[ ]: train_3['C20'].nunique()
```

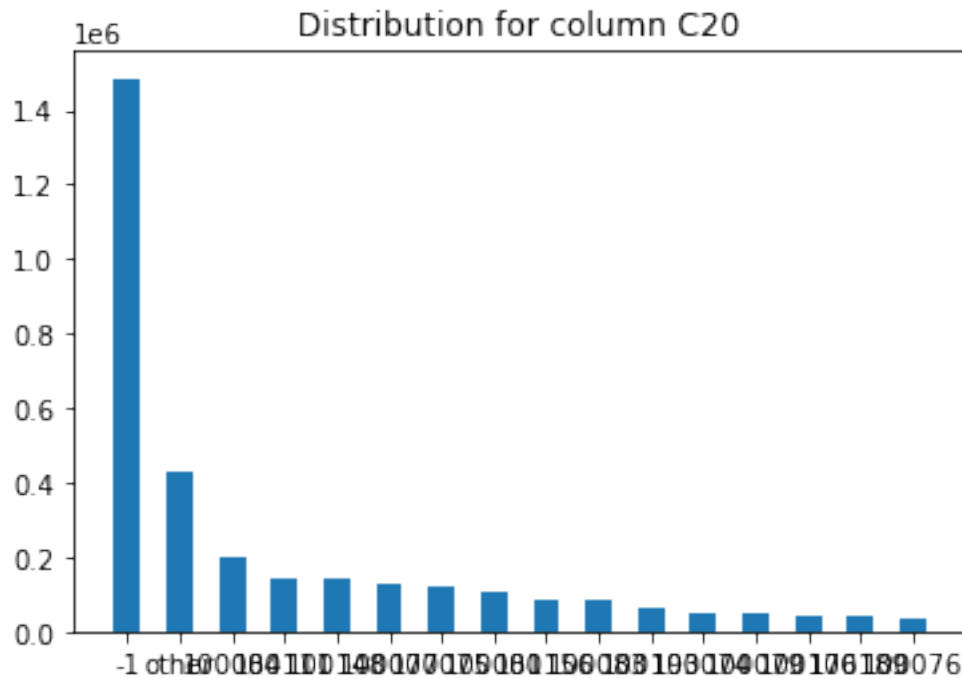
```
[ ]: 168
```

```
[ ]: train_3['C20'].value_counts(normalize = True).skew()
# as shown below, the skewness it is quite high
```

```
[ ]: 11.87592669177848
```

```
[ ]: categorical_replace(train_3, test, 'C20')
hist_bar_cat(train_3, "C20")
```

Train Test Replace Finished!  
Test data replace finished!



C21

```
[ ]: train_3['C21'].nunique()
```

```
[ ]: 55
```

```
[ ]: train_3['C21'].value_counts(normalize = True).skew()
# as shown below, the skewness it is not quite high
```

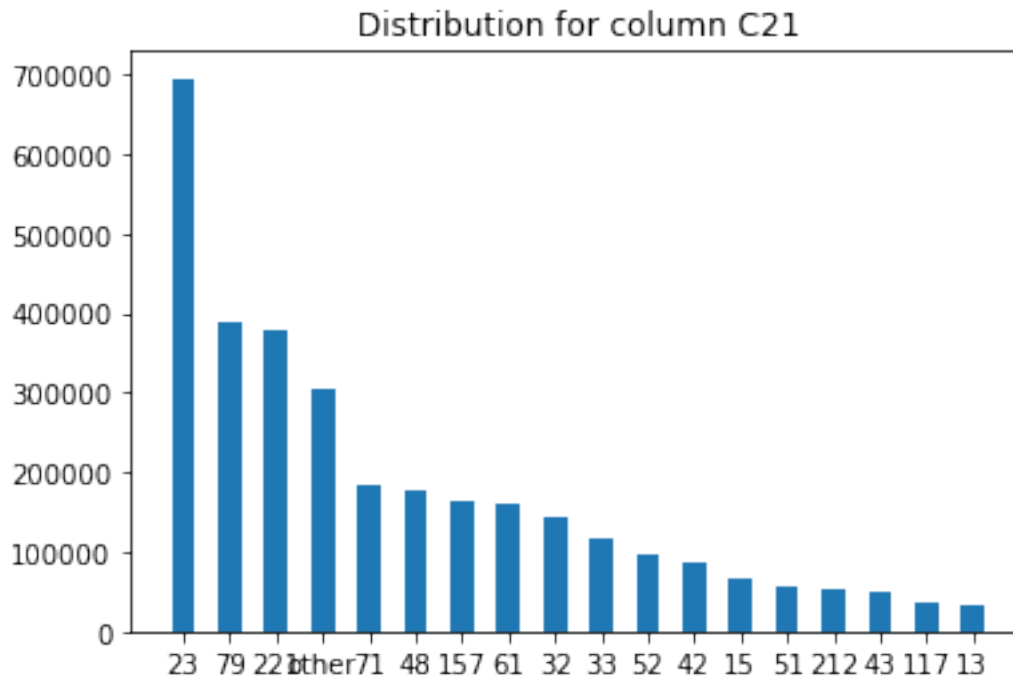
```
[ ]: 3.6204617840011357
```

```
[ ]: categorical_replace(train_3, test, 'C21')
```

Train Test Replace Finished!  
Test data replace finished!

```
[ ]: hist_bar_cat(train_3, "C21")
```





### 2.3.5 Encoding

#### One-hot

```
[ ]: # one-hot encoding: app_domain, device_id
encoder=ce.
↳ OneHotEncoder(cols=['app_domain','device_id'],handle_unknown='return_nan',return_df=True,us
one_hot_encoder = encoder.fit(train_3)
```

```
[ ]: train_3 = one_hot_encoder.transform(train_3)
```

```
[ ]: test = one_hot_encoder.transform(test)
```

#### Base 5

```
[ ]: # base 5 encoding: site_id, site_domain, app_id, C14, C17, C19, C20, C21
encoder1 = ce.
↳ BaseNEncoder(cols=['site_id','site_domain','app_id','C14','C17','C19','C20','C21'],u
↳ return_df=True, base=5)
base_5_encoder = encoder1.fit(train_3)
```

```
[ ]: train_3 = base_5_encoder.transform(train_3)
```

```
[ ]: test = base_5_encoder.transform(test)
```

## Base 10

```
[ ]: # base 10 encoding: site_category, app_category, device_model
encoder2 = ce.
    ↳BaseNEncoder(cols=['site_category', 'app_category', 'device_model'],
    ↳return_df=True, base=10)
base_10_encoder = encoder2.fit(train_3)

[ ]: train_3 = base_10_encoder.transform(train_3)

[ ]: test = base_10_encoder.transform(test)
```

## 2.4 Numerical Variables

- hour: detail shown below
- banner\_pos: encoded by professor
- device\_type: encoded by professor
- device\_conn\_type: encoded by professor

### 2.4.1 Hour

- Day: week number
- Hour:
  - 1: 00-06
  - 2: 07-12
  - 3: 13-18
  - 4: 19-24

```
[ ]: splitat = 6
train_3['day'], train_3['time'] = train_3['hour'].astype('str').str[:splitat],
    ↳train_3['hour'].astype('str').str[splitat:].astype('int')

[ ]: # time of day
def add_time_of_day(data):
    data['day'], data['time'] = data['hour'].astype('str').str[:splitat],
    ↳data['hour'].astype('str').str[splitat:].astype('int')
    conditions = [(data.time <=6),
                  (data.time > 6) & (data.time <= 12),
                  (data.time > 12) & (data.time <= 18),
                  (data.time >18)
                  ]
    values = [1, 2, 3, 4]
    data['time_of_day'] = np.select(conditions, values, 0)

[ ]: add_time_of_day(train_3)
add_time_of_day(test)

[ ]: train_3[['day', 'time', 'hour', 'time_of_day']].head(5)
```

```
[ ]:      day  time      hour  time_of_day
5616589  141022    10  14102210          2
31825749 141029    22  14102922          4
13548281 141024    13  14102413          3
11682204 141023    22  14102322          4
29186795 141029     4  14102904          1
```

```
[ ]: def add_day_of_week(data):
      data['day'] = pd.to_datetime(data['day'], format="%y%m%d")
      data['day_of_week'] = data['day'].dt.dayofweek
```

```
[ ]: add_day_of_week(train_3)
      add_day_of_week(test)
```

## 2.5 Finalize the data

```
[ ]: train_3.head()
```

```
[ ]:      click      hour    C1  banner_pos  site_id_0  site_id_1  \
5616589      0  14102210  1005           1          0          1
31825749      0  14102922  1005           0          0          2
13548281      1  14102413  1005           0          0          1
11682204      0  14102322  1005           0          0          2
29186795      0  14102904  1005           0          0          1

      site_domain_0  site_domain_1  site_category_0  site_category_1  ...  \
5616589           0           1           0           1  ...
31825749           0           2           0           2  ...
13548281           0           1           0           1  ...
11682204           0           2           0           2  ...
29186795           0           1           0           3  ...

      C19_0  C19_1  C20_0  C20_1  C21_0  C21_1      day  time  \
5616589      0      1      0      1      0      1  2014-10-22    10
31825749      0      2      0      2      0      2  2014-10-29    22
13548281      0      2      0      2      0      3  2014-10-24    13
11682204      0      1      0      3      0      4  2014-10-23    22
29186795      0      3      0      3      0      3  2014-10-29     4

      time_of_day  day_of_week
5616589          2           2
31825749          4           2
13548281          3           4
11682204          4           3
29186795          1           2
```

[5 rows x 42 columns]

```
[ ]:
[ ]: x_col = ['C1', 'banner_pos', 'site_id_0', 'site_id_1',
            'site_domain_0', 'site_domain_1', 'site_category_0', 'site_category_1',
            'app_id_0', 'app_id_1', 'app_domain_7801e8d9', 'app_domain_ae637522',
            'app_domain_2347f47a', 'app_domain_other', 'app_category_0',
            'app_category_1', 'device_id_other', 'device_id_a99f214a',
            'device_id_c357dbff', 'device_model_0', 'device_model_1', 'device_type',
            ↪ 'device_conn_type',
            'C14_0', 'C14_1', 'C15', 'C16', 'C17_0', 'C17_1', 'C18', 'C19_0',
            'C19_1', 'C20_0', 'C20_1', 'C21_0', 'C21_1',
            'time_of_day', 'day_of_week']
y_col = 'click'

[ ]: X_train3 = train_3[x_col]
y_train3 = train_3[y_col]
X_test = test[x_col]
y_pred_naive = test[y_col]
```

### 3 Model Training

```
[ ]: # Split to Sub-train & Validation dataset
X_train, X_val, y_train, y_val = train_test_split(X_train3, y_train3,
            ↪ test_size=0.2, random_state=42)
```

#### 3.1 Logistic Regression

```
[ ]: from sklearn.linear_model import LogisticRegression

[ ]: lr = LogisticRegression(random_state=42, n_jobs=-1, penalty="l2").fit(X_train,
            ↪ y_train)
```

/Users/rebecca/opt/anaconda3/lib/python3.9/site-packages/sklearn/linear\_model/\_logistic.py:444: ConvergenceWarning: lbfgs failed to converge (status=1):  
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max\_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

[https://scikit-learn.org/stable/modules/linear\\_model.html#logistic-regression](https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)

```
n_iter_i = _check_optimize_result(
```

```
[ ]: y_pred = lr.predict_proba(X_val)
```

```
[ ]: # print log loss of the validation data
log_loss(y_val, y_pred)
```

```
[ ]: 0.4360344851617682
```

Since the log-loss is quite high, LR is not a great model to be chosen as a final step.

## 3.2 Random Forest

### 3.2.1 TuneGridsearchCV

```
[ ]: from sklearn.ensemble import RandomForestClassifier
```

```
[ ]: rfc=RandomForestClassifier(random_state=42)
```

Random Search using random forest

```
[ ]: rs_space={'max_depth': [20, 30],
            'max_features': ['sqrt', 'log2'],
            'min_samples_split': [50, 100]
        }
```

```
[ ]: # param_grid = {
#     'n_estimators': [100, 200, 300, 400],
#     'max_features': ['sqrt', 'log2'],
#     'min_samples_split': [2, 10, 100],
#     'max_depth': [10, 20, 30],
# }
# gs_rfc = GridSearchCV(estimator=rfc,
#     param_grid=param_grid,
#     cv=3,
#     scoring='neg_log_loss',
#     n_jobs=-1,
#     verbose=3)
# gs_rfc.fit(X_train, y_train)
```

```
[ ]: from tune_sklearn import TuneGridSearchCV
from sklearn.model_selection import RandomizedSearchCV
```

```
[ ]: gs_rfc = TuneGridSearchCV(rfc, rs_space,
                             scoring='neg_log_loss',
                             n_jobs=-1,
                             cv=3,
                             max_iters=10,
                             early_stopping = True,
                             verbose=2)
```

```
[ ]: gs_rfc.fit(X_train, y_train)
```

<IPython.core.display.HTML object>

(raylet) Spilled 2460 MiB, 10 objects, write throughput 129 MiB/s.  
Set RAY\_verbose\_spill\_logs=0 to disable this message.

<IPython.core.display.HTML object>

2022-12-05 20:28:43,216 INFO tune.py:777 -- Total run time: 2454.85 seconds  
(2454.71 seconds for the tuning loop).

```
[ ]: TuneGridSearchCV(cv=3, early_stopping=True,
                      estimator=RandomForestClassifier(random_state=42),
                      max_iters=10, mode='max', n_jobs=-1,
                      param_grid={'max_depth': [20, 30],
                                   'max_features': ['sqrt', 'log2'],
                                   'min_samples_split': [50, 100]},
                      scoring='neg_log_loss', sk_n_jobs=1, verbose=2)
```

```
[ ]: # rfc_random = RandomizedSearchCV(rfc, rs_space,
#                                     scoring='neg_log_loss',
#                                     n_jobs=-1,
#                                     cv=2,
#                                     n_iter=10,
#                                     verbose=30)
# rfc_random.fit(X_train, y_train)
```

```
[ ]: print("Parameter: ", gs_rfc.best_params_)
print("Non-nested LogLoss: ", gs_rfc.best_score_)
print("Best Estimator: ", gs_rfc.best_estimator_)
```

Parameter: {'max\_depth': 20, 'max\_features': 'sqrt', 'min\_samples\_split': 100,  
'n\_estimators': 10}

Non-nested LogLoss: -0.40525057984573404

Best Estimator: RandomForestClassifier(max\_depth=20, min\_samples\_split=100,  
n\_estimators=10,  
random\_state=42)

```
[ ]: rf = RandomForestClassifier(n_estimators = 10,
                                random_state = 42,
                                min_samples_split=100,
                                max_depth=20,
                                n_jobs=-1)
```

```
[ ]: rf.fit(X_train, y_train)
```

```
[ ]: RandomForestClassifier(max_depth=20, min_samples_split=100, n_estimators=10,
                             n_jobs=-1, random_state=42)
```

```
[ ]: y_pred_rf = rf.predict_proba(X_val)
      log_loss(y_val, y_pred_rf)
```

```
[ ]: 0.4047542503530167
```

### 3.3 Neural Net

```
[ ]: # initialize nn
nnc = Sequential()
# add input layer
nnc.add(Dense(10, kernel_regularizer = regularizers.l2(0.003),
    ↪kernel_initializer=initializers.RandomNormal(stddev=0.01),
    ↪activation='relu', input_shape = (X_train.shape[1],), use_bias=True))
nnc.add(Dropout(0.4))
# add hidden layer
nnc.add(Dense(10, kernel_regularizer = regularizers.l2(0.003),
    ↪kernel_initializer=initializers.RandomNormal(stddev=0.01),
    ↪activation='relu'))
nnc.add(Dropout(0.3))
nnc.add(Dense(units = 6, activation = 'sigmoid', use_bias = True))
# add output layer
nnc.add(Dense(1, activation = 'sigmoid', use_bias = True))
nnc.compile(optimizer = Adam(learning_rate=1e-3), loss = 'binary_crossentropy',
    ↪metrics = ['accuracy'])
```

2022-12-05 17:21:14.358862: I tensorflow/core/platform/cpu\_feature\_guard.cc:193] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: SSE4.1 SSE4.2 AVX AVX2 FMA To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.

```
[ ]: nnc.fit(X_train, y_train, epochs=10, batch_size = 512, validation_data =
    ↪(X_val, y_val))
```

```
Epoch 1/10
4374/4374 [=====] - 8s 2ms/step - loss: 0.4494 -
accuracy: 0.8302 - val_loss: 0.4479 - val_accuracy: 0.8303
Epoch 2/10
4374/4374 [=====] - 6s 1ms/step - loss: 0.4493 -
accuracy: 0.8302 - val_loss: 0.4478 - val_accuracy: 0.8303
Epoch 3/10
4374/4374 [=====] - 6s 1ms/step - loss: 0.4493 -
accuracy: 0.8302 - val_loss: 0.4485 - val_accuracy: 0.8303
Epoch 4/10
4374/4374 [=====] - 6s 1ms/step - loss: 0.4504 -
accuracy: 0.8302 - val_loss: 0.4483 - val_accuracy: 0.8303
```

```
Epoch 5/10
4374/4374 [=====] - 6s 1ms/step - loss: 0.4516 -
accuracy: 0.8302 - val_loss: 0.4501 - val_accuracy: 0.8303
Epoch 6/10
4374/4374 [=====] - 6s 1ms/step - loss: 0.4520 -
accuracy: 0.8302 - val_loss: 0.4498 - val_accuracy: 0.8303
Epoch 7/10
4374/4374 [=====] - 6s 1ms/step - loss: 0.4532 -
accuracy: 0.8302 - val_loss: 0.4534 - val_accuracy: 0.8303
Epoch 8/10
4374/4374 [=====] - 6s 1ms/step - loss: 0.4552 -
accuracy: 0.8302 - val_loss: 0.4555 - val_accuracy: 0.8303
Epoch 9/10
4374/4374 [=====] - 6s 1ms/step - loss: 0.4555 -
accuracy: 0.8302 - val_loss: 0.4555 - val_accuracy: 0.8303
Epoch 10/10
4374/4374 [=====] - 6s 1ms/step - loss: 0.4555 -
accuracy: 0.8302 - val_loss: 0.4555 - val_accuracy: 0.8303
```

```
[ ]: <keras.callbacks.History at 0x7fc0a83c37c0>
```

```
[ ]: y_pred_nn = nnc.predict(X_val)
log_loss(y_val, y_pred_nn)
```

```
29992/29992 [=====] - 17s 570us/step
```

```
[ ]: 0.4554982736349612
```

### 3.4 LightGBM

```
[ ]: lgb = LGBMClassifier(random_state=42)
```

```
[ ]: lgb.fit(X_train,y_train,verbose=3,eval_metric='logloss')
```

```
[ ]: LGBMClassifier(random_state=42)
```

```
[ ]: parameters = {
    'learning_rate': [0.01,0.05,0.09,0.1],
    'num_leaves': [31,250,300],
}

gbm = LGBMClassifier(objective='binary',
                     n_jobs=-1,
                     metric = 'binary_logloss',
                     boosting_type='gbdt',
                     cat_smooth= 35)
```



```
gs_lgb = GridSearchCV(gbm, param_grid=parameters, scoring='neg_log_loss', cv=3,
↳ verbose = 1)
```

```
[ ]: gs_lgb.fit(X_train, y_train)
```

Fitting 3 folds for each of 12 candidates, totalling 36 fits

```
[ ]: GridSearchCV(cv=3,
    estimator=LGBMClassifier(cat_smooth=35, metric='binary_logloss',
        objective='binary'),
    param_grid={'learning_rate': [0.01, 0.05, 0.09, 0.1],
        'num_leaves': [31, 250, 300]},
    scoring='neg_log_loss', verbose=1)
```

```
[ ]: print('best parameter:{0}'.format(gs_lgb.best_params_))
```

best parameter: {'learning\_rate': 0.1, 'num\_leaves': 300}

```
[ ]: lgb = LGBMClassifier(learning_rate=0.1, num_leaves = 300, random_state=42, metric_
↳ 'binary_logloss', cat_smooth= 35).\
    fit(X_train, y_train, eval_metric='logloss')
```

```
[ ]: y_pred_lgb=lgb.predict_proba(X_val)[:, 1]
```

```
[ ]: log_loss(y_val, y_pred_lgb)
```

```
[ ]: 0.40189174104048725
```

### 3.5 CatBoost

```
[ ]: from catboost import CatBoostClassifier, Pool
```

```
[ ]: ctb = CatBoostClassifier(random_seed=42)
```

```
[ ]: ctb.fit(X_train, y_train)
```

Learning rate set to 0.277612

0:	learn: 0.5546360	total: 286ms	remaining: 4m 46s
1:	learn: 0.4901436	total: 466ms	remaining: 3m 52s
2:	learn: 0.4580588	total: 677ms	remaining: 3m 45s
3:	learn: 0.4419956	total: 875ms	remaining: 3m 37s
4:	learn: 0.4337565	total: 1.09s	remaining: 3m 37s
5:	learn: 0.4287606	total: 1.29s	remaining: 3m 34s
6:	learn: 0.4262102	total: 1.58s	remaining: 3m 44s
7:	learn: 0.4243884	total: 1.75s	remaining: 3m 37s
8:	learn: 0.4233272	total: 1.94s	remaining: 3m 33s
9:	learn: 0.4221519	total: 2.12s	remaining: 3m 30s

10:	learn: 0.4211780	total: 2.31s	remaining: 3m 28s
11:	learn: 0.4200521	total: 2.5s	remaining: 3m 25s
12:	learn: 0.4192943	total: 2.71s	remaining: 3m 25s
13:	learn: 0.4189905	total: 2.9s	remaining: 3m 24s
14:	learn: 0.4184677	total: 3.09s	remaining: 3m 23s
15:	learn: 0.4181608	total: 3.27s	remaining: 3m 20s
16:	learn: 0.4178235	total: 3.46s	remaining: 3m 20s
17:	learn: 0.4173590	total: 3.65s	remaining: 3m 19s
18:	learn: 0.4171426	total: 3.84s	remaining: 3m 18s
19:	learn: 0.4167615	total: 4.05s	remaining: 3m 18s
20:	learn: 0.4164154	total: 4.28s	remaining: 3m 19s
21:	learn: 0.4161804	total: 4.51s	remaining: 3m 20s
22:	learn: 0.4159316	total: 4.74s	remaining: 3m 21s
23:	learn: 0.4157913	total: 4.94s	remaining: 3m 21s
24:	learn: 0.4156897	total: 5.16s	remaining: 3m 21s
25:	learn: 0.4155136	total: 5.36s	remaining: 3m 20s
26:	learn: 0.4153259	total: 5.6s	remaining: 3m 21s
27:	learn: 0.4150533	total: 5.84s	remaining: 3m 22s
28:	learn: 0.4148490	total: 6.08s	remaining: 3m 23s
29:	learn: 0.4146303	total: 6.3s	remaining: 3m 23s
30:	learn: 0.4145423	total: 6.52s	remaining: 3m 23s
31:	learn: 0.4143679	total: 6.77s	remaining: 3m 24s
32:	learn: 0.4141760	total: 7.02s	remaining: 3m 25s
33:	learn: 0.4139500	total: 7.25s	remaining: 3m 26s
34:	learn: 0.4138078	total: 7.48s	remaining: 3m 26s
35:	learn: 0.4137127	total: 7.71s	remaining: 3m 26s
36:	learn: 0.4135380	total: 7.98s	remaining: 3m 27s
37:	learn: 0.4133628	total: 8.31s	remaining: 3m 30s
38:	learn: 0.4132991	total: 8.51s	remaining: 3m 29s
39:	learn: 0.4132065	total: 8.72s	remaining: 3m 29s
40:	learn: 0.4130997	total: 8.93s	remaining: 3m 28s
41:	learn: 0.4129088	total: 9.16s	remaining: 3m 28s
42:	learn: 0.4127877	total: 9.38s	remaining: 3m 28s
43:	learn: 0.4127018	total: 9.65s	remaining: 3m 29s
44:	learn: 0.4126035	total: 9.86s	remaining: 3m 29s
45:	learn: 0.4124666	total: 10.1s	remaining: 3m 28s
46:	learn: 0.4124038	total: 10.3s	remaining: 3m 28s
47:	learn: 0.4122297	total: 10.5s	remaining: 3m 28s
48:	learn: 0.4121711	total: 10.7s	remaining: 3m 28s
49:	learn: 0.4120944	total: 10.9s	remaining: 3m 27s
50:	learn: 0.4120003	total: 11.2s	remaining: 3m 28s
51:	learn: 0.4119157	total: 11.4s	remaining: 3m 28s
52:	learn: 0.4118710	total: 11.6s	remaining: 3m 27s
53:	learn: 0.4117106	total: 11.9s	remaining: 3m 27s
54:	learn: 0.4116283	total: 12.1s	remaining: 3m 28s
55:	learn: 0.4115617	total: 12.4s	remaining: 3m 28s
56:	learn: 0.4115376	total: 12.6s	remaining: 3m 28s
57:	learn: 0.4114261	total: 12.8s	remaining: 3m 28s

58:	learn: 0.4113178	total: 13.1s	remaining: 3m 28s
59:	learn: 0.4112835	total: 13.4s	remaining: 3m 29s
60:	learn: 0.4112375	total: 13.6s	remaining: 3m 29s
61:	learn: 0.4111765	total: 13.8s	remaining: 3m 29s
62:	learn: 0.4110480	total: 14.2s	remaining: 3m 30s
63:	learn: 0.4110022	total: 14.4s	remaining: 3m 31s
64:	learn: 0.4109316	total: 14.9s	remaining: 3m 33s
65:	learn: 0.4108577	total: 15.2s	remaining: 3m 34s
66:	learn: 0.4107676	total: 15.5s	remaining: 3m 35s
67:	learn: 0.4106700	total: 15.8s	remaining: 3m 36s
68:	learn: 0.4105561	total: 16.1s	remaining: 3m 37s
69:	learn: 0.4105163	total: 16.5s	remaining: 3m 38s
70:	learn: 0.4104150	total: 16.8s	remaining: 3m 40s
71:	learn: 0.4103828	total: 17.2s	remaining: 3m 41s
72:	learn: 0.4102453	total: 17.5s	remaining: 3m 42s
73:	learn: 0.4101556	total: 17.8s	remaining: 3m 43s
74:	learn: 0.4101179	total: 18.2s	remaining: 3m 44s
75:	learn: 0.4100612	total: 18.6s	remaining: 3m 45s
76:	learn: 0.4099731	total: 18.9s	remaining: 3m 46s
77:	learn: 0.4098977	total: 19.1s	remaining: 3m 46s
78:	learn: 0.4098373	total: 19.3s	remaining: 3m 45s
79:	learn: 0.4097720	total: 19.5s	remaining: 3m 44s
80:	learn: 0.4097240	total: 19.8s	remaining: 3m 44s
81:	learn: 0.4096137	total: 20s	remaining: 3m 43s
82:	learn: 0.4095649	total: 20.2s	remaining: 3m 42s
83:	learn: 0.4095271	total: 20.4s	remaining: 3m 42s
84:	learn: 0.4094921	total: 20.6s	remaining: 3m 41s
85:	learn: 0.4094402	total: 20.8s	remaining: 3m 40s
86:	learn: 0.4093803	total: 21s	remaining: 3m 40s
87:	learn: 0.4092884	total: 21.2s	remaining: 3m 39s
88:	learn: 0.4092372	total: 21.5s	remaining: 3m 39s
89:	learn: 0.4091567	total: 21.7s	remaining: 3m 39s
90:	learn: 0.4091229	total: 21.9s	remaining: 3m 38s
91:	learn: 0.4091006	total: 22.1s	remaining: 3m 37s
92:	learn: 0.4090737	total: 22.3s	remaining: 3m 37s
93:	learn: 0.4090340	total: 22.5s	remaining: 3m 36s
94:	learn: 0.4089990	total: 22.7s	remaining: 3m 35s
95:	learn: 0.4089723	total: 22.9s	remaining: 3m 35s
96:	learn: 0.4088957	total: 23.1s	remaining: 3m 35s
97:	learn: 0.4088538	total: 23.3s	remaining: 3m 34s
98:	learn: 0.4088058	total: 23.6s	remaining: 3m 34s
99:	learn: 0.4087818	total: 23.8s	remaining: 3m 34s
100:	learn: 0.4087388	total: 24s	remaining: 3m 33s
101:	learn: 0.4086906	total: 24.2s	remaining: 3m 33s
102:	learn: 0.4086492	total: 24.4s	remaining: 3m 32s
103:	learn: 0.4086265	total: 24.6s	remaining: 3m 32s
104:	learn: 0.4085876	total: 24.8s	remaining: 3m 31s
105:	learn: 0.4085429	total: 25s	remaining: 3m 31s

106:	learn: 0.4084917	total: 25.3s	remaining: 3m 30s
107:	learn: 0.4084537	total: 25.5s	remaining: 3m 30s
108:	learn: 0.4084092	total: 25.7s	remaining: 3m 29s
109:	learn: 0.4083451	total: 25.9s	remaining: 3m 29s
110:	learn: 0.4083154	total: 26.1s	remaining: 3m 29s
111:	learn: 0.4082757	total: 26.3s	remaining: 3m 28s
112:	learn: 0.4082531	total: 26.6s	remaining: 3m 28s
113:	learn: 0.4082184	total: 26.8s	remaining: 3m 28s
114:	learn: 0.4081724	total: 27s	remaining: 3m 27s
115:	learn: 0.4081503	total: 27.2s	remaining: 3m 27s
116:	learn: 0.4081357	total: 27.4s	remaining: 3m 26s
117:	learn: 0.4080655	total: 27.6s	remaining: 3m 26s
118:	learn: 0.4080417	total: 27.8s	remaining: 3m 26s
119:	learn: 0.4080179	total: 28s	remaining: 3m 25s
120:	learn: 0.4080007	total: 28.2s	remaining: 3m 25s
121:	learn: 0.4079785	total: 28.4s	remaining: 3m 24s
122:	learn: 0.4079478	total: 28.7s	remaining: 3m 24s
123:	learn: 0.4079144	total: 28.9s	remaining: 3m 23s
124:	learn: 0.4078750	total: 29.1s	remaining: 3m 23s
125:	learn: 0.4078398	total: 29.3s	remaining: 3m 23s
126:	learn: 0.4078000	total: 29.5s	remaining: 3m 22s
127:	learn: 0.4077733	total: 29.7s	remaining: 3m 22s
128:	learn: 0.4077444	total: 30s	remaining: 3m 22s
129:	learn: 0.4076883	total: 30.2s	remaining: 3m 22s
130:	learn: 0.4076662	total: 30.4s	remaining: 3m 21s
131:	learn: 0.4076328	total: 30.7s	remaining: 3m 21s
132:	learn: 0.4076176	total: 30.9s	remaining: 3m 21s
133:	learn: 0.4075565	total: 31.1s	remaining: 3m 21s
134:	learn: 0.4075280	total: 31.3s	remaining: 3m 20s
135:	learn: 0.4074894	total: 31.6s	remaining: 3m 21s
136:	learn: 0.4074472	total: 32s	remaining: 3m 21s
137:	learn: 0.4074319	total: 32.3s	remaining: 3m 21s
138:	learn: 0.4074043	total: 32.5s	remaining: 3m 21s
139:	learn: 0.4073753	total: 32.7s	remaining: 3m 20s
140:	learn: 0.4073439	total: 33s	remaining: 3m 20s
141:	learn: 0.4073318	total: 33.2s	remaining: 3m 20s
142:	learn: 0.4072999	total: 33.4s	remaining: 3m 20s
143:	learn: 0.4072732	total: 33.6s	remaining: 3m 19s
144:	learn: 0.4072630	total: 33.8s	remaining: 3m 19s
145:	learn: 0.4072472	total: 34s	remaining: 3m 19s
146:	learn: 0.4072141	total: 34.2s	remaining: 3m 18s
147:	learn: 0.4071776	total: 34.5s	remaining: 3m 18s
148:	learn: 0.4071652	total: 34.7s	remaining: 3m 18s
149:	learn: 0.4071312	total: 34.9s	remaining: 3m 17s
150:	learn: 0.4071084	total: 35.2s	remaining: 3m 17s
151:	learn: 0.4070794	total: 35.4s	remaining: 3m 17s
152:	learn: 0.4070640	total: 35.6s	remaining: 3m 17s
153:	learn: 0.4070345	total: 35.8s	remaining: 3m 16s

154:	learn: 0.4069977	total: 36.1s	remaining: 3m 16s
155:	learn: 0.4069743	total: 36.3s	remaining: 3m 16s
156:	learn: 0.4069504	total: 36.5s	remaining: 3m 15s
157:	learn: 0.4069286	total: 36.7s	remaining: 3m 15s
158:	learn: 0.4068965	total: 36.9s	remaining: 3m 15s
159:	learn: 0.4068773	total: 37.2s	remaining: 3m 15s
160:	learn: 0.4068513	total: 37.4s	remaining: 3m 14s
161:	learn: 0.4068374	total: 37.6s	remaining: 3m 14s
162:	learn: 0.4068174	total: 37.8s	remaining: 3m 14s
163:	learn: 0.4068034	total: 38s	remaining: 3m 13s
164:	learn: 0.4067877	total: 38.2s	remaining: 3m 13s
165:	learn: 0.4067633	total: 38.4s	remaining: 3m 13s
166:	learn: 0.4067357	total: 38.6s	remaining: 3m 12s
167:	learn: 0.4067110	total: 38.9s	remaining: 3m 12s
168:	learn: 0.4066987	total: 39.1s	remaining: 3m 12s
169:	learn: 0.4066886	total: 39.3s	remaining: 3m 11s
170:	learn: 0.4066533	total: 39.5s	remaining: 3m 11s
171:	learn: 0.4066345	total: 39.8s	remaining: 3m 11s
172:	learn: 0.4066089	total: 40s	remaining: 3m 11s
173:	learn: 0.4065808	total: 40.2s	remaining: 3m 10s
174:	learn: 0.4065537	total: 40.4s	remaining: 3m 10s
175:	learn: 0.4065433	total: 40.6s	remaining: 3m 10s
176:	learn: 0.4065240	total: 40.9s	remaining: 3m 9s
177:	learn: 0.4064885	total: 41.1s	remaining: 3m 9s
178:	learn: 0.4064755	total: 41.3s	remaining: 3m 9s
179:	learn: 0.4064459	total: 41.5s	remaining: 3m 9s
180:	learn: 0.4064299	total: 41.7s	remaining: 3m 8s
181:	learn: 0.4064031	total: 42s	remaining: 3m 8s
182:	learn: 0.4063876	total: 42.2s	remaining: 3m 8s
183:	learn: 0.4063384	total: 42.4s	remaining: 3m 7s
184:	learn: 0.4063215	total: 42.6s	remaining: 3m 7s
185:	learn: 0.4062923	total: 42.8s	remaining: 3m 7s
186:	learn: 0.4062749	total: 43s	remaining: 3m 7s
187:	learn: 0.4062672	total: 43.2s	remaining: 3m 6s
188:	learn: 0.4062549	total: 43.5s	remaining: 3m 6s
189:	learn: 0.4062397	total: 43.7s	remaining: 3m 6s
190:	learn: 0.4062136	total: 43.9s	remaining: 3m 5s
191:	learn: 0.4062005	total: 44.1s	remaining: 3m 5s
192:	learn: 0.4061774	total: 44.3s	remaining: 3m 5s
193:	learn: 0.4061610	total: 44.6s	remaining: 3m 5s
194:	learn: 0.4061446	total: 44.8s	remaining: 3m 5s
195:	learn: 0.4061208	total: 45s	remaining: 3m 4s
196:	learn: 0.4060996	total: 45.3s	remaining: 3m 4s
197:	learn: 0.4060861	total: 45.5s	remaining: 3m 4s
198:	learn: 0.4060670	total: 45.7s	remaining: 3m 4s
199:	learn: 0.4060516	total: 45.9s	remaining: 3m 3s
200:	learn: 0.4060291	total: 46.2s	remaining: 3m 3s
201:	learn: 0.4060139	total: 46.4s	remaining: 3m 3s

202:	learn: 0.4059953	total: 46.6s	remaining: 3m 2s
203:	learn: 0.4059759	total: 46.8s	remaining: 3m 2s
204:	learn: 0.4059664	total: 47s	remaining: 3m 2s
205:	learn: 0.4059429	total: 47.2s	remaining: 3m 2s
206:	learn: 0.4059226	total: 47.4s	remaining: 3m 1s
207:	learn: 0.4059115	total: 47.7s	remaining: 3m 1s
208:	learn: 0.4058949	total: 47.9s	remaining: 3m 1s
209:	learn: 0.4058798	total: 48.1s	remaining: 3m
210:	learn: 0.4058673	total: 48.3s	remaining: 3m
211:	learn: 0.4058561	total: 48.5s	remaining: 3m
212:	learn: 0.4058360	total: 48.8s	remaining: 3m
213:	learn: 0.4058271	total: 49s	remaining: 2m 59s
214:	learn: 0.4058121	total: 49.2s	remaining: 2m 59s
215:	learn: 0.4057984	total: 49.4s	remaining: 2m 59s
216:	learn: 0.4057882	total: 49.6s	remaining: 2m 58s
217:	learn: 0.4057719	total: 49.8s	remaining: 2m 58s
218:	learn: 0.4057617	total: 50s	remaining: 2m 58s
219:	learn: 0.4057519	total: 50.2s	remaining: 2m 58s
220:	learn: 0.4057273	total: 50.4s	remaining: 2m 57s
221:	learn: 0.4056960	total: 50.6s	remaining: 2m 57s
222:	learn: 0.4056795	total: 50.9s	remaining: 2m 57s
223:	learn: 0.4056680	total: 51.1s	remaining: 2m 56s
224:	learn: 0.4056563	total: 51.3s	remaining: 2m 56s
225:	learn: 0.4056457	total: 51.6s	remaining: 2m 56s
226:	learn: 0.4056328	total: 51.8s	remaining: 2m 56s
227:	learn: 0.4056212	total: 52s	remaining: 2m 55s
228:	learn: 0.4056033	total: 52.2s	remaining: 2m 55s
229:	learn: 0.4055903	total: 52.4s	remaining: 2m 55s
230:	learn: 0.4055802	total: 52.6s	remaining: 2m 55s
231:	learn: 0.4055721	total: 52.9s	remaining: 2m 54s
232:	learn: 0.4055312	total: 53.1s	remaining: 2m 54s
233:	learn: 0.4055076	total: 53.3s	remaining: 2m 54s
234:	learn: 0.4054938	total: 53.5s	remaining: 2m 54s
235:	learn: 0.4054813	total: 53.7s	remaining: 2m 53s
236:	learn: 0.4054677	total: 54s	remaining: 2m 53s
237:	learn: 0.4054590	total: 54.2s	remaining: 2m 53s
238:	learn: 0.4054464	total: 54.4s	remaining: 2m 53s
239:	learn: 0.4054315	total: 54.6s	remaining: 2m 53s
240:	learn: 0.4054149	total: 54.9s	remaining: 2m 52s
241:	learn: 0.4053760	total: 55.1s	remaining: 2m 52s
242:	learn: 0.4053650	total: 55.3s	remaining: 2m 52s
243:	learn: 0.4053556	total: 55.6s	remaining: 2m 52s
244:	learn: 0.4053454	total: 55.8s	remaining: 2m 51s
245:	learn: 0.4053390	total: 56s	remaining: 2m 51s
246:	learn: 0.4053257	total: 56.2s	remaining: 2m 51s
247:	learn: 0.4053130	total: 56.4s	remaining: 2m 51s
248:	learn: 0.4053038	total: 56.6s	remaining: 2m 50s
249:	learn: 0.4052820	total: 56.8s	remaining: 2m 50s

250:	learn: 0.4052560	total: 57s	remaining: 2m 50s
251:	learn: 0.4052438	total: 57.3s	remaining: 2m 49s
252:	learn: 0.4052355	total: 57.5s	remaining: 2m 49s
253:	learn: 0.4052228	total: 57.7s	remaining: 2m 49s
254:	learn: 0.4051931	total: 57.9s	remaining: 2m 49s
255:	learn: 0.4051823	total: 58.1s	remaining: 2m 48s
256:	learn: 0.4051648	total: 58.3s	remaining: 2m 48s
257:	learn: 0.4051546	total: 58.5s	remaining: 2m 48s
258:	learn: 0.4051246	total: 58.8s	remaining: 2m 48s
259:	learn: 0.4051168	total: 59s	remaining: 2m 47s
260:	learn: 0.4051040	total: 59.2s	remaining: 2m 47s
261:	learn: 0.4050899	total: 59.4s	remaining: 2m 47s
262:	learn: 0.4050770	total: 59.6s	remaining: 2m 46s
263:	learn: 0.4050651	total: 59.8s	remaining: 2m 46s
264:	learn: 0.4050544	total: 60s	remaining: 2m 46s
265:	learn: 0.4050407	total: 1m	remaining: 2m 46s
266:	learn: 0.4050311	total: 1m	remaining: 2m 45s
267:	learn: 0.4050281	total: 1m	remaining: 2m 45s
268:	learn: 0.4050195	total: 1m	remaining: 2m 45s
269:	learn: 0.4050085	total: 1m 1s	remaining: 2m 45s
270:	learn: 0.4050033	total: 1m 1s	remaining: 2m 44s
271:	learn: 0.4049935	total: 1m 1s	remaining: 2m 44s
272:	learn: 0.4049822	total: 1m 1s	remaining: 2m 44s
273:	learn: 0.4049738	total: 1m 1s	remaining: 2m 44s
274:	learn: 0.4049665	total: 1m 2s	remaining: 2m 43s
275:	learn: 0.4049579	total: 1m 2s	remaining: 2m 43s
276:	learn: 0.4049482	total: 1m 2s	remaining: 2m 43s
277:	learn: 0.4049412	total: 1m 2s	remaining: 2m 43s
278:	learn: 0.4049282	total: 1m 2s	remaining: 2m 42s
279:	learn: 0.4049244	total: 1m 3s	remaining: 2m 42s
280:	learn: 0.4049184	total: 1m 3s	remaining: 2m 42s
281:	learn: 0.4049084	total: 1m 3s	remaining: 2m 41s
282:	learn: 0.4048928	total: 1m 3s	remaining: 2m 41s
283:	learn: 0.4048869	total: 1m 4s	remaining: 2m 41s
284:	learn: 0.4048683	total: 1m 4s	remaining: 2m 41s
285:	learn: 0.4048518	total: 1m 4s	remaining: 2m 41s
286:	learn: 0.4048329	total: 1m 4s	remaining: 2m 40s
287:	learn: 0.4048286	total: 1m 4s	remaining: 2m 40s
288:	learn: 0.4048170	total: 1m 5s	remaining: 2m 40s
289:	learn: 0.4048022	total: 1m 5s	remaining: 2m 40s
290:	learn: 0.4047942	total: 1m 5s	remaining: 2m 39s
291:	learn: 0.4047865	total: 1m 5s	remaining: 2m 39s
292:	learn: 0.4047767	total: 1m 5s	remaining: 2m 39s
293:	learn: 0.4047712	total: 1m 6s	remaining: 2m 38s
294:	learn: 0.4047599	total: 1m 6s	remaining: 2m 38s
295:	learn: 0.4047497	total: 1m 6s	remaining: 2m 38s
296:	learn: 0.4047346	total: 1m 6s	remaining: 2m 38s
297:	learn: 0.4047153	total: 1m 7s	remaining: 2m 37s

298:	learn: 0.4047060	total: 1m 7s	remaining: 2m 37s
299:	learn: 0.4047011	total: 1m 7s	remaining: 2m 37s
300:	learn: 0.4046944	total: 1m 7s	remaining: 2m 37s
301:	learn: 0.4046841	total: 1m 7s	remaining: 2m 36s
302:	learn: 0.4046742	total: 1m 8s	remaining: 2m 36s
303:	learn: 0.4046664	total: 1m 8s	remaining: 2m 36s
304:	learn: 0.4046568	total: 1m 8s	remaining: 2m 36s
305:	learn: 0.4046493	total: 1m 8s	remaining: 2m 35s
306:	learn: 0.4046417	total: 1m 8s	remaining: 2m 35s
307:	learn: 0.4046333	total: 1m 9s	remaining: 2m 35s
308:	learn: 0.4046233	total: 1m 9s	remaining: 2m 35s
309:	learn: 0.4046101	total: 1m 9s	remaining: 2m 34s
310:	learn: 0.4045998	total: 1m 9s	remaining: 2m 34s
311:	learn: 0.4045916	total: 1m 10s	remaining: 2m 34s
312:	learn: 0.4045847	total: 1m 10s	remaining: 2m 34s
313:	learn: 0.4045777	total: 1m 10s	remaining: 2m 33s
314:	learn: 0.4045701	total: 1m 10s	remaining: 2m 33s
315:	learn: 0.4045600	total: 1m 10s	remaining: 2m 33s
316:	learn: 0.4045482	total: 1m 11s	remaining: 2m 33s
317:	learn: 0.4045382	total: 1m 11s	remaining: 2m 32s
318:	learn: 0.4045285	total: 1m 11s	remaining: 2m 32s
319:	learn: 0.4045192	total: 1m 11s	remaining: 2m 32s
320:	learn: 0.4045016	total: 1m 11s	remaining: 2m 32s
321:	learn: 0.4044942	total: 1m 12s	remaining: 2m 31s
322:	learn: 0.4044870	total: 1m 12s	remaining: 2m 31s
323:	learn: 0.4044795	total: 1m 12s	remaining: 2m 31s
324:	learn: 0.4044713	total: 1m 12s	remaining: 2m 31s
325:	learn: 0.4044562	total: 1m 12s	remaining: 2m 30s
326:	learn: 0.4044477	total: 1m 13s	remaining: 2m 30s
327:	learn: 0.4044398	total: 1m 13s	remaining: 2m 30s
328:	learn: 0.4044316	total: 1m 13s	remaining: 2m 30s
329:	learn: 0.4044225	total: 1m 13s	remaining: 2m 29s
330:	learn: 0.4044170	total: 1m 14s	remaining: 2m 29s
331:	learn: 0.4044082	total: 1m 14s	remaining: 2m 29s
332:	learn: 0.4043991	total: 1m 14s	remaining: 2m 29s
333:	learn: 0.4043951	total: 1m 14s	remaining: 2m 28s
334:	learn: 0.4043878	total: 1m 14s	remaining: 2m 28s
335:	learn: 0.4043777	total: 1m 15s	remaining: 2m 28s
336:	learn: 0.4043654	total: 1m 15s	remaining: 2m 28s
337:	learn: 0.4043588	total: 1m 15s	remaining: 2m 28s
338:	learn: 0.4043501	total: 1m 15s	remaining: 2m 27s
339:	learn: 0.4043456	total: 1m 16s	remaining: 2m 27s
340:	learn: 0.4043331	total: 1m 16s	remaining: 2m 27s
341:	learn: 0.4043244	total: 1m 16s	remaining: 2m 27s
342:	learn: 0.4043182	total: 1m 16s	remaining: 2m 26s
343:	learn: 0.4043133	total: 1m 16s	remaining: 2m 26s
344:	learn: 0.4043056	total: 1m 17s	remaining: 2m 26s
345:	learn: 0.4042980	total: 1m 17s	remaining: 2m 26s



346:	learn: 0.4042896	total: 1m 17s	remaining: 2m 25s
347:	learn: 0.4042847	total: 1m 17s	remaining: 2m 25s
348:	learn: 0.4042792	total: 1m 17s	remaining: 2m 25s
349:	learn: 0.4042700	total: 1m 18s	remaining: 2m 25s
350:	learn: 0.4042587	total: 1m 18s	remaining: 2m 24s
351:	learn: 0.4042526	total: 1m 18s	remaining: 2m 24s
352:	learn: 0.4042446	total: 1m 18s	remaining: 2m 24s
353:	learn: 0.4042339	total: 1m 18s	remaining: 2m 24s
354:	learn: 0.4042270	total: 1m 19s	remaining: 2m 23s
355:	learn: 0.4042166	total: 1m 19s	remaining: 2m 23s
356:	learn: 0.4042099	total: 1m 19s	remaining: 2m 23s
357:	learn: 0.4042053	total: 1m 19s	remaining: 2m 23s
358:	learn: 0.4041939	total: 1m 20s	remaining: 2m 22s
359:	learn: 0.4041841	total: 1m 20s	remaining: 2m 22s
360:	learn: 0.4041781	total: 1m 20s	remaining: 2m 22s
361:	learn: 0.4041533	total: 1m 20s	remaining: 2m 22s
362:	learn: 0.4041413	total: 1m 20s	remaining: 2m 22s
363:	learn: 0.4041387	total: 1m 21s	remaining: 2m 21s
364:	learn: 0.4041178	total: 1m 21s	remaining: 2m 21s
365:	learn: 0.4041116	total: 1m 21s	remaining: 2m 21s
366:	learn: 0.4041036	total: 1m 21s	remaining: 2m 21s
367:	learn: 0.4040965	total: 1m 22s	remaining: 2m 20s
368:	learn: 0.4040894	total: 1m 22s	remaining: 2m 20s
369:	learn: 0.4040849	total: 1m 22s	remaining: 2m 20s
370:	learn: 0.4040799	total: 1m 22s	remaining: 2m 20s
371:	learn: 0.4040747	total: 1m 22s	remaining: 2m 19s
372:	learn: 0.4040620	total: 1m 23s	remaining: 2m 19s
373:	learn: 0.4040541	total: 1m 23s	remaining: 2m 19s
374:	learn: 0.4040487	total: 1m 23s	remaining: 2m 19s
375:	learn: 0.4040389	total: 1m 23s	remaining: 2m 19s
376:	learn: 0.4040325	total: 1m 24s	remaining: 2m 18s
377:	learn: 0.4040226	total: 1m 24s	remaining: 2m 18s
378:	learn: 0.4040149	total: 1m 24s	remaining: 2m 18s
379:	learn: 0.4040036	total: 1m 24s	remaining: 2m 18s
380:	learn: 0.4039975	total: 1m 24s	remaining: 2m 17s
381:	learn: 0.4039907	total: 1m 25s	remaining: 2m 17s
382:	learn: 0.4039751	total: 1m 25s	remaining: 2m 17s
383:	learn: 0.4039666	total: 1m 25s	remaining: 2m 17s
384:	learn: 0.4039637	total: 1m 25s	remaining: 2m 17s
385:	learn: 0.4039527	total: 1m 25s	remaining: 2m 16s
386:	learn: 0.4039453	total: 1m 26s	remaining: 2m 16s
387:	learn: 0.4039395	total: 1m 26s	remaining: 2m 16s
388:	learn: 0.4039329	total: 1m 26s	remaining: 2m 16s
389:	learn: 0.4039139	total: 1m 26s	remaining: 2m 15s
390:	learn: 0.4039024	total: 1m 27s	remaining: 2m 15s
391:	learn: 0.4038922	total: 1m 27s	remaining: 2m 15s
392:	learn: 0.4038837	total: 1m 27s	remaining: 2m 15s
393:	learn: 0.4038770	total: 1m 27s	remaining: 2m 14s

394:	learn: 0.4038738	total: 1m 27s	remaining: 2m 14s
395:	learn: 0.4038686	total: 1m 28s	remaining: 2m 14s
396:	learn: 0.4038600	total: 1m 28s	remaining: 2m 14s
397:	learn: 0.4038522	total: 1m 28s	remaining: 2m 14s
398:	learn: 0.4038489	total: 1m 28s	remaining: 2m 13s
399:	learn: 0.4038430	total: 1m 29s	remaining: 2m 13s
400:	learn: 0.4038402	total: 1m 29s	remaining: 2m 13s
401:	learn: 0.4038236	total: 1m 29s	remaining: 2m 13s
402:	learn: 0.4038122	total: 1m 29s	remaining: 2m 12s
403:	learn: 0.4038069	total: 1m 29s	remaining: 2m 12s
404:	learn: 0.4037984	total: 1m 30s	remaining: 2m 12s
405:	learn: 0.4037915	total: 1m 30s	remaining: 2m 12s
406:	learn: 0.4037868	total: 1m 30s	remaining: 2m 11s
407:	learn: 0.4037812	total: 1m 30s	remaining: 2m 11s
408:	learn: 0.4037663	total: 1m 30s	remaining: 2m 11s
409:	learn: 0.4037614	total: 1m 31s	remaining: 2m 11s
410:	learn: 0.4037561	total: 1m 31s	remaining: 2m 10s
411:	learn: 0.4037523	total: 1m 31s	remaining: 2m 10s
412:	learn: 0.4037438	total: 1m 31s	remaining: 2m 10s
413:	learn: 0.4037360	total: 1m 32s	remaining: 2m 10s
414:	learn: 0.4037307	total: 1m 32s	remaining: 2m 10s
415:	learn: 0.4037230	total: 1m 32s	remaining: 2m 9s
416:	learn: 0.4037179	total: 1m 32s	remaining: 2m 9s
417:	learn: 0.4037112	total: 1m 32s	remaining: 2m 9s
418:	learn: 0.4037058	total: 1m 33s	remaining: 2m 9s
419:	learn: 0.4036983	total: 1m 33s	remaining: 2m 9s
420:	learn: 0.4036899	total: 1m 33s	remaining: 2m 9s
421:	learn: 0.4036855	total: 1m 34s	remaining: 2m 8s
422:	learn: 0.4036775	total: 1m 34s	remaining: 2m 8s
423:	learn: 0.4036693	total: 1m 34s	remaining: 2m 8s
424:	learn: 0.4036594	total: 1m 34s	remaining: 2m 8s
425:	learn: 0.4036503	total: 1m 35s	remaining: 2m 8s
426:	learn: 0.4036436	total: 1m 35s	remaining: 2m 7s
427:	learn: 0.4036339	total: 1m 35s	remaining: 2m 7s
428:	learn: 0.4036271	total: 1m 35s	remaining: 2m 7s
429:	learn: 0.4036212	total: 1m 36s	remaining: 2m 7s
430:	learn: 0.4036081	total: 1m 36s	remaining: 2m 7s
431:	learn: 0.4035991	total: 1m 36s	remaining: 2m 6s
432:	learn: 0.4035927	total: 1m 36s	remaining: 2m 6s
433:	learn: 0.4035819	total: 1m 36s	remaining: 2m 6s
434:	learn: 0.4035750	total: 1m 37s	remaining: 2m 6s
435:	learn: 0.4035671	total: 1m 37s	remaining: 2m 5s
436:	learn: 0.4035594	total: 1m 37s	remaining: 2m 5s
437:	learn: 0.4035505	total: 1m 37s	remaining: 2m 5s
438:	learn: 0.4035362	total: 1m 37s	remaining: 2m 5s
439:	learn: 0.4035278	total: 1m 38s	remaining: 2m 4s
440:	learn: 0.4035191	total: 1m 38s	remaining: 2m 4s
441:	learn: 0.4035073	total: 1m 38s	remaining: 2m 4s

442:	learn: 0.4034928	total: 1m 38s	remaining: 2m 4s
443:	learn: 0.4034908	total: 1m 39s	remaining: 2m 4s
444:	learn: 0.4034849	total: 1m 39s	remaining: 2m 3s
445:	learn: 0.4034825	total: 1m 39s	remaining: 2m 3s
446:	learn: 0.4034709	total: 1m 39s	remaining: 2m 3s
447:	learn: 0.4034659	total: 1m 39s	remaining: 2m 3s
448:	learn: 0.4034608	total: 1m 40s	remaining: 2m 2s
449:	learn: 0.4034547	total: 1m 40s	remaining: 2m 2s
450:	learn: 0.4034493	total: 1m 40s	remaining: 2m 2s
451:	learn: 0.4034411	total: 1m 40s	remaining: 2m 2s
452:	learn: 0.4034367	total: 1m 40s	remaining: 2m 1s
453:	learn: 0.4034324	total: 1m 41s	remaining: 2m 1s
454:	learn: 0.4034201	total: 1m 41s	remaining: 2m 1s
455:	learn: 0.4034119	total: 1m 41s	remaining: 2m 1s
456:	learn: 0.4034057	total: 1m 41s	remaining: 2m
457:	learn: 0.4034007	total: 1m 42s	remaining: 2m
458:	learn: 0.4033971	total: 1m 42s	remaining: 2m
459:	learn: 0.4033943	total: 1m 42s	remaining: 2m
460:	learn: 0.4033897	total: 1m 42s	remaining: 2m
461:	learn: 0.4033815	total: 1m 42s	remaining: 1m 59s
462:	learn: 0.4033740	total: 1m 43s	remaining: 1m 59s
463:	learn: 0.4033700	total: 1m 43s	remaining: 1m 59s
464:	learn: 0.4033637	total: 1m 43s	remaining: 1m 59s
465:	learn: 0.4033566	total: 1m 43s	remaining: 1m 58s
466:	learn: 0.4033492	total: 1m 43s	remaining: 1m 58s
467:	learn: 0.4033378	total: 1m 44s	remaining: 1m 58s
468:	learn: 0.4033315	total: 1m 44s	remaining: 1m 58s
469:	learn: 0.4033267	total: 1m 44s	remaining: 1m 57s
470:	learn: 0.4033203	total: 1m 44s	remaining: 1m 57s
471:	learn: 0.4033163	total: 1m 45s	remaining: 1m 57s
472:	learn: 0.4033119	total: 1m 45s	remaining: 1m 57s
473:	learn: 0.4033059	total: 1m 45s	remaining: 1m 56s
474:	learn: 0.4032999	total: 1m 45s	remaining: 1m 56s
475:	learn: 0.4032937	total: 1m 45s	remaining: 1m 56s
476:	learn: 0.4032887	total: 1m 46s	remaining: 1m 56s
477:	learn: 0.4032820	total: 1m 46s	remaining: 1m 56s
478:	learn: 0.4032790	total: 1m 46s	remaining: 1m 55s
479:	learn: 0.4032735	total: 1m 46s	remaining: 1m 55s
480:	learn: 0.4032705	total: 1m 46s	remaining: 1m 55s
481:	learn: 0.4032659	total: 1m 47s	remaining: 1m 55s
482:	learn: 0.4032555	total: 1m 47s	remaining: 1m 54s
483:	learn: 0.4032477	total: 1m 47s	remaining: 1m 54s
484:	learn: 0.4032417	total: 1m 47s	remaining: 1m 54s
485:	learn: 0.4032333	total: 1m 48s	remaining: 1m 54s
486:	learn: 0.4032280	total: 1m 48s	remaining: 1m 54s
487:	learn: 0.4032244	total: 1m 48s	remaining: 1m 53s
488:	learn: 0.4032136	total: 1m 48s	remaining: 1m 53s
489:	learn: 0.4032055	total: 1m 48s	remaining: 1m 53s

490:	learn: 0.4031998	total: 1m 49s	remaining: 1m 53s
491:	learn: 0.4031979	total: 1m 49s	remaining: 1m 52s
492:	learn: 0.4031886	total: 1m 49s	remaining: 1m 52s
493:	learn: 0.4031808	total: 1m 49s	remaining: 1m 52s
494:	learn: 0.4031771	total: 1m 50s	remaining: 1m 52s
495:	learn: 0.4031680	total: 1m 50s	remaining: 1m 52s
496:	learn: 0.4031602	total: 1m 50s	remaining: 1m 51s
497:	learn: 0.4031575	total: 1m 50s	remaining: 1m 51s
498:	learn: 0.4031538	total: 1m 50s	remaining: 1m 51s
499:	learn: 0.4031491	total: 1m 51s	remaining: 1m 51s
500:	learn: 0.4031455	total: 1m 51s	remaining: 1m 50s
501:	learn: 0.4031383	total: 1m 51s	remaining: 1m 50s
502:	learn: 0.4031360	total: 1m 51s	remaining: 1m 50s
503:	learn: 0.4031290	total: 1m 51s	remaining: 1m 50s
504:	learn: 0.4031216	total: 1m 52s	remaining: 1m 49s
505:	learn: 0.4031173	total: 1m 52s	remaining: 1m 49s
506:	learn: 0.4031090	total: 1m 52s	remaining: 1m 49s
507:	learn: 0.4031009	total: 1m 52s	remaining: 1m 49s
508:	learn: 0.4030906	total: 1m 53s	remaining: 1m 49s
509:	learn: 0.4030846	total: 1m 53s	remaining: 1m 48s
510:	learn: 0.4030787	total: 1m 53s	remaining: 1m 48s
511:	learn: 0.4030740	total: 1m 53s	remaining: 1m 48s
512:	learn: 0.4030682	total: 1m 53s	remaining: 1m 48s
513:	learn: 0.4030635	total: 1m 54s	remaining: 1m 47s
514:	learn: 0.4030552	total: 1m 54s	remaining: 1m 47s
515:	learn: 0.4030497	total: 1m 54s	remaining: 1m 47s
516:	learn: 0.4030449	total: 1m 54s	remaining: 1m 47s
517:	learn: 0.4030383	total: 1m 54s	remaining: 1m 46s
518:	learn: 0.4030321	total: 1m 55s	remaining: 1m 46s
519:	learn: 0.4030269	total: 1m 55s	remaining: 1m 46s
520:	learn: 0.4030122	total: 1m 55s	remaining: 1m 46s
521:	learn: 0.4030080	total: 1m 55s	remaining: 1m 46s
522:	learn: 0.4030037	total: 1m 56s	remaining: 1m 45s
523:	learn: 0.4029985	total: 1m 56s	remaining: 1m 45s
524:	learn: 0.4029909	total: 1m 56s	remaining: 1m 45s
525:	learn: 0.4029862	total: 1m 56s	remaining: 1m 45s
526:	learn: 0.4029821	total: 1m 56s	remaining: 1m 44s
527:	learn: 0.4029769	total: 1m 57s	remaining: 1m 44s
528:	learn: 0.4029694	total: 1m 57s	remaining: 1m 44s
529:	learn: 0.4029649	total: 1m 57s	remaining: 1m 44s
530:	learn: 0.4029612	total: 1m 57s	remaining: 1m 44s
531:	learn: 0.4029518	total: 1m 57s	remaining: 1m 43s
532:	learn: 0.4029481	total: 1m 58s	remaining: 1m 43s
533:	learn: 0.4029448	total: 1m 58s	remaining: 1m 43s
534:	learn: 0.4029389	total: 1m 58s	remaining: 1m 43s
535:	learn: 0.4029338	total: 1m 58s	remaining: 1m 42s
536:	learn: 0.4029284	total: 1m 59s	remaining: 1m 42s
537:	learn: 0.4029233	total: 1m 59s	remaining: 1m 42s

538:	learn: 0.4029179	total: 1m 59s	remaining: 1m 42s
539:	learn: 0.4029124	total: 1m 59s	remaining: 1m 41s
540:	learn: 0.4029079	total: 1m 59s	remaining: 1m 41s
541:	learn: 0.4029042	total: 2m	remaining: 1m 41s
542:	learn: 0.4029000	total: 2m	remaining: 1m 41s
543:	learn: 0.4028950	total: 2m	remaining: 1m 41s
544:	learn: 0.4028910	total: 2m	remaining: 1m 40s
545:	learn: 0.4028850	total: 2m 1s	remaining: 1m 40s
546:	learn: 0.4028800	total: 2m 1s	remaining: 1m 40s
547:	learn: 0.4028706	total: 2m 1s	remaining: 1m 40s
548:	learn: 0.4028653	total: 2m 1s	remaining: 1m 40s
549:	learn: 0.4028552	total: 2m 2s	remaining: 1m 39s
550:	learn: 0.4028488	total: 2m 2s	remaining: 1m 39s
551:	learn: 0.4028436	total: 2m 2s	remaining: 1m 39s
552:	learn: 0.4028375	total: 2m 2s	remaining: 1m 39s
553:	learn: 0.4028321	total: 2m 2s	remaining: 1m 38s
554:	learn: 0.4028267	total: 2m 3s	remaining: 1m 38s
555:	learn: 0.4028197	total: 2m 3s	remaining: 1m 38s
556:	learn: 0.4028123	total: 2m 3s	remaining: 1m 38s
557:	learn: 0.4028078	total: 2m 3s	remaining: 1m 38s
558:	learn: 0.4028030	total: 2m 3s	remaining: 1m 37s
559:	learn: 0.4027996	total: 2m 4s	remaining: 1m 37s
560:	learn: 0.4027905	total: 2m 4s	remaining: 1m 37s
561:	learn: 0.4027852	total: 2m 4s	remaining: 1m 37s
562:	learn: 0.4027780	total: 2m 4s	remaining: 1m 36s
563:	learn: 0.4027714	total: 2m 5s	remaining: 1m 36s
564:	learn: 0.4027650	total: 2m 5s	remaining: 1m 36s
565:	learn: 0.4027556	total: 2m 5s	remaining: 1m 36s
566:	learn: 0.4027503	total: 2m 5s	remaining: 1m 36s
567:	learn: 0.4027432	total: 2m 6s	remaining: 1m 35s
568:	learn: 0.4027354	total: 2m 6s	remaining: 1m 35s
569:	learn: 0.4027322	total: 2m 6s	remaining: 1m 35s
570:	learn: 0.4027298	total: 2m 6s	remaining: 1m 35s
571:	learn: 0.4027265	total: 2m 6s	remaining: 1m 34s
572:	learn: 0.4027239	total: 2m 7s	remaining: 1m 34s
573:	learn: 0.4027158	total: 2m 7s	remaining: 1m 34s
574:	learn: 0.4027115	total: 2m 7s	remaining: 1m 34s
575:	learn: 0.4027060	total: 2m 7s	remaining: 1m 34s
576:	learn: 0.4027023	total: 2m 7s	remaining: 1m 33s
577:	learn: 0.4026939	total: 2m 8s	remaining: 1m 33s
578:	learn: 0.4026832	total: 2m 8s	remaining: 1m 33s
579:	learn: 0.4026789	total: 2m 8s	remaining: 1m 33s
580:	learn: 0.4026739	total: 2m 8s	remaining: 1m 32s
581:	learn: 0.4026658	total: 2m 9s	remaining: 1m 32s
582:	learn: 0.4026614	total: 2m 9s	remaining: 1m 32s
583:	learn: 0.4026569	total: 2m 9s	remaining: 1m 32s
584:	learn: 0.4026536	total: 2m 9s	remaining: 1m 32s
585:	learn: 0.4026498	total: 2m 9s	remaining: 1m 31s

586:	learn: 0.4026459	total: 2m 10s	remaining: 1m 31s
587:	learn: 0.4026422	total: 2m 10s	remaining: 1m 31s
588:	learn: 0.4026398	total: 2m 10s	remaining: 1m 31s
589:	learn: 0.4026385	total: 2m 10s	remaining: 1m 30s
590:	learn: 0.4026339	total: 2m 10s	remaining: 1m 30s
591:	learn: 0.4026287	total: 2m 11s	remaining: 1m 30s
592:	learn: 0.4026234	total: 2m 11s	remaining: 1m 30s
593:	learn: 0.4026174	total: 2m 11s	remaining: 1m 29s
594:	learn: 0.4026088	total: 2m 11s	remaining: 1m 29s
595:	learn: 0.4026010	total: 2m 12s	remaining: 1m 29s
596:	learn: 0.4025992	total: 2m 12s	remaining: 1m 29s
597:	learn: 0.4025964	total: 2m 12s	remaining: 1m 29s
598:	learn: 0.4025917	total: 2m 12s	remaining: 1m 28s
599:	learn: 0.4025833	total: 2m 12s	remaining: 1m 28s
600:	learn: 0.4025678	total: 2m 13s	remaining: 1m 28s
601:	learn: 0.4025607	total: 2m 13s	remaining: 1m 28s
602:	learn: 0.4025555	total: 2m 13s	remaining: 1m 27s
603:	learn: 0.4025469	total: 2m 13s	remaining: 1m 27s
604:	learn: 0.4025411	total: 2m 13s	remaining: 1m 27s
605:	learn: 0.4025354	total: 2m 14s	remaining: 1m 27s
606:	learn: 0.4025300	total: 2m 14s	remaining: 1m 27s
607:	learn: 0.4025222	total: 2m 14s	remaining: 1m 26s
608:	learn: 0.4025186	total: 2m 14s	remaining: 1m 26s
609:	learn: 0.4025103	total: 2m 15s	remaining: 1m 26s
610:	learn: 0.4025074	total: 2m 15s	remaining: 1m 26s
611:	learn: 0.4024992	total: 2m 15s	remaining: 1m 25s
612:	learn: 0.4024919	total: 2m 15s	remaining: 1m 25s
613:	learn: 0.4024857	total: 2m 15s	remaining: 1m 25s
614:	learn: 0.4024817	total: 2m 16s	remaining: 1m 25s
615:	learn: 0.4024690	total: 2m 16s	remaining: 1m 25s
616:	learn: 0.4024657	total: 2m 16s	remaining: 1m 24s
617:	learn: 0.4024609	total: 2m 16s	remaining: 1m 24s
618:	learn: 0.4024567	total: 2m 17s	remaining: 1m 24s
619:	learn: 0.4024507	total: 2m 17s	remaining: 1m 24s
620:	learn: 0.4024448	total: 2m 17s	remaining: 1m 23s
621:	learn: 0.4024375	total: 2m 17s	remaining: 1m 23s
622:	learn: 0.4024319	total: 2m 17s	remaining: 1m 23s
623:	learn: 0.4024246	total: 2m 18s	remaining: 1m 23s
624:	learn: 0.4024197	total: 2m 18s	remaining: 1m 23s
625:	learn: 0.4024156	total: 2m 18s	remaining: 1m 22s
626:	learn: 0.4024014	total: 2m 18s	remaining: 1m 22s
627:	learn: 0.4023957	total: 2m 19s	remaining: 1m 22s
628:	learn: 0.4023900	total: 2m 19s	remaining: 1m 22s
629:	learn: 0.4023866	total: 2m 19s	remaining: 1m 21s
630:	learn: 0.4023829	total: 2m 19s	remaining: 1m 21s
631:	learn: 0.4023789	total: 2m 19s	remaining: 1m 21s
632:	learn: 0.4023748	total: 2m 20s	remaining: 1m 21s
633:	learn: 0.4023691	total: 2m 20s	remaining: 1m 21s

634:	learn: 0.4023654	total: 2m 20s	remaining: 1m 20s
635:	learn: 0.4023537	total: 2m 20s	remaining: 1m 20s
636:	learn: 0.4023527	total: 2m 21s	remaining: 1m 20s
637:	learn: 0.4023477	total: 2m 21s	remaining: 1m 20s
638:	learn: 0.4023416	total: 2m 21s	remaining: 1m 19s
639:	learn: 0.4023368	total: 2m 21s	remaining: 1m 19s
640:	learn: 0.4023307	total: 2m 21s	remaining: 1m 19s
641:	learn: 0.4023279	total: 2m 22s	remaining: 1m 19s
642:	learn: 0.4023220	total: 2m 22s	remaining: 1m 19s
643:	learn: 0.4023163	total: 2m 22s	remaining: 1m 18s
644:	learn: 0.4023124	total: 2m 22s	remaining: 1m 18s
645:	learn: 0.4023072	total: 2m 22s	remaining: 1m 18s
646:	learn: 0.4023025	total: 2m 23s	remaining: 1m 18s
647:	learn: 0.4022968	total: 2m 23s	remaining: 1m 17s
648:	learn: 0.4022916	total: 2m 23s	remaining: 1m 17s
649:	learn: 0.4022886	total: 2m 23s	remaining: 1m 17s
650:	learn: 0.4022810	total: 2m 24s	remaining: 1m 17s
651:	learn: 0.4022769	total: 2m 24s	remaining: 1m 17s
652:	learn: 0.4022710	total: 2m 24s	remaining: 1m 16s
653:	learn: 0.4022655	total: 2m 24s	remaining: 1m 16s
654:	learn: 0.4022588	total: 2m 24s	remaining: 1m 16s
655:	learn: 0.4022559	total: 2m 25s	remaining: 1m 16s
656:	learn: 0.4022499	total: 2m 25s	remaining: 1m 15s
657:	learn: 0.4022455	total: 2m 25s	remaining: 1m 15s
658:	learn: 0.4022432	total: 2m 25s	remaining: 1m 15s
659:	learn: 0.4022402	total: 2m 25s	remaining: 1m 15s
660:	learn: 0.4022363	total: 2m 26s	remaining: 1m 14s
661:	learn: 0.4022313	total: 2m 26s	remaining: 1m 14s
662:	learn: 0.4022284	total: 2m 26s	remaining: 1m 14s
663:	learn: 0.4022242	total: 2m 26s	remaining: 1m 14s
664:	learn: 0.4022204	total: 2m 27s	remaining: 1m 14s
665:	learn: 0.4022142	total: 2m 27s	remaining: 1m 13s
666:	learn: 0.4022075	total: 2m 27s	remaining: 1m 13s
667:	learn: 0.4022011	total: 2m 27s	remaining: 1m 13s
668:	learn: 0.4021965	total: 2m 27s	remaining: 1m 13s
669:	learn: 0.4021950	total: 2m 28s	remaining: 1m 12s
670:	learn: 0.4021920	total: 2m 28s	remaining: 1m 12s
671:	learn: 0.4021856	total: 2m 28s	remaining: 1m 12s
672:	learn: 0.4021782	total: 2m 28s	remaining: 1m 12s
673:	learn: 0.4021734	total: 2m 28s	remaining: 1m 12s
674:	learn: 0.4021722	total: 2m 29s	remaining: 1m 11s
675:	learn: 0.4021692	total: 2m 29s	remaining: 1m 11s
676:	learn: 0.4021640	total: 2m 29s	remaining: 1m 11s
677:	learn: 0.4021611	total: 2m 29s	remaining: 1m 11s
678:	learn: 0.4021553	total: 2m 30s	remaining: 1m 10s
679:	learn: 0.4021496	total: 2m 30s	remaining: 1m 10s
680:	learn: 0.4021425	total: 2m 30s	remaining: 1m 10s
681:	learn: 0.4021355	total: 2m 30s	remaining: 1m 10s

682:	learn: 0.4021307	total: 2m 30s	remaining: 1m 10s
683:	learn: 0.4021251	total: 2m 31s	remaining: 1m 9s
684:	learn: 0.4021213	total: 2m 31s	remaining: 1m 9s
685:	learn: 0.4021174	total: 2m 31s	remaining: 1m 9s
686:	learn: 0.4021132	total: 2m 31s	remaining: 1m 9s
687:	learn: 0.4021068	total: 2m 31s	remaining: 1m 8s
688:	learn: 0.4021040	total: 2m 32s	remaining: 1m 8s
689:	learn: 0.4020947	total: 2m 32s	remaining: 1m 8s
690:	learn: 0.4020900	total: 2m 32s	remaining: 1m 8s
691:	learn: 0.4020852	total: 2m 32s	remaining: 1m 8s
692:	learn: 0.4020796	total: 2m 33s	remaining: 1m 7s
693:	learn: 0.4020768	total: 2m 33s	remaining: 1m 7s
694:	learn: 0.4020718	total: 2m 33s	remaining: 1m 7s
695:	learn: 0.4020690	total: 2m 33s	remaining: 1m 7s
696:	learn: 0.4020645	total: 2m 33s	remaining: 1m 6s
697:	learn: 0.4020590	total: 2m 34s	remaining: 1m 6s
698:	learn: 0.4020456	total: 2m 34s	remaining: 1m 6s
699:	learn: 0.4020410	total: 2m 34s	remaining: 1m 6s
700:	learn: 0.4020380	total: 2m 34s	remaining: 1m 6s
701:	learn: 0.4020311	total: 2m 35s	remaining: 1m 5s
702:	learn: 0.4020263	total: 2m 35s	remaining: 1m 5s
703:	learn: 0.4020234	total: 2m 35s	remaining: 1m 5s
704:	learn: 0.4020158	total: 2m 35s	remaining: 1m 5s
705:	learn: 0.4020080	total: 2m 35s	remaining: 1m 4s
706:	learn: 0.4020018	total: 2m 36s	remaining: 1m 4s
707:	learn: 0.4019939	total: 2m 36s	remaining: 1m 4s
708:	learn: 0.4019894	total: 2m 36s	remaining: 1m 4s
709:	learn: 0.4019869	total: 2m 36s	remaining: 1m 4s
710:	learn: 0.4019819	total: 2m 37s	remaining: 1m 3s
711:	learn: 0.4019770	total: 2m 37s	remaining: 1m 3s
712:	learn: 0.4019721	total: 2m 37s	remaining: 1m 3s
713:	learn: 0.4019647	total: 2m 37s	remaining: 1m 3s
714:	learn: 0.4019566	total: 2m 37s	remaining: 1m 2s
715:	learn: 0.4019520	total: 2m 38s	remaining: 1m 2s
716:	learn: 0.4019477	total: 2m 38s	remaining: 1m 2s
717:	learn: 0.4019418	total: 2m 38s	remaining: 1m 2s
718:	learn: 0.4019384	total: 2m 38s	remaining: 1m 2s
719:	learn: 0.4019311	total: 2m 39s	remaining: 1m 1s
720:	learn: 0.4019257	total: 2m 39s	remaining: 1m 1s
721:	learn: 0.4019202	total: 2m 39s	remaining: 1m 1s
722:	learn: 0.4019164	total: 2m 39s	remaining: 1m 1s
723:	learn: 0.4019100	total: 2m 39s	remaining: 1m
724:	learn: 0.4019044	total: 2m 40s	remaining: 1m
725:	learn: 0.4019015	total: 2m 40s	remaining: 1m
726:	learn: 0.4018971	total: 2m 40s	remaining: 1m
727:	learn: 0.4018935	total: 2m 40s	remaining: 1m
728:	learn: 0.4018887	total: 2m 40s	remaining: 59.8s
729:	learn: 0.4018851	total: 2m 41s	remaining: 59.6s



730:	learn: 0.4018792	total: 2m 41s	remaining: 59.4s
731:	learn: 0.4018761	total: 2m 41s	remaining: 59.2s
732:	learn: 0.4018729	total: 2m 41s	remaining: 59s
733:	learn: 0.4018677	total: 2m 42s	remaining: 58.7s
734:	learn: 0.4018572	total: 2m 42s	remaining: 58.5s
735:	learn: 0.4018527	total: 2m 42s	remaining: 58.3s
736:	learn: 0.4018468	total: 2m 42s	remaining: 58.1s
737:	learn: 0.4018422	total: 2m 42s	remaining: 57.9s
738:	learn: 0.4018386	total: 2m 43s	remaining: 57.6s
739:	learn: 0.4018359	total: 2m 43s	remaining: 57.4s
740:	learn: 0.4018319	total: 2m 43s	remaining: 57.2s
741:	learn: 0.4018288	total: 2m 43s	remaining: 57s
742:	learn: 0.4018234	total: 2m 44s	remaining: 56.7s
743:	learn: 0.4018181	total: 2m 44s	remaining: 56.5s
744:	learn: 0.4018149	total: 2m 44s	remaining: 56.3s
745:	learn: 0.4018088	total: 2m 44s	remaining: 56.1s
746:	learn: 0.4018040	total: 2m 44s	remaining: 55.9s
747:	learn: 0.4017997	total: 2m 45s	remaining: 55.6s
748:	learn: 0.4017950	total: 2m 45s	remaining: 55.4s
749:	learn: 0.4017926	total: 2m 45s	remaining: 55.2s
750:	learn: 0.4017865	total: 2m 45s	remaining: 55s
751:	learn: 0.4017787	total: 2m 45s	remaining: 54.7s
752:	learn: 0.4017728	total: 2m 46s	remaining: 54.5s
753:	learn: 0.4017659	total: 2m 46s	remaining: 54.3s
754:	learn: 0.4017614	total: 2m 46s	remaining: 54.1s
755:	learn: 0.4017559	total: 2m 46s	remaining: 53.8s
756:	learn: 0.4017512	total: 2m 47s	remaining: 53.6s
757:	learn: 0.4017490	total: 2m 47s	remaining: 53.4s
758:	learn: 0.4017445	total: 2m 47s	remaining: 53.2s
759:	learn: 0.4017391	total: 2m 47s	remaining: 52.9s
760:	learn: 0.4017271	total: 2m 47s	remaining: 52.7s
761:	learn: 0.4017206	total: 2m 48s	remaining: 52.5s
762:	learn: 0.4017158	total: 2m 48s	remaining: 52.3s
763:	learn: 0.4017060	total: 2m 48s	remaining: 52.1s
764:	learn: 0.4017040	total: 2m 48s	remaining: 51.8s
765:	learn: 0.4016992	total: 2m 48s	remaining: 51.6s
766:	learn: 0.4016949	total: 2m 49s	remaining: 51.4s
767:	learn: 0.4016901	total: 2m 49s	remaining: 51.2s
768:	learn: 0.4016871	total: 2m 49s	remaining: 50.9s
769:	learn: 0.4016831	total: 2m 49s	remaining: 50.7s
770:	learn: 0.4016785	total: 2m 49s	remaining: 50.5s
771:	learn: 0.4016708	total: 2m 50s	remaining: 50.3s
772:	learn: 0.4016663	total: 2m 50s	remaining: 50.1s
773:	learn: 0.4016604	total: 2m 50s	remaining: 49.8s
774:	learn: 0.4016554	total: 2m 50s	remaining: 49.6s
775:	learn: 0.4016511	total: 2m 51s	remaining: 49.4s
776:	learn: 0.4016486	total: 2m 51s	remaining: 49.2s
777:	learn: 0.4016432	total: 2m 51s	remaining: 48.9s

778:	learn: 0.4016379	total: 2m 51s	remaining: 48.7s
779:	learn: 0.4016317	total: 2m 51s	remaining: 48.5s
780:	learn: 0.4016288	total: 2m 52s	remaining: 48.3s
781:	learn: 0.4016251	total: 2m 52s	remaining: 48s
782:	learn: 0.4016214	total: 2m 52s	remaining: 47.8s
783:	learn: 0.4016151	total: 2m 52s	remaining: 47.6s
784:	learn: 0.4016078	total: 2m 53s	remaining: 47.4s
785:	learn: 0.4016044	total: 2m 53s	remaining: 47.2s
786:	learn: 0.4016003	total: 2m 53s	remaining: 46.9s
787:	learn: 0.4015964	total: 2m 53s	remaining: 46.7s
788:	learn: 0.4015921	total: 2m 53s	remaining: 46.5s
789:	learn: 0.4015875	total: 2m 54s	remaining: 46.3s
790:	learn: 0.4015830	total: 2m 54s	remaining: 46.1s
791:	learn: 0.4015765	total: 2m 54s	remaining: 45.8s
792:	learn: 0.4015748	total: 2m 54s	remaining: 45.6s
793:	learn: 0.4015702	total: 2m 54s	remaining: 45.4s
794:	learn: 0.4015674	total: 2m 55s	remaining: 45.2s
795:	learn: 0.4015625	total: 2m 55s	remaining: 45s
796:	learn: 0.4015586	total: 2m 55s	remaining: 44.7s
797:	learn: 0.4015544	total: 2m 55s	remaining: 44.5s
798:	learn: 0.4015492	total: 2m 56s	remaining: 44.3s
799:	learn: 0.4015448	total: 2m 56s	remaining: 44.1s
800:	learn: 0.4015407	total: 2m 56s	remaining: 43.8s
801:	learn: 0.4015382	total: 2m 56s	remaining: 43.6s
802:	learn: 0.4015358	total: 2m 56s	remaining: 43.4s
803:	learn: 0.4015296	total: 2m 57s	remaining: 43.2s
804:	learn: 0.4015263	total: 2m 57s	remaining: 43s
805:	learn: 0.4015228	total: 2m 57s	remaining: 42.7s
806:	learn: 0.4015176	total: 2m 57s	remaining: 42.5s
807:	learn: 0.4015132	total: 2m 58s	remaining: 42.3s
808:	learn: 0.4015057	total: 2m 58s	remaining: 42.1s
809:	learn: 0.4015017	total: 2m 58s	remaining: 41.9s
810:	learn: 0.4014946	total: 2m 58s	remaining: 41.6s
811:	learn: 0.4014929	total: 2m 58s	remaining: 41.4s
812:	learn: 0.4014885	total: 2m 59s	remaining: 41.2s
813:	learn: 0.4014833	total: 2m 59s	remaining: 41s
814:	learn: 0.4014807	total: 2m 59s	remaining: 40.7s
815:	learn: 0.4014699	total: 2m 59s	remaining: 40.5s
816:	learn: 0.4014681	total: 2m 59s	remaining: 40.3s
817:	learn: 0.4014655	total: 3m	remaining: 40.1s
818:	learn: 0.4014600	total: 3m	remaining: 39.8s
819:	learn: 0.4014553	total: 3m	remaining: 39.6s
820:	learn: 0.4014530	total: 3m	remaining: 39.4s
821:	learn: 0.4014512	total: 3m	remaining: 39.2s
822:	learn: 0.4014482	total: 3m 1s	remaining: 39s
823:	learn: 0.4014429	total: 3m 1s	remaining: 38.7s
824:	learn: 0.4014374	total: 3m 1s	remaining: 38.5s
825:	learn: 0.4014346	total: 3m 1s	remaining: 38.3s

826:	learn: 0.4014299	total: 3m 2s	remaining: 38.1s
827:	learn: 0.4014284	total: 3m 2s	remaining: 37.9s
828:	learn: 0.4014254	total: 3m 2s	remaining: 37.6s
829:	learn: 0.4014201	total: 3m 2s	remaining: 37.4s
830:	learn: 0.4014158	total: 3m 2s	remaining: 37.2s
831:	learn: 0.4014121	total: 3m 3s	remaining: 37s
832:	learn: 0.4014099	total: 3m 3s	remaining: 36.8s
833:	learn: 0.4014082	total: 3m 3s	remaining: 36.5s
834:	learn: 0.4014062	total: 3m 3s	remaining: 36.3s
835:	learn: 0.4014044	total: 3m 3s	remaining: 36.1s
836:	learn: 0.4013981	total: 3m 4s	remaining: 35.9s
837:	learn: 0.4013933	total: 3m 4s	remaining: 35.7s
838:	learn: 0.4013896	total: 3m 4s	remaining: 35.4s
839:	learn: 0.4013847	total: 3m 4s	remaining: 35.2s
840:	learn: 0.4013804	total: 3m 5s	remaining: 35s
841:	learn: 0.4013758	total: 3m 5s	remaining: 34.8s
842:	learn: 0.4013723	total: 3m 5s	remaining: 34.5s
843:	learn: 0.4013658	total: 3m 5s	remaining: 34.3s
844:	learn: 0.4013637	total: 3m 5s	remaining: 34.1s
845:	learn: 0.4013611	total: 3m 6s	remaining: 33.9s
846:	learn: 0.4013592	total: 3m 6s	remaining: 33.7s
847:	learn: 0.4013551	total: 3m 6s	remaining: 33.4s
848:	learn: 0.4013507	total: 3m 6s	remaining: 33.2s
849:	learn: 0.4013440	total: 3m 6s	remaining: 33s
850:	learn: 0.4013402	total: 3m 7s	remaining: 32.8s
851:	learn: 0.4013355	total: 3m 7s	remaining: 32.6s
852:	learn: 0.4013293	total: 3m 7s	remaining: 32.3s
853:	learn: 0.4013255	total: 3m 7s	remaining: 32.1s
854:	learn: 0.4013212	total: 3m 8s	remaining: 31.9s
855:	learn: 0.4013167	total: 3m 8s	remaining: 31.7s
856:	learn: 0.4013128	total: 3m 8s	remaining: 31.4s
857:	learn: 0.4013090	total: 3m 8s	remaining: 31.2s
858:	learn: 0.4013047	total: 3m 8s	remaining: 31s
859:	learn: 0.4012984	total: 3m 9s	remaining: 30.8s
860:	learn: 0.4012943	total: 3m 9s	remaining: 30.6s
861:	learn: 0.4012906	total: 3m 9s	remaining: 30.4s
862:	learn: 0.4012894	total: 3m 9s	remaining: 30.1s
863:	learn: 0.4012861	total: 3m 10s	remaining: 29.9s
864:	learn: 0.4012822	total: 3m 10s	remaining: 29.7s
865:	learn: 0.4012794	total: 3m 10s	remaining: 29.5s
866:	learn: 0.4012738	total: 3m 10s	remaining: 29.2s
867:	learn: 0.4012684	total: 3m 10s	remaining: 29s
868:	learn: 0.4012652	total: 3m 11s	remaining: 28.8s
869:	learn: 0.4012616	total: 3m 11s	remaining: 28.6s
870:	learn: 0.4012576	total: 3m 11s	remaining: 28.4s
871:	learn: 0.4012544	total: 3m 11s	remaining: 28.1s
872:	learn: 0.4012502	total: 3m 11s	remaining: 27.9s
873:	learn: 0.4012453	total: 3m 12s	remaining: 27.7s

874:	learn: 0.4012389	total: 3m 12s	remaining: 27.5s
875:	learn: 0.4012346	total: 3m 12s	remaining: 27.3s
876:	learn: 0.4012262	total: 3m 12s	remaining: 27s
877:	learn: 0.4012231	total: 3m 12s	remaining: 26.8s
878:	learn: 0.4012193	total: 3m 13s	remaining: 26.6s
879:	learn: 0.4012160	total: 3m 13s	remaining: 26.4s
880:	learn: 0.4012107	total: 3m 13s	remaining: 26.2s
881:	learn: 0.4012039	total: 3m 13s	remaining: 25.9s
882:	learn: 0.4011989	total: 3m 14s	remaining: 25.7s
883:	learn: 0.4011959	total: 3m 14s	remaining: 25.5s
884:	learn: 0.4011931	total: 3m 14s	remaining: 25.3s
885:	learn: 0.4011916	total: 3m 14s	remaining: 25.1s
886:	learn: 0.4011881	total: 3m 14s	remaining: 24.8s
887:	learn: 0.4011848	total: 3m 15s	remaining: 24.6s
888:	learn: 0.4011804	total: 3m 15s	remaining: 24.4s
889:	learn: 0.4011764	total: 3m 15s	remaining: 24.2s
890:	learn: 0.4011700	total: 3m 15s	remaining: 24s
891:	learn: 0.4011665	total: 3m 16s	remaining: 23.7s
892:	learn: 0.4011632	total: 3m 16s	remaining: 23.5s
893:	learn: 0.4011601	total: 3m 16s	remaining: 23.3s
894:	learn: 0.4011591	total: 3m 16s	remaining: 23.1s
895:	learn: 0.4011551	total: 3m 16s	remaining: 22.9s
896:	learn: 0.4011517	total: 3m 17s	remaining: 22.6s
897:	learn: 0.4011462	total: 3m 17s	remaining: 22.4s
898:	learn: 0.4011418	total: 3m 17s	remaining: 22.2s
899:	learn: 0.4011396	total: 3m 17s	remaining: 22s
900:	learn: 0.4011369	total: 3m 17s	remaining: 21.8s
901:	learn: 0.4011350	total: 3m 18s	remaining: 21.5s
902:	learn: 0.4011314	total: 3m 18s	remaining: 21.3s
903:	learn: 0.4011271	total: 3m 18s	remaining: 21.1s
904:	learn: 0.4011235	total: 3m 18s	remaining: 20.9s
905:	learn: 0.4011203	total: 3m 19s	remaining: 20.7s
906:	learn: 0.4011158	total: 3m 19s	remaining: 20.4s
907:	learn: 0.4011128	total: 3m 19s	remaining: 20.2s
908:	learn: 0.4011078	total: 3m 19s	remaining: 20s
909:	learn: 0.4011048	total: 3m 19s	remaining: 19.8s
910:	learn: 0.4011017	total: 3m 20s	remaining: 19.6s
911:	learn: 0.4010986	total: 3m 20s	remaining: 19.3s
912:	learn: 0.4010958	total: 3m 20s	remaining: 19.1s
913:	learn: 0.4010938	total: 3m 20s	remaining: 18.9s
914:	learn: 0.4010915	total: 3m 20s	remaining: 18.7s
915:	learn: 0.4010890	total: 3m 21s	remaining: 18.5s
916:	learn: 0.4010866	total: 3m 21s	remaining: 18.2s
917:	learn: 0.4010818	total: 3m 21s	remaining: 18s
918:	learn: 0.4010787	total: 3m 21s	remaining: 17.8s
919:	learn: 0.4010766	total: 3m 22s	remaining: 17.6s
920:	learn: 0.4010721	total: 3m 22s	remaining: 17.4s
921:	learn: 0.4010696	total: 3m 22s	remaining: 17.1s

922:	learn: 0.4010624	total: 3m 22s	remaining: 16.9s
923:	learn: 0.4010603	total: 3m 22s	remaining: 16.7s
924:	learn: 0.4010581	total: 3m 23s	remaining: 16.5s
925:	learn: 0.4010561	total: 3m 23s	remaining: 16.3s
926:	learn: 0.4010481	total: 3m 23s	remaining: 16s
927:	learn: 0.4010457	total: 3m 23s	remaining: 15.8s
928:	learn: 0.4010420	total: 3m 23s	remaining: 15.6s
929:	learn: 0.4010401	total: 3m 24s	remaining: 15.4s
930:	learn: 0.4010350	total: 3m 24s	remaining: 15.1s
931:	learn: 0.4010328	total: 3m 24s	remaining: 14.9s
932:	learn: 0.4010302	total: 3m 24s	remaining: 14.7s
933:	learn: 0.4010278	total: 3m 25s	remaining: 14.5s
934:	learn: 0.4010234	total: 3m 25s	remaining: 14.3s
935:	learn: 0.4010209	total: 3m 25s	remaining: 14s
936:	learn: 0.4010126	total: 3m 25s	remaining: 13.8s
937:	learn: 0.4010085	total: 3m 25s	remaining: 13.6s
938:	learn: 0.4010022	total: 3m 26s	remaining: 13.4s
939:	learn: 0.4009975	total: 3m 26s	remaining: 13.2s
940:	learn: 0.4009920	total: 3m 26s	remaining: 12.9s
941:	learn: 0.4009906	total: 3m 26s	remaining: 12.7s
942:	learn: 0.4009880	total: 3m 26s	remaining: 12.5s
943:	learn: 0.4009843	total: 3m 27s	remaining: 12.3s
944:	learn: 0.4009805	total: 3m 27s	remaining: 12.1s
945:	learn: 0.4009745	total: 3m 27s	remaining: 11.9s
946:	learn: 0.4009715	total: 3m 27s	remaining: 11.6s
947:	learn: 0.4009701	total: 3m 28s	remaining: 11.4s
948:	learn: 0.4009651	total: 3m 28s	remaining: 11.2s
949:	learn: 0.4009610	total: 3m 28s	remaining: 11s
950:	learn: 0.4009577	total: 3m 28s	remaining: 10.8s
951:	learn: 0.4009535	total: 3m 28s	remaining: 10.5s
952:	learn: 0.4009499	total: 3m 29s	remaining: 10.3s
953:	learn: 0.4009459	total: 3m 29s	remaining: 10.1s
954:	learn: 0.4009423	total: 3m 29s	remaining: 9.87s
955:	learn: 0.4009383	total: 3m 29s	remaining: 9.65s
956:	learn: 0.4009354	total: 3m 29s	remaining: 9.43s
957:	learn: 0.4009319	total: 3m 30s	remaining: 9.21s
958:	learn: 0.4009284	total: 3m 30s	remaining: 8.99s
959:	learn: 0.4009220	total: 3m 30s	remaining: 8.77s
960:	learn: 0.4009189	total: 3m 30s	remaining: 8.55s
961:	learn: 0.4009149	total: 3m 31s	remaining: 8.34s
962:	learn: 0.4009116	total: 3m 31s	remaining: 8.12s
963:	learn: 0.4009098	total: 3m 31s	remaining: 7.9s
964:	learn: 0.4009053	total: 3m 31s	remaining: 7.68s
965:	learn: 0.4009020	total: 3m 31s	remaining: 7.46s
966:	learn: 0.4008986	total: 3m 32s	remaining: 7.24s
967:	learn: 0.4008957	total: 3m 32s	remaining: 7.02s
968:	learn: 0.4008931	total: 3m 32s	remaining: 6.8s
969:	learn: 0.4008887	total: 3m 32s	remaining: 6.58s

970:	learn: 0.4008854	total: 3m 32s	remaining: 6.36s
971:	learn: 0.4008822	total: 3m 33s	remaining: 6.14s
972:	learn: 0.4008781	total: 3m 33s	remaining: 5.92s
973:	learn: 0.4008764	total: 3m 33s	remaining: 5.7s
974:	learn: 0.4008710	total: 3m 33s	remaining: 5.48s
975:	learn: 0.4008671	total: 3m 34s	remaining: 5.26s
976:	learn: 0.4008628	total: 3m 34s	remaining: 5.04s
977:	learn: 0.4008539	total: 3m 34s	remaining: 4.82s
978:	learn: 0.4008506	total: 3m 34s	remaining: 4.6s
979:	learn: 0.4008460	total: 3m 34s	remaining: 4.38s
980:	learn: 0.4008407	total: 3m 35s	remaining: 4.17s
981:	learn: 0.4008379	total: 3m 35s	remaining: 3.95s
982:	learn: 0.4008349	total: 3m 35s	remaining: 3.73s
983:	learn: 0.4008288	total: 3m 35s	remaining: 3.51s
984:	learn: 0.4008264	total: 3m 35s	remaining: 3.29s
985:	learn: 0.4008220	total: 3m 36s	remaining: 3.07s
986:	learn: 0.4008187	total: 3m 36s	remaining: 2.85s
987:	learn: 0.4008168	total: 3m 36s	remaining: 2.63s
988:	learn: 0.4008143	total: 3m 36s	remaining: 2.41s
989:	learn: 0.4008076	total: 3m 37s	remaining: 2.19s
990:	learn: 0.4007979	total: 3m 37s	remaining: 1.97s
991:	learn: 0.4007951	total: 3m 37s	remaining: 1.75s
992:	learn: 0.4007883	total: 3m 37s	remaining: 1.53s
993:	learn: 0.4007845	total: 3m 37s	remaining: 1.31s
994:	learn: 0.4007806	total: 3m 38s	remaining: 1.1s
995:	learn: 0.4007787	total: 3m 38s	remaining: 877ms
996:	learn: 0.4007758	total: 3m 38s	remaining: 658ms
997:	learn: 0.4007712	total: 3m 38s	remaining: 438ms
998:	learn: 0.4007676	total: 3m 38s	remaining: 219ms
999:	learn: 0.4007641	total: 3m 39s	remaining: 0us

```
[ ]: <catboost.core.CatBoostClassifier at 0x7fc0acbd3fa0>
```

```
[ ]: y_pred_ctb=ctb.predict_proba(X_val)[:, 1]
```

```
[ ]: log_loss(y_val, y_pred_ctb)
```

```
[ ]: 0.40239392328189716
```

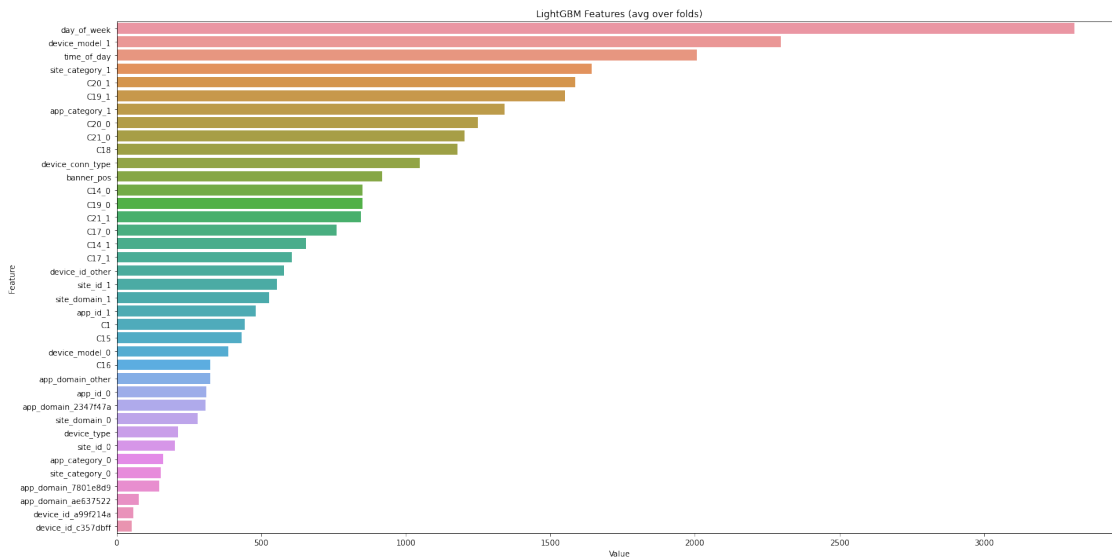
## 4 Final Result

Lightgbm

## 4.1 Feature Importance

```
[ ]: feature_imp = pd.DataFrame(sorted(zip(lgb.feature_importances_,X_train.
    ↪columns)), columns=['Value', 'Feature'])

[ ]: plt.figure(figsize=(20, 10))
sns.barplot(x="Value", y="Feature", data=feature_imp.sort_values(by="Value",
    ↪ascending=False))
plt.title('LightGBM Features (avg over folds)')
plt.tight_layout()
plt.show()
plt.savefig('lgbm_importances-01.png')
```



<Figure size 432x288 with 0 Axes>

## 4.2 ROC

```
[ ]: ##### Import Libraries & Modules
    ↪#####

from sklearn.metrics import roc_curve, auc
from numpy import interp
from sklearn.model_selection import StratifiedKFold
lgb = LGBMClassifier(learning_rate=0.1,num_leaves = 300,random_state=42, metric_
    ↪='binary_logloss',cat_smooth= 35)

[ ]: ##### Cross - Validation
    ↪#####
```

```
# This cross-validation object is a variation of KFold that returns stratified
↳ folds
# The folds are made by preserving the percentage of samples for each class
cv = list(StratifiedKFold(n_splits=5, # number of folds. Must
↳ be at least 2

).split(X_train, y_train)) # generate indices to
↳ split data into training and test set
```

```
[ ]: ##### Visualization
↳ #####

fig = plt.figure(figsize=(7, 5)) # set figure size

mean_tpr = 0.0
mean_fpr = np.linspace(0, 1, 100)
all_tpr = []

for i, (train, test) in enumerate(cv): # enumerate allows us to loop over
↳ something and have an automatic counter (e.g, 0, 1, 2, etc.)
    probas = lgb.fit(X_train.iloc[train],y_train.iloc[train]).
    ↳ predict_proba(X_train.iloc[test]) # make predictions based on classifiers

    # roc_curve will compute Receiver operating characteristic (ROC)
    fpr, tpr, thresholds = roc_curve(y_train.iloc[test], # data for ROC curves
    ↳ (true labels)

                                probas[:, 1], # predictions based on
    ↳ estimators

                                pos_label=1) # the label of the positive
    ↳ class

    mean_tpr += interp(mean_fpr, fpr, tpr) # one-dimensional linear
    ↳ interpolation (continuous ROC curve)

    mean_tpr[0] = 0.0
    # auc will compute Area Under the Curve (AUC)
    roc_auc = auc(fpr, tpr)
    plt.plot(fpr, # the horizontal
    ↳ coordinates of the data points

            tpr, # the vertical coordinates
    ↳ of the data points

            label='ROC fold %d (area = %0.2f)'
                % (i+1, roc_auc))

plt.plot([0, 1], # plot random guessing
    ↳ classifier

        [0, 1],

        linestyle='--',
```



```

        color=(0.6, 0.6, 0.6),
        label='random guessing')

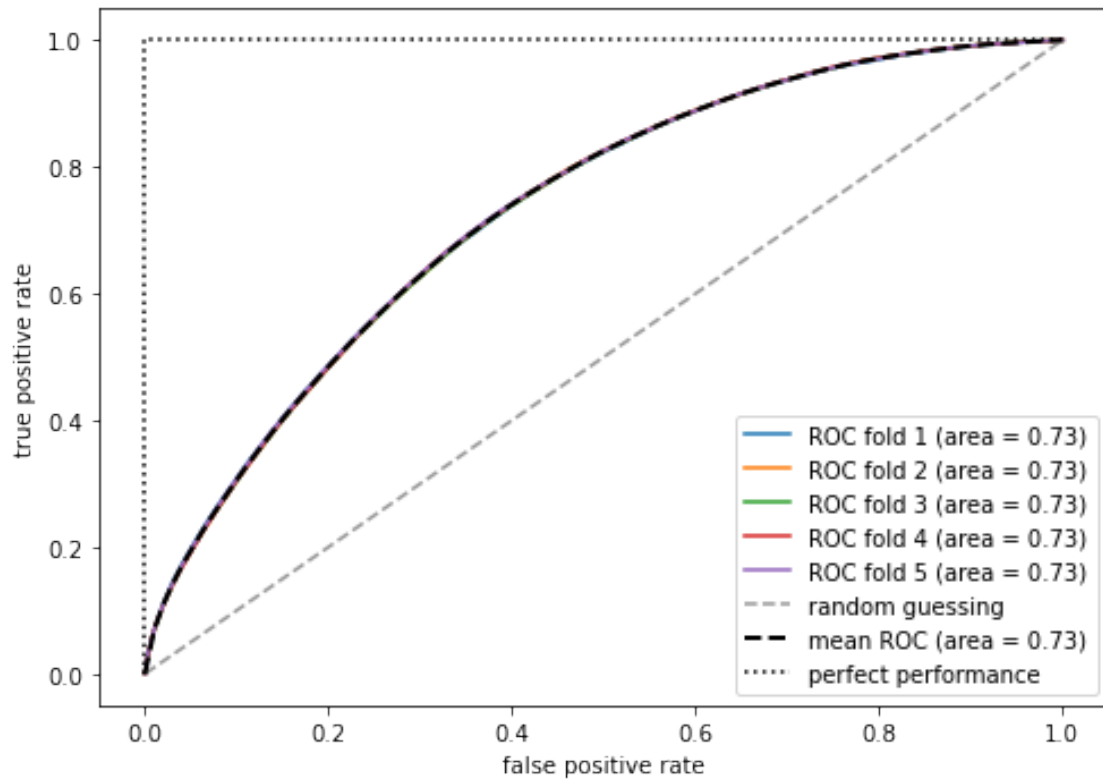
mean_tpr /= len(cv)                                # plot mearn ROC curve
mean_tpr[-1] = 1.0
mean_auc = auc(mean_fpr, mean_tpr)
plt.plot(mean_fpr,
         mean_tpr,
         'k--',
         label='mean ROC (area = %0.2f)' % mean_auc, lw=2)

plt.plot([0, 0, 1],                                # plot perfect classifier
         [0, 1, 1],
         linestyle=':',
         color='black',
         label='perfect performance')

# Figure paramaters: x axis limits, y axis limits, labels of axes, legend_
↪position
plt.xlim([-0.05, 1.05])
plt.ylim([-0.05, 1.05])
plt.xlabel('false positive rate')
plt.ylabel('true positive rate')
plt.legend(loc="lower right")

plt.tight_layout()
# plt.savefig('ROC_CrossValidation_Oneclassifier.png', dpi=300)
plt.show()                                         # display figure

```



### 4.3 Submission

```
[ ]: y_sub = lgb.predict_proba(X_test)[: ,1]
```

```
[ ]: # Import the submission data
sub = pd.read_csv('Project Data/ProjectSubmission-Team4.csv')
```

```
[ ]: sub.iloc[: ,1] = y_sub
```

```
[ ]: sub.to_csv('ProjectSubmission-Team4.csv', index = False)
```

```
[ ]:
```