

# Import required packages

In [169]:

```

import numpy as np
import pandas as pd

from sklearn.model_selection import train_test_split, RandomizedSearchCV, GridSearchCV,
cross_val_score
from sklearn.preprocessing import StandardScaler
from sklearn.pipeline import Pipeline
from sklearn.decomposition import PCA
from sklearn.metrics import accuracy_score, plot_confusion_matrix, classification_report
from sklearn.feature_selection import SelectFromModel
from sklearn.neighbors import KNeighborsClassifier
from sklearn.naive_bayes import GaussianNB
from sklearn.ensemble import RandomForestClassifier, ExtraTreesClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from sklearn.neighbors import KNeighborsClassifier

from skopt import BayesSearchCV
from skopt.space import Integer, Categorical, Real

import xgboost as xgb
import lightgbm as lgb
import catboost as ctb

from mlxtend.classifier import StackingClassifier, StackingCVClassifier

from pprint import pprint
import matplotlib.pyplot as plt
from matplotlib.ticker import FuncFormatter
import seaborn as sns

RANDOM_SEED = 33

```

## 1. Load data and show the summary statistics

In [45]:

```

data = pd.read_csv(
    'dataset/parkinsons.csv') # Load data and convert to DataFrame object

X = data.drop(['id', 'class'], axis=1) # extracts feature data from the data
y = data['class'].to_numpy(
    np.int32) # extracts labels for each samples and covert to an array

FEATURES = X.columns.to_numpy(
) # extracts feature names and convert to an array

```

In [46]:

```

labels, label_counts = np.unique(
    y, return_counts=True) # counts the number of each label

for i in range(len(labels)):

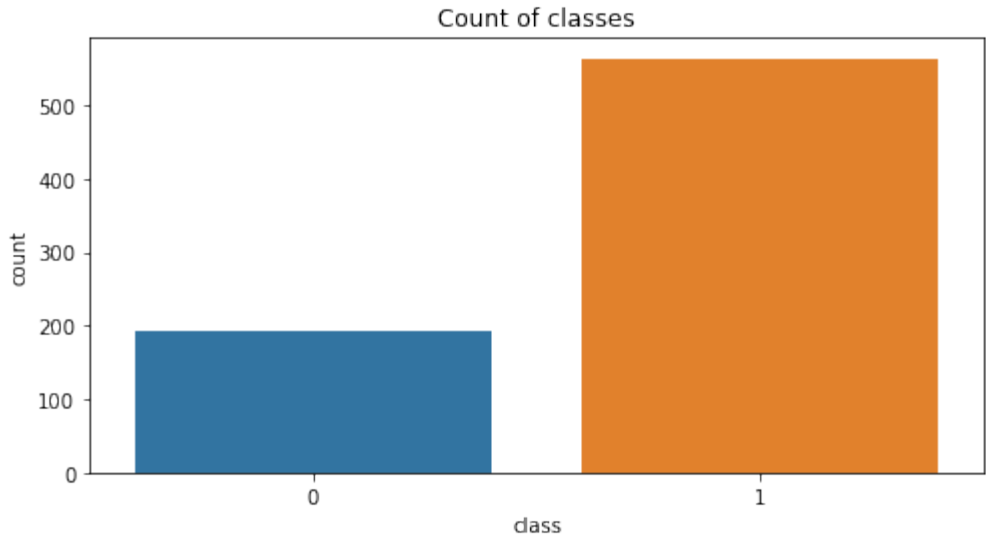
```

```
print('Number of samples labelled to {}: {}'.format(
    labels[i], label_counts[i]))

print('Number of total samples: {}'.format(sum(label_counts)))
print('Number of total features: {}'.format(len(FEATURES)))

plt.figure(figsize=(8, 4))
sns.countplot(x='class', data=data)
plt.title('Count of classes')
plt.show()
```

Number of samples labelled to 0: 192  
Number of samples labelled to 1: 564  
Number of total samples: 756  
Number of total features: 753



In [47]:

```
X.describe() # display the summary stats of the dataset
```

Out[47]:

	gender	PPE	DFA	RPDE	numPulses	numPeriodsPulses	meanPeriodPulses	stdDevPeriodPulses	locP
count	756.000000	756.000000	756.000000	756.000000	756.000000	756.000000	756.000000	756.000000	756.000000
mean	0.515873	0.746284	0.700414	0.489058	323.972222	322.678571	0.006360	0.000383	0.000000
std	0.500079	0.169294	0.069718	0.137442	99.219059	99.402499	0.001826	0.000728	0.000000
min	0.000000	0.041551	0.543500	0.154300	2.000000	1.000000	0.002107	0.000011	0.000000
25%	0.000000	0.762833	0.647053	0.386537	251.000000	250.000000	0.005003	0.000049	0.000000
50%	1.000000	0.809655	0.700525	0.484355	317.000000	316.000000	0.006048	0.000077	0.000000
75%	1.000000	0.834315	0.754985	0.586515	384.250000	383.250000	0.007528	0.000171	0.000000
max	1.000000	0.907660	0.852640	0.871230	907.000000	905.000000	0.012966	0.003483	0.000000

8 rows × 753 columns

In [48]:

```
data.drop(['id'], axis=1).head() # Show the details of first 5 samples
```

Out[48]:

	gender	PPE	DFA	RPDE	numPulses	numPeriodsPulses	meanPeriodPulses	stdDevPeriodPulses	locPctJitter	locAbsJitt
0	1	0.85247	0.71826	0.57227	240	239	0.008064	0.000087	0.00218	0.0000
1	1	0.76686	0.69481	0.53966	234	233	0.008258	0.000073	0.00195	0.0000
2	1	0.85083	0.67604	0.58982	232	231	0.008340	0.000060	0.00176	0.0000
3	0	0.41121	0.79672	0.59257	178	177	0.010858	0.000183	0.00419	0.0000
4	0	0.32790	0.79782	0.53028	236	235	0.008162	0.002669	0.00535	0.0000

5 rows × 754 columns

### 3. Split samples to training dataset and test dataset

In [49]:

```
# Split samples to training data (80%) and test data(20%)
X_trn, X_tst, y_train, y_test = train_test_split(X,
                                                y,
                                                test_size=0.2,
                                                random_state=RANDOM_SEED)
```

### 3. Data Preprocessing

Transform data to usable and meaningful information using [sklearn.preprocessing.StandardScaler](#)

$$x' = \frac{x - \bar{X}}{S}$$

In [50]:

```
## Standardize features
scaler = StandardScaler().fit(X_trn)
X_train = scaler.transform(X_trn)
X_test = scaler.transform(X_tst)
```

### 4. Feature Selection

- 1. Calculate the feature weights with [sklearn.ensemble.RandomForestClassifier](#)
- 2. Remove unrelated/unimportant feature with [sklearn.feature\\_selection.SelectFromModel](#)

#### 4.1 Train the data with random forest classifier

In [51]:

```
clf = RandomForestClassifier(n_estimators=200,
                            random_state=RANDOM_SEED,
                            n_jobs=-1)

clf.fit(X_train, y_train)
```

Out[51]:

```
RandomForestClassifier(n_estimators=200, n_jobs=-1, random state=33)
```

In [52]:

```
# print the feature name and corresponding feature importance
for feature in zip(FEATURES, clf.feature_importances_):
```

```

print(feature)

('gender', 0.0)
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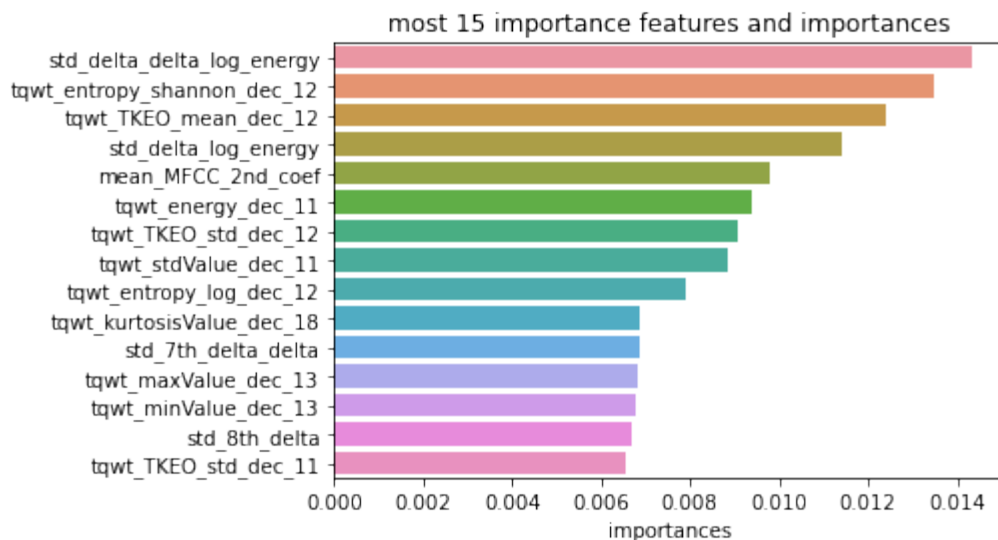


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```

In [53]:

```
# Get the indices of the most 15 important features
sort_i = np.argsort(clf.feature_importances_ * -1)[:15]
```

```
# Plot the most 15 important features and importances
sns.barplot(y=FEATURES[sort_i], x=clf.feature_importances_[sort_i], orient='h')
plt.xlabel('importances')
plt.title('most 15 importance features and importances')
plt.show()
```



## 4.2 Identify and select most important features

In [54]:

```
# Create a selector object that will use the random forest classifier to identify features
sfm = SelectFromModel(clf)

# Train the selector
sfm.fit(X_train, y_train)
```

Out[54]:

```
SelectFromModel(estimator=RandomForestClassifier(n_estimators=200, n_jobs=-1,
                                                  random_state=33))
```

In [55]:

```
print('Threshold of the feature importance: {}'.format(sfm.threshold_))
```

```
Threshold of the feature importance: 0.0013280212483399738
```

In [56]:

```
# Get the index list of the selected features
selected_i = sfm.get_support(indices=True)
```

In [57]:

```
# Print the name of selected features
IMPORTANT_FEATURES = FEATURES[selected_i]
print(IMPORTANT_FEATURES)
```

```
['PPE' 'DFA' 'locPctJitter' 'locAbsJitter' 'rapJitter' 'ddpJitter'
 'locDbShimmer' 'apq11Shimmer' 'minIntensity' 'meanIntensity' 'f1'
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 'tqwt\_kurtosisValue\_dec\_18' 'tqwt\_kurtosisValue\_dec\_19'  
 'tqwt\_kurtosisValue\_dec\_20' 'tqwt\_kurtosisValue\_dec\_22'  
 'tqwt\_kurtosisValue\_dec\_23' 'tqwt\_kurtosisValue\_dec\_26'  
 'tqwt\_kurtosisValue\_dec\_27' 'tqwt\_kurtosisValue\_dec\_28'  
 'tqwt\_kurtosisValue\_dec\_31' 'tqwt\_kurtosisValue\_dec\_33'

```
'tqwt kurtosisValue dec 34' 'tqwt kurtosisValue dec 36']
```

## 4.3 Create new dataset with the selected features

In [58]:

```
X_important_train = sfm.transform(X_train)
X_important_test = sfm.transform(X_test)
print('The shape of X_train changed from {} to {}'.format(
    X_train.shape, X_important_train.shape))
print('\nThe shape of X_test changed from {} to {}'.format(
    X_test.shape, X_important_test.shape))
```

The shape of X\_train changed from (604, 753) to (604, 220)

The shape of X test changed from (152, 753) to (152, 220)

## 5. Extra Trees Classifier Tuning using Random Hyperparameter Search

### 5.1 Ste up Random Hyperparameter Grid

To use RandomizedSearchCV, we first need to create a parameter grid On each iteration, the algorithm will choose a difference combination of the features. There are overall  $5 \times 6 \times 4 \times 3 \times 3 = 1080$  settings. However, the benefit of a random search is that we are not trying every combination, but selecting at random to sample a wide range of values.

In [59]:

```
# Create Parameter Grid
param_grid = {
    'n_estimators': [100, 150, 200, 250,
                    300], # Number of trees in random forest
    'max_features': ['auto', 'log2', 0.25, 0.5, 0.75,
                    1], # Number of features to consider at every split
    'max_depth': [40, 70, 100, None], # Maximum number of levels in tree
    'min_samples_split':
    [2, 5, 10], # Minimum number of samples required to split a node
    'min_samples_leaf': [1, 2, 4]
}

pprint(param_grid)

{'max_depth': [40, 70, 100, None],
 'max_features': ['auto', 'log2', 0.25, 0.5, 0.75, 1],
 'min_samples_leaf': [1, 2, 4],
 'min_samples_split': [2, 5, 10],
 'n_estimators': [100, 150, 200, 250, 300]}
```

### 5.2 Random Search Training

Random search of parameter distributions, using 5-fold cross validation, search acrosss 100 different combinations, and use all available cores for [sklearn.ensemble.ExtraTreesClassifier](#)

In [60]:

```
%%time
# Create a Random forest classifier object
xtrees_model = ExtraTreesClassifier(random_state=RANDOM_SEED, n_jobs=-1)

# Create a RandomGridSearchCV object
rs_xtrees = RandomizedSearchCV(xtrees_model,
                               param_distributions=param_grid,
                               n_iter=100,
```

```
cv=5,
verbose=2,
n_jobs=-1,
random_state=RANDOM_SEED)
```

```
rs_xtrees.fit(X_important_train, y_train)
rs_xtrees.best_params_, rs_xtrees.best_score_
```

```
Fitting 5 folds for each of 100 candidates, totalling 500 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done 25 tasks      | elapsed: 4.1s
[Parallel(n_jobs=-1)]: Done 146 tasks    | elapsed: 21.5s
[Parallel(n_jobs=-1)]: Done 349 tasks    | elapsed: 43.2s
[Parallel(n_jobs=-1)]: Done 500 out of 500 | elapsed: 59.3s finished
Wall time: 59.8 s
```

Out[60]:

```
({'n_estimators': 200,
  'min_samples_split': 2,
  'min_samples_leaf': 1,
  'max_features': 0.5,
  'max_depth': None},
0.8989256198347106)
```

### 5.3 Description of best extra trees classifier

In [61]:

```
# Get the best parameter list
print('The parameter list for the best estimator')
best_xtrees = rs_xtrees.best_estimator_
pprint(best_xtrees.get_params())
```

```
The parameter list for the best estimator
{'bootstrap': False,
 'ccp_alpha': 0.0,
 'class_weight': None,
 'criterion': 'gini',
 'max_depth': None,
 'max_features': 0.5,
 'max_leaf_nodes': None,
 'max_samples': None,
 'min_impurity_decrease': 0.0,
 'min_impurity_split': None,
 'min_samples_leaf': 1,
 'min_samples_split': 2,
 'min_weight_fraction_leaf': 0.0,
 'n_estimators': 200,
 'n_jobs': -1,
 'oob_score': False,
 'random_state': 33,
 'verbose': 0,
 'warm_start': False}
```

### 5.4 Analysis the performance of the best estimator

In [62]:

```
print('Training accuracy: {:.2f}%'.format(
    best_xtrees.score(X_important_train, y_train) * 100))
print('Test accuracy:      {:.2f}%'.format(
    best_xtrees.score(X_important_test, y_test) * 100))
```

```
Training accuracy: 100.00%
Test accuracy:     91.45%
```

In [63]:

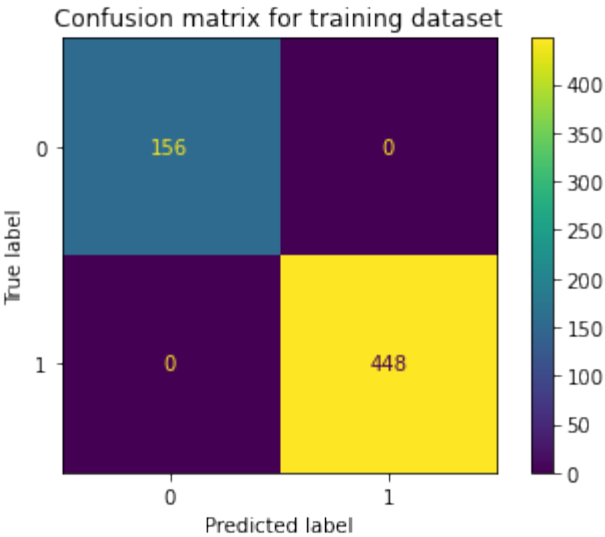
```
print('Performance on training data:\n')
```



```
print(
    classification_report(y_train,
                          best_xtrees.predict(X_important_train),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_xtrees, X_important_train, y_train, labels=[0, 1])
plt.title('Confusion matrix for training dataset')
plt.show()
```

Performance on training data:

	precision	recall	f1-score	support
0	1.0000	1.0000	1.0000	156
1	1.0000	1.0000	1.0000	448
accuracy			1.0000	604
macro avg	1.0000	1.0000	1.0000	604
weighted avg	1.0000	1.0000	1.0000	604

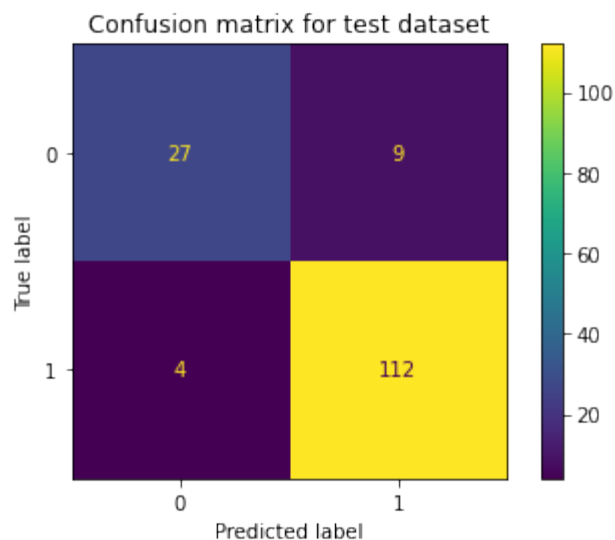


In [64]:

```
print('Performance on test data:\n')
print(
    classification_report(y_test,
                          best_xtrees.predict(X_important_test),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_xtrees, X_important_test, y_test, labels=[0, 1])
plt.title('Confusion matrix for test dataset')
plt.show()
```

Performance on test data:

	precision	recall	f1-score	support
0	0.8710	0.7500	0.8060	36
1	0.9256	0.9655	0.9451	116
accuracy			0.9145	152
macro avg	0.8983	0.8578	0.8756	152
weighted avg	0.9127	0.9145	0.9122	152



## 6 Logistic Regression Tuning using Grid Hyperparameter Search

### 6.1 Set up parameter grid

In [65]:

```
param_grid = {
    'penalty': ['l1', 'l2', 'elasticnet'],
    'solver': ['newton-cg', 'lbfgs', 'liblinear', 'sag', 'saga'],
    'C': np.logspace(-5, 0, num=6)
}

pprint(param_grid)

{'C': array([1.e-05, 1.e-04, 1.e-03, 1.e-02, 1.e-01, 1.e+00]),
 'penalty': ['l1', 'l2', 'elasticnet'],
 'solver': ['newton-cg', 'lbfgs', 'liblinear', 'sag', 'saga']}
```

### 6.2 Grid hyperparameter search training

In [66]:

```
lr_model = LogisticRegression(max_iter=10000, random_state=RANDOM_SEED)

gs_lr = GridSearchCV(lr_model,
                     param_grid=param_grid,
                     cv=5,
                     verbose=2,
                     n_jobs=-1)

gs_lr.fit(X_important_train, y_train)
gs_lr.best_params_, gs_lr.best_score_
```

```
Fitting 5 folds for each of 90 candidates, totalling 450 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done 34 tasks | elapsed: 0.0s
[Parallel(n_jobs=-1)]: Done 450 out of 450 | elapsed: 12.4s finished
```

Out[66]:

```
({'C': 0.1, 'penalty': 'l1', 'solver': 'liblinear'}, 0.8558953168044077)
```

### 6.3 Parameter list for the best LogisticRegression model

In [67]:

```
best_lr = gs_lr.best_estimator_
print('Paramater list for the estimator')
pprint(best_lr.get_params())
```

Paramater list for the estimator

```
{'C': 0.1,
 'class_weight': None,
 'dual': False,
 'fit_intercept': True,
 'intercept_scaling': 1,
 'l1_ratio': None,
 'max_iter': 10000,
 'multi_class': 'auto',
 'n_jobs': None,
 'penalty': 'l1',
 'random_state': 33,
 'solver': 'liblinear',
 'tol': 0.0001,
 'verbose': 0,
 'warm_start': False}
```

## 6.4 Analysis the performance of the best estimator

In [68]:

```
print('Training accuracy: {:.2f}%'.format(
    best_lr.score(X_important_train, y_train) * 100))
print('Test accuracy:      {:.2f}%'.format(
    best_lr.score(X_important_test, y_test) * 100))
```

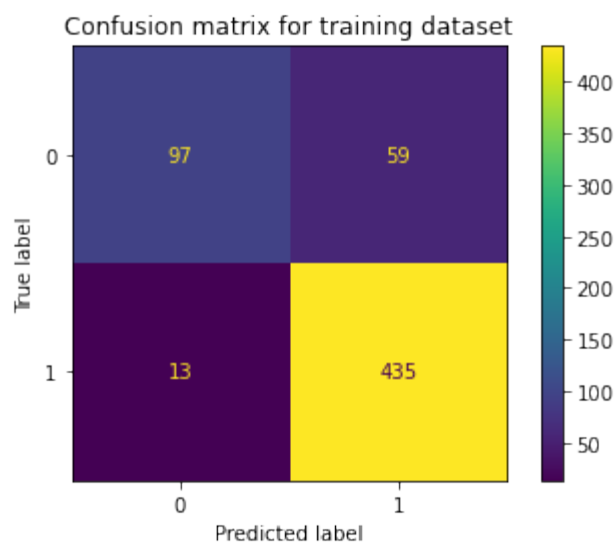
```
Training accuracy: 88.08%
Test accuracy:     91.45%
```

In [69]:

```
print('Performance on training data:\n')
print(
    classification_report(y_train,
                          best_lr.predict(X_important_train),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_lr, X_important_train, y_train, labels=[0, 1])
plt.title('Confusion matrix for training dataset')
plt.show()
```

Performance on training data:

	precision	recall	f1-score	support
0	0.8818	0.6218	0.7293	156
1	0.8806	0.9710	0.9236	448
accuracy			0.8808	604
macro avg	0.8812	0.7964	0.8264	604
weighted avg	0.8809	0.8808	0.8734	604

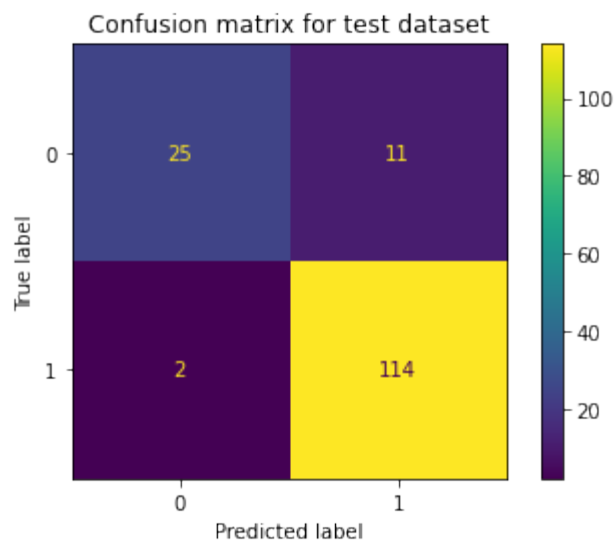


In [70]:

```
print('Performance on test data:\n')
print(
    classification_report(y_test,
                          best_lr.predict(X_important_test),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_lr, X_important_test, y_test, labels=[0, 1])
plt.title('Confusion matrix for test dataset')
plt.show()
```

Performance on test data:

	precision	recall	f1-score	support
0	0.9259	0.6944	0.7937	36
1	0.9120	0.9828	0.9461	116
accuracy			0.9145	152
macro avg	0.9190	0.8386	0.8699	152
weighted avg	0.9153	0.9145	0.9100	152



## 7 SVM Tuning using Random Hyperparameter Search

## 7.1 Set up parameter distribution

In [71]:

```
param_grid = {
    'C': np.logspace(-2, 3, num=6),
    'gamma': np.logspace(-4, 0, num=5),
    'kernel': ['rbf', 'poly', 'sigmoid', 'linear']
}

pprint(param_grid)

{'C': array([1.e-02, 1.e-01, 1.e+00, 1.e+01, 1.e+02, 1.e+03]),
 'gamma': array([1.e-04, 1.e-03, 1.e-02, 1.e-01, 1.e+00]),
 'kernel': ['rbf', 'poly', 'sigmoid', 'linear']}
```

## 7.2 Grid hypterparameter search training

In [72]:

```
svm_model = SVC(max_iter=10000, random_state=RANDOM_SEED)

gs_svm = GridSearchCV(svm_model,
                      param_grid=param_grid,
                      cv=5,
                      verbose=2,
                      n_jobs=-1)

gs_svm.fit(X_important_train, y_train)
gs_svm.best_params_, gs_svm.best_score_

Fitting 5 folds for each of 120 candidates, totalling 600 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done 34 tasks | elapsed: 0.2s
[Parallel(n_jobs=-1)]: Done 504 tasks | elapsed: 3.8s
[Parallel(n_jobs=-1)]: Done 585 out of 600 | elapsed: 4.4s remaining: 0.0s
[Parallel(n_jobs=-1)]: Done 600 out of 600 | elapsed: 4.5s finished

({'C': 100.0, 'gamma': 0.001, 'kernel': 'rbf'}, 0.889090909090909)
```

Out[72]:

## 7.3 Patameter list for the best SVM

In [73]:

```
best_svm = gs_svm.best_estimator_
print('Paramater list for the estimator')
pprint(best_svm.get_params())

Paramater list for the estimator
{'C': 100.0,
 'break_ties': False,
 'cache_size': 200,
 'class_weight': None,
 'coef0': 0.0,
 'decision_function_shape': 'ovr',
 'degree': 3,
 'gamma': 0.001,
 'kernel': 'rbf',
 'max_iter': 10000,
 'probability': False,
 'random_state': 33,
 'shrinking': True,
 'tol': 0.001,
 'verbose': False}
```

## 7.4 Analysis the performance of the best estimator



In [74]:

```
print('Training accuracy: {:.2f}%'.format(
    best_svm.score(X_important_train, y_train) * 100))
print('Test accuracy:      {:.2f}%'.format(
    best_svm.score(X_important_test, y_test) * 100))
```

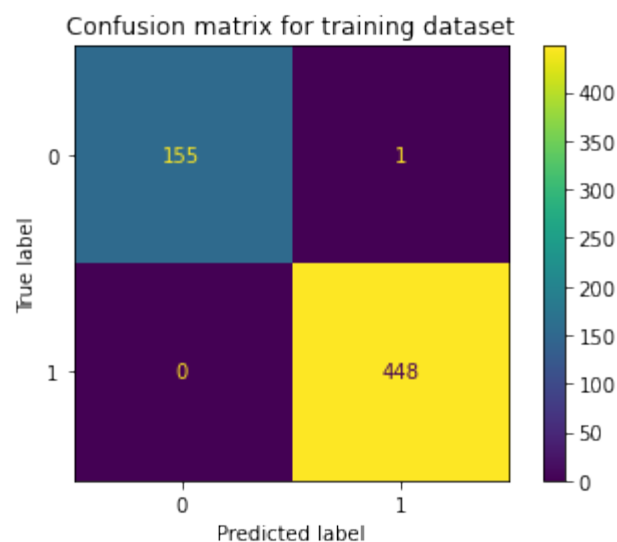
```
Training accuracy: 99.83%
Test accuracy:     92.76%
```

In [75]:

```
print('Performance on training data:\n')
print(
    classification_report(y_train,
                          best_svm.predict(X_important_train),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_svm, X_important_train, y_train, labels=[0, 1])
plt.title('Confusion matrix for training dataset')
plt.show()
```

Performance on training data:

	precision	recall	f1-score	support
0	1.0000	0.9936	0.9968	156
1	0.9978	1.0000	0.9989	448
accuracy			0.9983	604
macro avg	0.9989	0.9968	0.9978	604
weighted avg	0.9983	0.9983	0.9983	604

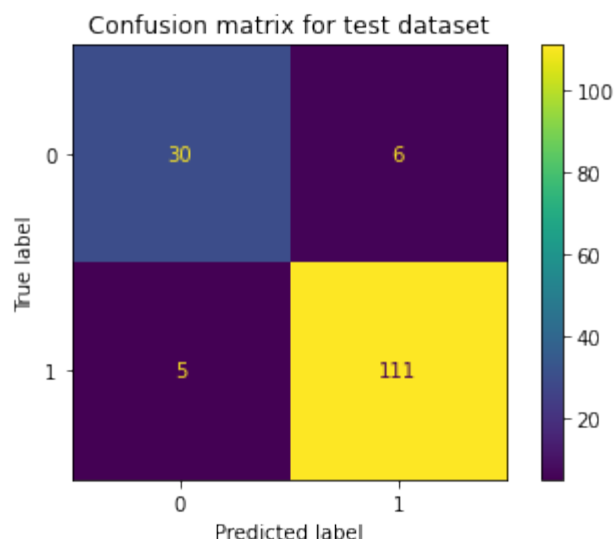


In [76]:

```
print('Performance on test data:\n')
print(
    classification_report(y_test,
                          best_svm.predict(X_important_test),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_svm, X_important_test, y_test, labels=[0, 1])
plt.title('Confusion matrix for test dataset')
plt.show()
```

Performance on test data:

	precision	recall	f1-score	support
0	0.8571	0.8333	0.8451	36
1	0.9487	0.9569	0.9528	116
accuracy			0.9276	152
macro avg	0.9029	0.8951	0.8989	152
weighted avg	0.9270	0.9276	0.9273	152



## 8 XGBoost Tuning using Grid Hyperparameter Search

### 8.1 Find the parameters for the best estimator

#### 8.1.1 Find best max\_depth and max\_depth

In [77]:

```
param_grid1 = {'max_depth': [7, 8, 9], 'min_child_weight': [2, 4, 6, 8]}
```

```
pprint(param_grid1)
```

```
{'max depth': [7, 8, 9], 'min child weight': [2, 4, 6, 8]}
```

In [78]:

```
xgb_clf1 = xgb.XGBClassifier(learning_rate=0.05,
                             n_estimators=10000,
                             gamma=0,
                             objective='binary:logistic',
                             nthread=4,
                             scale_pos_weight=1,
                             seed=RANDOM_SEED)
```

```
gs_xgb1 = GridSearchCV(
    xgb_clf1,
    param_grid=param_grid1,
    verbose=2,
    cv=5,
    n_jobs=-1,
)
```

```
gs_xgb1.fit(X_important_train, y_train)
gs_xgb1.best_params_, gs_xgb1.best_score_
```

```
Fitting 5 folds for each of 12 candidates, totalling 60 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done 25 tasks | elapsed: 2.0min
[Parallel(n_jobs=-1)]: Done 60 out of 60 | elapsed: 3.8min finished
```

Out[78]:

```
({'max_depth': 8, 'min_child_weight': 2}, 0.8973278236914602)
```

In [79]:

```
best_xgb1 = gs_xgb1.best_estimator_
print('Training accuracy: {:.2f}%'.format(
    best_xgb1.score(X_important_train, y_train) * 100))
print('Test accuracy: {:.2f}%'.format(
    best_xgb1.score(X_important_test, y_test) * 100))
```

```
Training accuracy: 100.00%
Test accuracy: 92.11%
```

## 8.1.2 Tune subsample and colsample\_bytree

In [80]:

```
param_grid2 = {
    'subsample': np.arange(0.6, 1, 0.1),
    'colsample_bytree': np.arange(0.6, 1, 0.1)
}
```

```
pprint(param_grid2)
```

```
{'colsample_bytree': array([0.6, 0.7, 0.8, 0.9]),
 'subsample': array([0.6, 0.7, 0.8, 0.9])}
```

In [81]:

```
xgb_clf2 = xgb.XGBClassifier(max_depth=8,
                             min_child_weight=2,
                             learning_rate=0.05,
                             n_estimators=10000,
                             gamma=0,
                             objective='binary:logistic',
                             nthread=4,
                             booster='gbtree',
                             scale_pos_weight=1,
                             seed=RANDOM_SEED)
```

```
gs_xgb2 = GridSearchCV(
    xgb_clf2,
    param_grid=param_grid2,
    verbose=2,
    cv=5,
    n_jobs=-1,
)
```

```
gs_xgb2.fit(X_important_train, y_train)
gs_xgb2.best_params_, gs_xgb2.best_score_
```

```
Fitting 5 folds for each of 16 candidates, totalling 80 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done 25 tasks | elapsed: 1.7min
[Parallel(n_jobs=-1)]: Done 80 out of 80 | elapsed: 4.6min finished
```

Out[81]:

```
({'colsample_bytree': 0.8999999999999999, 'subsample': 0.6},
 0.9089118457300277)
```

In [82]:

```
best_xgb2 = gs_xgb2.best_estimator_
```

```
print('Training accuracy: {:.2f}%'.format(
    best_xgb2.score(X_important_train, y_train) * 100))
print('Test accuracy:      {:.2f}%'.format(
    best_xgb2.score(X_important_test, y_test) * 100))
```

```
Training accuracy: 100.00%
Test accuracy:     90.79%
```

### 8.1.3 Tune learning rate and number of gradient boosted trees

In [83]:

```
param_grid3 = {
    'learning_rate': [0.1, 0.15, 0.2],
    'n_estimators': [400, 600, 800, 1000]
}
pprint(param_grid3)

{'learning_rate': [0.1, 0.15, 0.2], 'n_estimators': [400, 600, 800, 1000]}
```

In [84]:

```
xgb_clf3 = xgb.XGBClassifier(max_depth=4,
                             min_child_weight=2,
                             gamma=0,
                             objective='binary:logistic',
                             colsample_bytree=0.6,
                             subsample=0.9,
                             nthread=4,
                             scale_pos_weight=1,
                             seed=27)
```

```
gs_xgb3 = GridSearchCV(
    xgb_clf3,
    param_grid=param_grid3,
    verbose=2,
    cv=5,
    n_jobs=-1,
)
```

```
gs_xgb3.fit(X_important_train, y_train)
gs_xgb3.best_params_, gs_xgb3.best_score_
```

```
Fitting 5 folds for each of 12 candidates, totalling 60 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done 25 tasks      | elapsed: 10.3s
[Parallel(n_jobs=-1)]: Done 60 out of 60 | elapsed: 20.8s finished
```

Out[84]:

```
({'learning_rate': 0.1, 'n_estimators': 400}, 0.9006473829201103)
```

### 8.2 Parameter list for the best XGBoost Classifier

In [85]:

```
best_xgb = gs_xgb3.best_estimator_
print('Parameter list for the estimator')
pprint(best_xgb.get_params())
```

```
Parameter list for the estimator
{'base_score': 0.5,
 'booster': 'gbtree',
 'colsample_bylevel': 1,
 'colsample_bynode': 1,
 'colsample_bytree': 0.6,
 'gamma': 0,
```

```

'gpu_id': -1,
'importance_type': 'gain',
'interaction_constraints': '',
'learning_rate': 0.1,
'max_delta_step': 0,
'max_depth': 4,
'min_child_weight': 2,
'missing': nan,
'monotone_constraints': '()',
'n_estimators': 400,
'n_jobs': 4,
'nthread': 4,
'num_parallel_tree': 1,
'objective': 'binary:logistic',
'random_state': 27,
'reg_alpha': 0,
'reg_lambda': 1,
'scale_pos_weight': 1,
'seed': 27,
'subsample': 0.9,
'tree_method': 'exact',
'validate_parameters': 1,
'verbosity': None}

```

### 8.3 Analysis the performance of the best estimator

In [86]:

```
best_xgb.fit(X_important_train, y_train, verbose=2)
```

Out[86]:

```

XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1,
              colsample_bynode=1, colsample_bytree=0.6, gamma=0, gpu_id=-1,
              importance_type='gain', interaction_constraints='',
              learning_rate=0.1, max_delta_step=0, max_depth=4,
              min_child_weight=2, missing=nan, monotone_constraints='()',
              n_estimators=400, n_jobs=4, nthread=4, num_parallel_tree=1,
              random_state=27, reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
              seed=27, subsample=0.9, tree_method='exact',
              validate_parameters=1, verbosity=None)

```

In [87]:

```

print('Training accuracy: {:.2f}%'.format(
    best_xgb.score(X_important_train, y_train) * 100))
print('Test accuracy:      {:.2f}%'.format(
    best_xgb.score(X_important_test, y_test) * 100))

```

```

Training accuracy: 100.00%
Test accuracy:     93.42%

```

In [88]:

```

print('Performance on training data:\n')
print(
    classification_report(y_train,
                          best_xgb.predict(X_important_train),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_xgb, X_important_train, y_train, labels=[0, 1])
plt.title('Confusion matrix for training dataset')
plt.show()

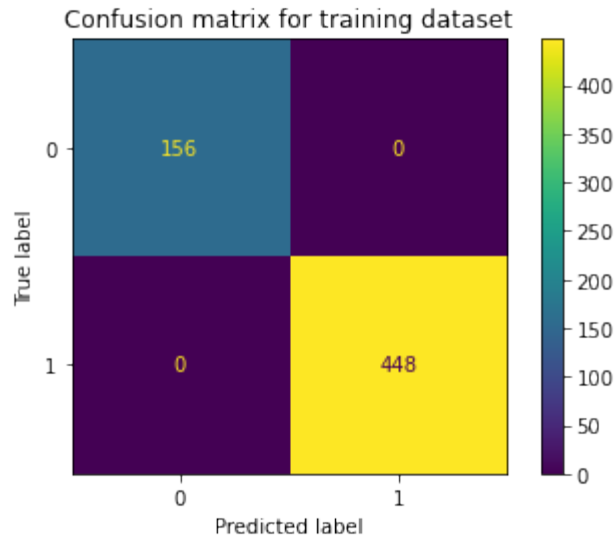
```

Performance on training data:

	precision	recall	f1-score	support
0	1.0000	1.0000	1.0000	156
1	1.0000	1.0000	1.0000	448



accuracy			1.0000	604
macro avg	1.0000	1.0000	1.0000	604
weighted avg	1.0000	1.0000	1.0000	604



In [89]:

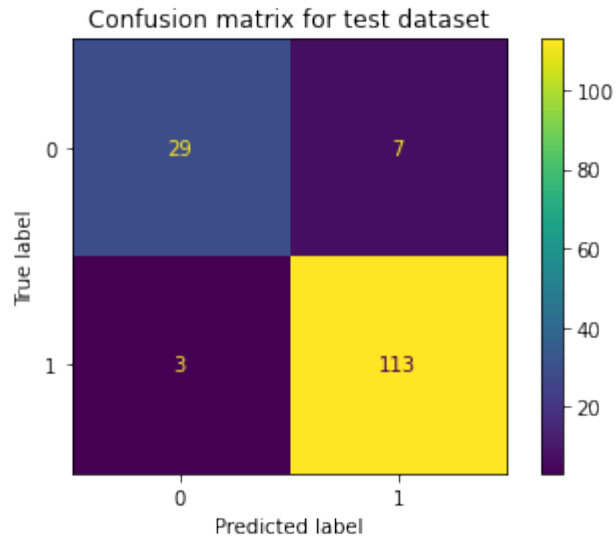
```
print('Performance on test data:\n')
print(
    classification_report(y_test,
                          best_xgb.predict(X_important_test),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_xgb, X_important_test, y_test, labels=[0, 1])
plt.title('Confusion matrix for test dataset')
plt.show()
```

Performance on test data:

	precision	recall	f1-score	support
0	0.9062	0.8056	0.8529	36
1	0.9417	0.9741	0.9576	116

accuracy			0.9342	152
macro avg	0.9240	0.8898	0.9053	152
weighted avg	0.9333	0.9342	0.9328	152



## 9 CatBoost Classifier Tuning

### 9.1 Set up parameter distribution

In [90]:

```
param_grid = {'depth': np.arange(6, 9, 1), 'l2_leaf_reg': [1, 3, 5]}

pprint(param_grid)

{'depth': array([6, 7, 8]), 'l2 leaf reg': [1, 3, 5]}
```

### 9.2 Grid hypterparameter search training

In [91]:

```
catb_model = ctb.CatBoostClassifier(iterations=400,
                                    learning_rate=0.1,
                                    random_seed=RANDOM_SEED)

grid_search_result = catb_model.grid_search(param_grid,
                                             X=X_important_train,
                                             y=y_train,
                                             verbose=2,
                                             cv=5,
                                             partition_random_seed=RANDOM_SEED,
                                             plot=True)

bestTest = 0.2537038718
bestIteration = 125

0:      loss: 0.2537039 best: 0.2537039 (0)      total: 9.36s      remaining: 1m 14s

bestTest = 0.2260130241
bestIteration = 116

bestTest = 0.2376994556
bestIteration = 202

2:      loss: 0.2376995 best: 0.2260130 (1)      total: 27.8s      remaining: 55.7s

bestTest = 0.2548930323
bestIteration = 75

bestTest = 0.2605842385
bestIteration = 95

4:      loss: 0.2605842 best: 0.2260130 (1)      total: 1m 2s      remaining: 50s

bestTest = 0.2392508007
bestIteration = 167

bestTest = 0.2138337157
bestIteration = 70

6:      loss: 0.2138337 best: 0.2138337 (6)      total: 1m 54s      remaining: 32.6s

bestTest = 0.2704721113
bestIteration = 94

bestTest = 0.2440361871
bestIteration = 122
```

```
8:      loss: 0.2440362 best: 0.2138337 (6)      total: 3m 2s      remaining: 0us
Estimating final quality...
```

### 9.3 Parameter list for the best CatBoost Classifier

In [92]:

```
best_ctb = ctb.CatBoostClassifier(random_seed=RANDOM_SEED,
                                  **grid_search_result['params'])
best_ctb.fit(X_important_train, y_train, logging_level='Silent')
```

Out[92]:

```
<catboost.core.CatBoostClassifier at 0x19eb6fddc40>
```

In [93]:

```
best_ctb.get_params()
```

Out[93]:

```
{'depth': 8, 'l2 leaf reg': 1, 'random seed': 33}
```

### 9.4 Analysis the performance of the best estimator

In [94]:

```
print('Training accuracy: {:.2f}%'.format(
    best_ctb.score(X_important_train, y_train) * 100))
print('Test accuracy:      {:.2f}%'.format(
    best_ctb.score(X_important_test, y_test) * 100))
```

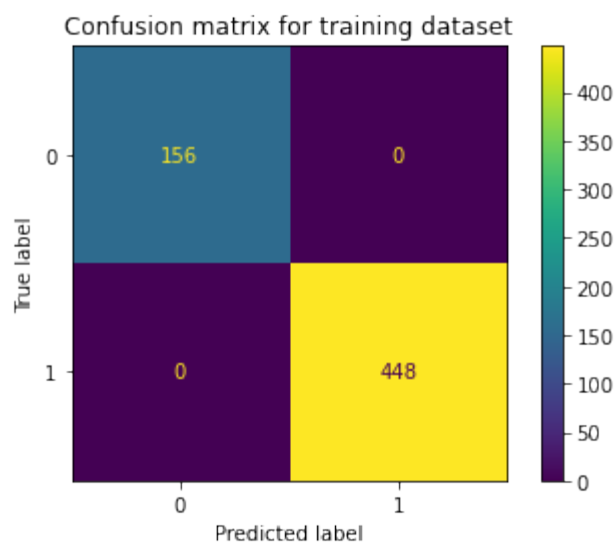
```
Training accuracy: 100.00%
Test accuracy:     93.42%
```

In [95]:

```
print('Performance on training data:\n')
print(
    classification_report(y_train,
                          best_ctb.predict(X_important_train),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_ctb, X_important_train, y_train, labels=[0, 1])
plt.title('Confusion matrix for training dataset')
plt.show()
```

Performance on training data:

	precision	recall	f1-score	support
0	1.0000	1.0000	1.0000	156
1	1.0000	1.0000	1.0000	448
accuracy			1.0000	604
macro avg	1.0000	1.0000	1.0000	604
weighted avg	1.0000	1.0000	1.0000	604

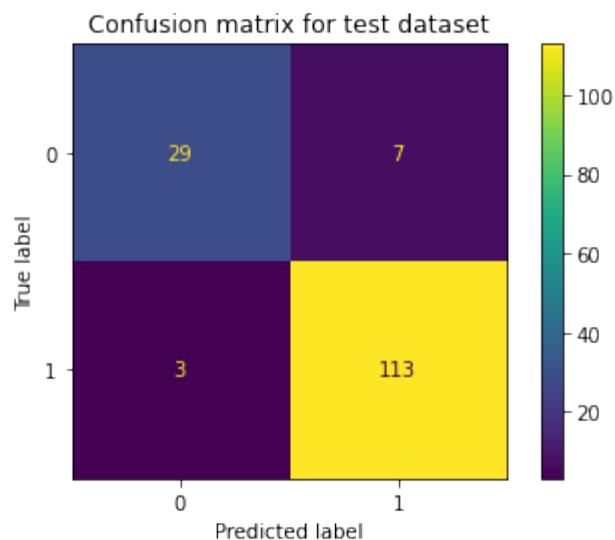


In [96]:

```
print('Performance on test data:\n')
print(
    classification_report(y_test,
                          best_ctb.predict(X_important_test),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_ctb, X_important_test, y_test, labels=[0, 1])
plt.title('Confusion matrix for test dataset')
plt.show()
```

Performance on test data:

	precision	recall	f1-score	support
0	0.9062	0.8056	0.8529	36
1	0.9417	0.9741	0.9576	116
accuracy			0.9342	152
macro avg	0.9240	0.8898	0.9053	152
weighted avg	0.9333	0.9342	0.9328	152



## 10 LGBMClassifier Tuning using Bayes Hyperparameter Search

## 10.1 Set up search spaces

In [97]:

```
search_spaces = {
    'num_leaves': Integer(10, 40),
    'max_depth': Integer(4, 10),
    'boosting_type': Categorical(['gbdt', 'dart']),
    'colsample_bytree': Real(0.8, 1),
    'subsample': Real(0.8, 1),
}

pprint(search_spaces)

{'boosting_type': Categorical(categories=('gbdt', 'dart'), prior=None),
 'colsample_bytree': Real(low=0.8, high=1, prior='uniform', transform='identity'),
 'max_depth': Integer(low=4, high=10, prior='uniform', transform='identity'),
 'num_leaves': Integer(low=10, high=40, prior='uniform', transform='identity'),
 'subsample': Real(low=0.8, high=1, prior='uniform', transform='identity')}
```

## 10.2 Bayes hyperparameter search training

In [98]:

```
lgbm_model = lgb.LGBMClassifier(
    learning_rate=0.1,
    n_estimators=400,
    objective='binary',
    subsample_for_bin=200,
    subsample=0.8,
    subsample_freq=1,
    min_split_gain=0.5,
    min_child_weight=1,
    min_child_samples=5,
    scale_pos_weight=1,
    n_jobs=-1,
    random_state=RANDOM_SEED,
    silent=True,
)

bs_lgbm = BayesSearchCV(lgbm_model,
                        search_spaces=search_spaces,
                        n_iter=300,
                        cv=5,
                        random_state=RANDOM_SEED,
                        n_jobs=-1,
                        verbose=2)

bs_lgbm.fit(X_important_train, y_train)

bs_lgbm.best_score_, bs_lgbm.best_params_

Fitting 5 folds for each of 1 candidates, totalling 5 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done 2 out of 5 | elapsed: 7.3s remaining: 11.0s
[Parallel(n_jobs=-1)]: Done 5 out of 5 | elapsed: 7.3s remaining: 0.0s
[Parallel(n_jobs=-1)]: Done 5 out of 5 | elapsed: 7.3s finished
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
Fitting 5 folds for each of 1 candidates, totalling 5 fits
[Parallel(n_jobs=-1)]: Done 2 out of 5 | elapsed: 0.2s remaining: 0.4s
[Parallel(n_jobs=-1)]: Done 5 out of 5 | elapsed: 1.4s remaining: 0.0s
[Parallel(n_jobs=-1)]: Done 5 out of 5 | elapsed: 1.4s finished
```

[illegible]



[illegible]

[illegible]

[illegible]

[illegible]



[illegible]

[illegible]



[illegible]

[illegible]

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[illegible]

[illegible]

[illegible]

[illegible]



[illegible]

[illegible]



[illegible]

[illegible]

```
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.6s finished
Fitting 5 folds for each of 1 candidates, totalling 5 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done    2 out of    5 | elapsed:    0.6s remaining:    1.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.6s remaining:    0.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.6s finished
Fitting 5 folds for each of 1 candidates, totalling 5 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done    2 out of    5 | elapsed:    0.6s remaining:    1.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.7s remaining:    0.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.7s finished
Fitting 5 folds for each of 1 candidates, totalling 5 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done    2 out of    5 | elapsed:    0.6s remaining:    1.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.7s remaining:    0.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.7s finished
Fitting 5 folds for each of 1 candidates, totalling 5 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done    2 out of    5 | elapsed:    0.6s remaining:    1.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.6s remaining:    0.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.6s finished
Fitting 5 folds for each of 1 candidates, totalling 5 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done    2 out of    5 | elapsed:    0.2s remaining:    0.4s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.2s remaining:    0.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.2s finished
Fitting 5 folds for each of 1 candidates, totalling 5 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done    2 out of    5 | elapsed:    0.6s remaining:    1.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.6s remaining:    0.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.6s finished
Fitting 5 folds for each of 1 candidates, totalling 5 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done    2 out of    5 | elapsed:    0.7s remaining:    1.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.7s remaining:    0.0s
[Parallel(n_jobs=-1)]: Done    5 out of    5 | elapsed:    0.7s finished
```

Out[98]:

```
(0.9122516556291391,
 OrderedDict([('boosting_type', 'dart'),
              ('colsample_bytree', 0.8016779670151458),
              ('max_depth', 10),
              ('num_leaves', 26),
              ('subsample', 0.8012317612884369)]))
```

### 10.3 Patameter list for the best LightGBM classifier

In [99]:

```
best_lgbm = bs_lgbm.best_estimator_
print('Paramater list for the estimator')
pprint(best_lgbm.get_params())
```

```
Paramater list for the estimator
{'boosting_type': 'dart',
 'class_weight': None,
 'colsample_bytree': 0.8016779670151458,
 'importance_type': 'split',
 'learning_rate': 0.1,
 'max_depth': 10,
 'min_child_samples': 5,
 'min_child_weight': 1,
 'min_split_gain': 0.5,
 'n_estimators': 400,
 'n_jobs': -1,
```

```
'num_leaves': 26,
'objective': 'binary',
'random_state': 33,
'reg_alpha': 0.0,
'reg_lambda': 0.0,
'scale_pos_weight': 1,
'silent': True,
'subsample': 0.8012317612884369,
'subsample_for_bin': 200,
'subsample_freq': 1}
```

10.4 Analysis the performance of the best estimator

In [100]:

```
print('Training accuracy: {:.2f}%'.format(
    best_lgbm.score(X_important_train, y_train) * 100))
print('Test accuracy:      {:.2f}%'.format(
    best_lgbm.score(X_important_test, y_test) * 100))
```

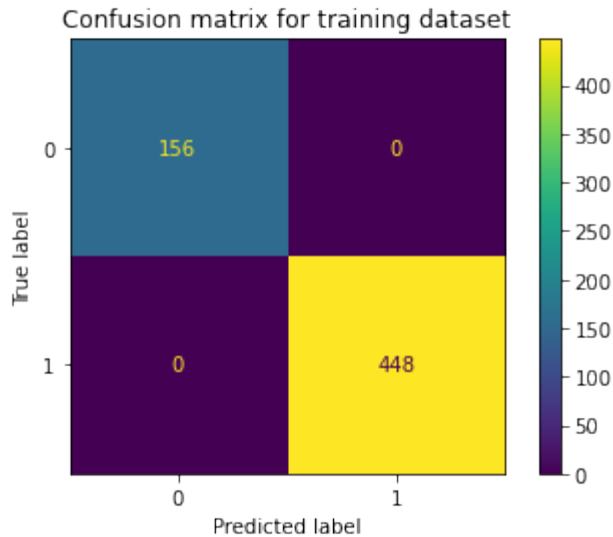
Training accuracy: 100.00%  
Test accuracy: 93.42%

In [101]:

```
print('Performance on training data:\n')
print(
    classification_report(y_train,
                          best_lgbm.predict(X_important_train),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_lgbm, X_important_train, y_train, labels=[0, 1])
plt.title('Confusion matrix for training dataset')
plt.show()
```

Performance on training data:

	precision	recall	f1-score	support
0	1.0000	1.0000	1.0000	156
1	1.0000	1.0000	1.0000	448
accuracy			1.0000	604
macro avg	1.0000	1.0000	1.0000	604
weighted avg	1.0000	1.0000	1.0000	604



In [102]:

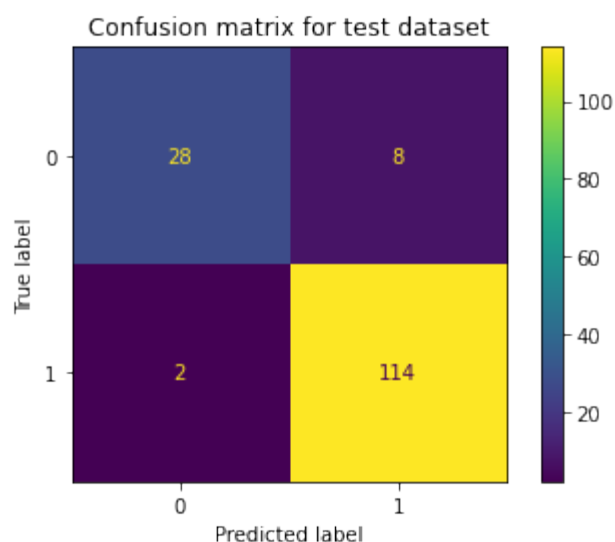
```

print('Performance on test data:\n')
print(
    classification_report(y_test,
                          best_lgbm.predict(X_important_test),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_lgbm, X_important_test, y_test, labels=[0, 1])
plt.title('Confusion matrix for test dataset')
plt.show()

```

Performance on test data:

	precision	recall	f1-score	support
0	0.9333	0.7778	0.8485	36
1	0.9344	0.9828	0.9580	116
accuracy			0.9342	152
macro avg	0.9339	0.8803	0.9032	152
weighted avg	0.9342	0.9342	0.9320	152



## 11 K-nearest Neighbor Classifier Tuning using Grid Hyperparameter Search

### 11.1 Set up search spaces

```

param_grid = {'n_neighbors': np.arange(1, 6), 'p': [1, 2, 3]}

pprint(param_grid)

{'n_neighbors': array([1, 2, 3, 4, 5]), 'p': [1, 2, 3]}

```

In [103]:

### 11.2 Bayes hyperparameter search training

```

knn_clf = KNeighborsClassifier(algorithm='auto', n_jobs=-1)

gs_knn = GridSearchCV(
    knn_clf,
    param_grid=param_grid,
    verbose=2,
    cv=5,

```

In [104]:



```

        n_jobs=-1,
    )

gs_knn.fit(X_important_train, y_train)
gs_knn.best_params_, gs_knn.best_score_

Fitting 5 folds for each of 15 candidates, totalling 75 fits
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
[Parallel(n_jobs=-1)]: Done   34 tasks      | elapsed:    1.1s
[Parallel(n_jobs=-1)]: Done   60 out of   75 | elapsed:    1.9s remaining:    0.4s
[Parallel(n_jobs=-1)]: Done   75 out of   75 | elapsed:    2.4s finished

```

Out[104]:

```

({ 'n_neighbors': 1, 'p': 1}, 0.941969696969697)

```

### 11.3 Patameter list for the best KNN classifier

```

best_knn = gs_knn.best_estimator_
print('Paramater list for the estimator')
pprint(best_knn.get_params())

```

```

Paramater list for the estimator
{'algorithm': 'auto',
 'leaf_size': 30,
 'metric': 'minkowski',
 'metric_params': None,
 'n_jobs': -1,
 'n_neighbors': 1,
 'p': 1,
 'weights': 'uniform'}

```

### 11.4 Analysis the performance of the best estimator

```

print('Training accuracy: {:.2f}%'.format(
    best_knn.score(X_important_train, y_train) * 100))
print('Test accuracy:      {:.2f}%'.format(
    best_knn.score(X_important_test, y_test) * 100))

```

```

Training accuracy: 100.00%
Test accuracy:     97.37%

```

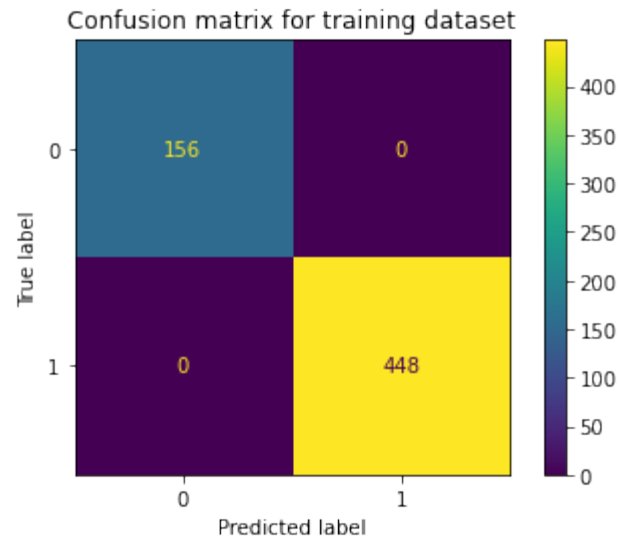
```

print('Performance on training data:\n')
print(
    classification_report(y_train,
                          best_knn.predict(X_important_train),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_knn, X_important_train, y_train, labels=[0, 1])
plt.title('Confusion matrix for training dataset')
plt.show()

```

Performance on training data:

	precision	recall	f1-score	support
0	1.0000	1.0000	1.0000	156
1	1.0000	1.0000	1.0000	448
accuracy			1.0000	604
macro avg	1.0000	1.0000	1.0000	604
weighted avg	1.0000	1.0000	1.0000	604

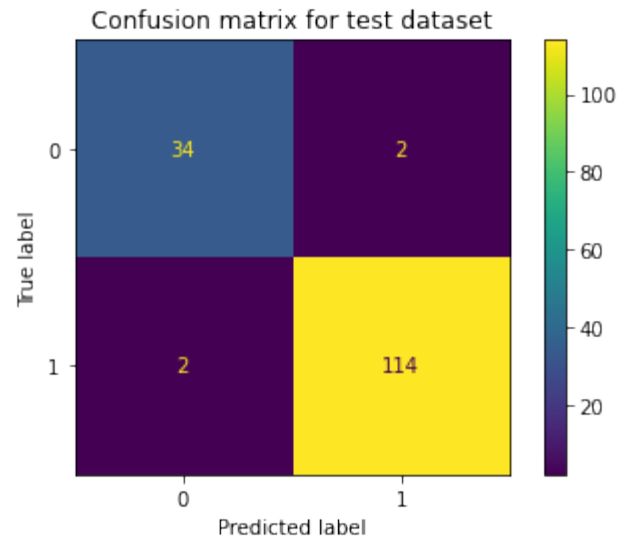


In [108]:

```
print('Performance on test data:\n')
print(
    classification_report(y_test,
                          best_knn.predict(X_important_test),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(best_knn, X_important_test, y_test, labels=[0, 1])
plt.title('Confusion matrix for test dataset')
plt.show()
```

Performance on test data:

	precision	recall	f1-score	support
0	0.9444	0.9444	0.9444	36
1	0.9828	0.9828	0.9828	116
accuracy			0.9737	152
macro avg	0.9636	0.9636	0.9636	152
weighted avg	0.9737	0.9737	0.9737	152



## 12 Combine Estimators by StakcingCVClassifier

The `StackingCVClassifier` extends the standard stacking algorithm (implemented as `StackingClassifier`) using cross-validation to prepare the input data for the level-2 classifier.

## 12.1 Collect the estimators

In [157]:

```
import warnings
warnings.simplefilter('ignore')

estimators = [best_xtrees, best_svm, best_xgb, best_ctb, best_lgbm, best_knn]

est_names = [
    'Extra Trees', 'SVM', 'XGBoost Classifier', 'CatBoost Classifier',
    'LightGBM Classifier', 'KNN', 'StackingCVClassifier'
]
```

## 12.2 Training the data

In [136]:

```
sclf = StackingCVClassifier(classifiers=estimators,
                           meta_classifier=best_lr,
                           cv=5,
                           n_jobs=-1,
                           verbose=0,
                           random_state=RANDOM_SEED
                           )

sclf.fit(X_important_train, y_train)
```

0:	learn: 0.6665931	total: 112ms	remaining: 1m 52s
1:	learn: 0.6369221	total: 199ms	remaining: 1m 39s
2:	learn: 0.6121429	total: 287ms	remaining: 1m 35s
3:	learn: 0.5893073	total: 377ms	remaining: 1m 33s
4:	learn: 0.5652527	total: 472ms	remaining: 1m 33s
5:	learn: 0.5444878	total: 558ms	remaining: 1m 32s
6:	learn: 0.5275610	total: 652ms	remaining: 1m 32s
7:	learn: 0.5080990	total: 744ms	remaining: 1m 32s
8:	learn: 0.4885503	total: 831ms	remaining: 1m 31s
9:	learn: 0.4732062	total: 931ms	remaining: 1m 32s
10:	learn: 0.4574211	total: 1.02s	remaining: 1m 31s
11:	learn: 0.4416857	total: 1.11s	remaining: 1m 31s
12:	learn: 0.4278373	total: 1.2s	remaining: 1m 31s
13:	learn: 0.4137917	total: 1.3s	remaining: 1m 31s
14:	learn: 0.4023856	total: 1.4s	remaining: 1m 31s
15:	learn: 0.3918084	total: 1.5s	remaining: 1m 32s
16:	learn: 0.3808521	total: 1.58s	remaining: 1m 31s
17:	learn: 0.3699718	total: 1.68s	remaining: 1m 31s
18:	learn: 0.3571421	total: 1.76s	remaining: 1m 31s
19:	learn: 0.3466100	total: 1.86s	remaining: 1m 31s
20:	learn: 0.3359778	total: 1.95s	remaining: 1m 30s
21:	learn: 0.3260458	total: 2.03s	remaining: 1m 30s
22:	learn: 0.3160314	total: 2.12s	remaining: 1m 30s
23:	learn: 0.3076384	total: 2.21s	remaining: 1m 29s
24:	learn: 0.2973433	total: 2.31s	remaining: 1m 29s
25:	learn: 0.2910566	total: 2.4s	remaining: 1m 29s
26:	learn: 0.2817621	total: 2.49s	remaining: 1m 29s
27:	learn: 0.2734052	total: 2.59s	remaining: 1m 29s
28:	learn: 0.2658563	total: 2.67s	remaining: 1m 29s
29:	learn: 0.2593366	total: 2.77s	remaining: 1m 29s
30:	learn: 0.2511561	total: 2.87s	remaining: 1m 29s
31:	learn: 0.2436404	total: 2.97s	remaining: 1m 29s
32:	learn: 0.2378629	total: 3.07s	remaining: 1m 29s
33:	learn: 0.2328022	total: 3.15s	remaining: 1m 29s

34:	learn:	0.2268950	total:	3.25s	remaining:	1m 29s
35:	learn:	0.2205209	total:	3.34s	remaining:	1m 29s
36:	learn:	0.2148865	total:	3.44s	remaining:	1m 29s
37:	learn:	0.2109698	total:	3.53s	remaining:	1m 29s
38:	learn:	0.2061818	total:	3.63s	remaining:	1m 29s
39:	learn:	0.2002269	total:	3.72s	remaining:	1m 29s
40:	learn:	0.1945628	total:	3.82s	remaining:	1m 29s
41:	learn:	0.1901853	total:	3.91s	remaining:	1m 29s
42:	learn:	0.1855108	total:	4s	remaining:	1m 29s
43:	learn:	0.1813334	total:	4.09s	remaining:	1m 28s
44:	learn:	0.1774749	total:	4.17s	remaining:	1m 28s
45:	learn:	0.1742857	total:	4.26s	remaining:	1m 28s
46:	learn:	0.1709442	total:	4.35s	remaining:	1m 28s
47:	learn:	0.1672066	total:	4.44s	remaining:	1m 28s
48:	learn:	0.1637110	total:	4.53s	remaining:	1m 27s
49:	learn:	0.1610062	total:	4.62s	remaining:	1m 27s
50:	learn:	0.1574836	total:	4.7s	remaining:	1m 27s
51:	learn:	0.1553096	total:	4.8s	remaining:	1m 27s
52:	learn:	0.1526120	total:	4.89s	remaining:	1m 27s
53:	learn:	0.1505715	total:	4.99s	remaining:	1m 27s
54:	learn:	0.1469025	total:	5.07s	remaining:	1m 27s
55:	learn:	0.1442962	total:	5.16s	remaining:	1m 26s
56:	learn:	0.1410970	total:	5.25s	remaining:	1m 26s
57:	learn:	0.1382398	total:	5.34s	remaining:	1m 26s
58:	learn:	0.1354866	total:	5.43s	remaining:	1m 26s
59:	learn:	0.1334406	total:	5.52s	remaining:	1m 26s
60:	learn:	0.1311775	total:	5.61s	remaining:	1m 26s
61:	learn:	0.1291246	total:	5.71s	remaining:	1m 26s
62:	learn:	0.1267156	total:	5.79s	remaining:	1m 26s
63:	learn:	0.1238921	total:	5.89s	remaining:	1m 26s
64:	learn:	0.1217356	total:	5.98s	remaining:	1m 26s
65:	learn:	0.1195336	total:	6.07s	remaining:	1m 25s
66:	learn:	0.1175291	total:	6.16s	remaining:	1m 25s
67:	learn:	0.1149171	total:	6.25s	remaining:	1m 25s
68:	learn:	0.1131604	total:	6.33s	remaining:	1m 25s
69:	learn:	0.1109330	total:	6.42s	remaining:	1m 25s
70:	learn:	0.1090313	total:	6.51s	remaining:	1m 25s
71:	learn:	0.1076120	total:	6.6s	remaining:	1m 25s
72:	learn:	0.1055416	total:	6.69s	remaining:	1m 24s
73:	learn:	0.1038735	total:	6.78s	remaining:	1m 24s
74:	learn:	0.1022902	total:	6.87s	remaining:	1m 24s
75:	learn:	0.1006721	total:	6.96s	remaining:	1m 24s
76:	learn:	0.0987378	total:	7.05s	remaining:	1m 24s
77:	learn:	0.0973889	total:	7.14s	remaining:	1m 24s
78:	learn:	0.0954455	total:	7.22s	remaining:	1m 24s
79:	learn:	0.0936286	total:	7.31s	remaining:	1m 24s
80:	learn:	0.0923240	total:	7.39s	remaining:	1m 23s
81:	learn:	0.0911380	total:	7.48s	remaining:	1m 23s
82:	learn:	0.0892142	total:	7.57s	remaining:	1m 23s
83:	learn:	0.0881188	total:	7.66s	remaining:	1m 23s
84:	learn:	0.0872838	total:	7.74s	remaining:	1m 23s
85:	learn:	0.0860535	total:	7.83s	remaining:	1m 23s
86:	learn:	0.0848872	total:	7.92s	remaining:	1m 23s
87:	learn:	0.0839426	total:	8.01s	remaining:	1m 22s
88:	learn:	0.0822884	total:	8.1s	remaining:	1m 22s
89:	learn:	0.0808639	total:	8.19s	remaining:	1m 22s
90:	learn:	0.0800833	total:	8.27s	remaining:	1m 22s
91:	learn:	0.0791321	total:	8.36s	remaining:	1m 22s
92:	learn:	0.0780913	total:	8.45s	remaining:	1m 22s
93:	learn:	0.0769898	total:	8.54s	remaining:	1m 22s
94:	learn:	0.0758830	total:	8.63s	remaining:	1m 22s
95:	learn:	0.0749293	total:	8.72s	remaining:	1m 22s
96:	learn:	0.0738071	total:	8.81s	remaining:	1m 22s
97:	learn:	0.0723964	total:	8.91s	remaining:	1m 21s
98:	learn:	0.0711163	total:	8.99s	remaining:	1m 21s
99:	learn:	0.0702033	total:	9.09s	remaining:	1m 21s
100:	learn:	0.0692712	total:	9.18s	remaining:	1m 21s
101:	learn:	0.0682152	total:	9.28s	remaining:	1m 21s
102:	learn:	0.0671951	total:	9.36s	remaining:	1m 21s

103:	learn:	0.0661143	total:	9.46s	remaining:	1m 21s
104:	learn:	0.0651683	total:	9.54s	remaining:	1m 21s
105:	learn:	0.0645015	total:	9.63s	remaining:	1m 21s
106:	learn:	0.0635525	total:	9.72s	remaining:	1m 21s
107:	learn:	0.0629830	total:	9.82s	remaining:	1m 21s
108:	learn:	0.0619736	total:	9.91s	remaining:	1m 21s
109:	learn:	0.0610629	total:	10s	remaining:	1m 20s
110:	learn:	0.0600692	total:	10.1s	remaining:	1m 20s
111:	learn:	0.0594773	total:	10.2s	remaining:	1m 20s
112:	learn:	0.0586926	total:	10.3s	remaining:	1m 20s
113:	learn:	0.0579906	total:	10.4s	remaining:	1m 20s
114:	learn:	0.0574282	total:	10.5s	remaining:	1m 20s
115:	learn:	0.0567268	total:	10.6s	remaining:	1m 20s
116:	learn:	0.0559868	total:	10.6s	remaining:	1m 20s
117:	learn:	0.0552869	total:	10.7s	remaining:	1m 20s
118:	learn:	0.0545884	total:	10.8s	remaining:	1m 20s
119:	learn:	0.0539090	total:	10.9s	remaining:	1m 19s
120:	learn:	0.0528602	total:	11s	remaining:	1m 19s
121:	learn:	0.0522435	total:	11.1s	remaining:	1m 19s
122:	learn:	0.0513587	total:	11.2s	remaining:	1m 19s
123:	learn:	0.0509132	total:	11.3s	remaining:	1m 19s
124:	learn:	0.0505370	total:	11.3s	remaining:	1m 19s
125:	learn:	0.0499324	total:	11.4s	remaining:	1m 19s
126:	learn:	0.0493611	total:	11.5s	remaining:	1m 19s
127:	learn:	0.0486005	total:	11.6s	remaining:	1m 19s
128:	learn:	0.0482156	total:	11.7s	remaining:	1m 18s
129:	learn:	0.0478700	total:	11.8s	remaining:	1m 18s
130:	learn:	0.0473576	total:	11.9s	remaining:	1m 18s
131:	learn:	0.0469390	total:	12s	remaining:	1m 18s
132:	learn:	0.0461538	total:	12.1s	remaining:	1m 18s
133:	learn:	0.0451014	total:	12.1s	remaining:	1m 18s
134:	learn:	0.0444075	total:	12.2s	remaining:	1m 18s
135:	learn:	0.0439429	total:	12.3s	remaining:	1m 18s
136:	learn:	0.0433780	total:	12.4s	remaining:	1m 18s
137:	learn:	0.0430927	total:	12.5s	remaining:	1m 18s
138:	learn:	0.0425382	total:	12.6s	remaining:	1m 17s
139:	learn:	0.0422400	total:	12.7s	remaining:	1m 17s
140:	learn:	0.0417839	total:	12.8s	remaining:	1m 17s
141:	learn:	0.0413530	total:	12.9s	remaining:	1m 17s
142:	learn:	0.0406295	total:	12.9s	remaining:	1m 17s
143:	learn:	0.0401271	total:	13s	remaining:	1m 17s
144:	learn:	0.0396290	total:	13.1s	remaining:	1m 17s
145:	learn:	0.0390866	total:	13.2s	remaining:	1m 17s
146:	learn:	0.0387293	total:	13.3s	remaining:	1m 17s
147:	learn:	0.0381789	total:	13.4s	remaining:	1m 16s
148:	learn:	0.0376551	total:	13.5s	remaining:	1m 16s
149:	learn:	0.0372854	total:	13.6s	remaining:	1m 16s
150:	learn:	0.0368793	total:	13.6s	remaining:	1m 16s
151:	learn:	0.0363714	total:	13.7s	remaining:	1m 16s
152:	learn:	0.0358615	total:	13.8s	remaining:	1m 16s
153:	learn:	0.0354923	total:	13.9s	remaining:	1m 16s
154:	learn:	0.0351307	total:	14s	remaining:	1m 16s
155:	learn:	0.0346607	total:	14.1s	remaining:	1m 16s
156:	learn:	0.0342436	total:	14.2s	remaining:	1m 16s
157:	learn:	0.0338963	total:	14.3s	remaining:	1m 15s
158:	learn:	0.0335239	total:	14.3s	remaining:	1m 15s
159:	learn:	0.0332274	total:	14.4s	remaining:	1m 15s
160:	learn:	0.0330024	total:	14.5s	remaining:	1m 15s
161:	learn:	0.0327476	total:	14.6s	remaining:	1m 15s
162:	learn:	0.0323317	total:	14.7s	remaining:	1m 15s
163:	learn:	0.0320039	total:	14.8s	remaining:	1m 15s
164:	learn:	0.0317032	total:	14.9s	remaining:	1m 15s
165:	learn:	0.0313857	total:	14.9s	remaining:	1m 15s
166:	learn:	0.0309395	total:	15s	remaining:	1m 14s
167:	learn:	0.0306148	total:	15.1s	remaining:	1m 14s
168:	learn:	0.0301048	total:	15.2s	remaining:	1m 14s
169:	learn:	0.0298195	total:	15.3s	remaining:	1m 14s
170:	learn:	0.0295175	total:	15.4s	remaining:	1m 14s
171:	learn:	0.0292236	total:	15.5s	remaining:	1m 14s



172:	learn:	0.0288997	total:	15.6s	remaining:	1m 14s
173:	learn:	0.0285403	total:	15.7s	remaining:	1m 14s
174:	learn:	0.0281333	total:	15.7s	remaining:	1m 14s
175:	learn:	0.0278478	total:	15.8s	remaining:	1m 14s
176:	learn:	0.0275233	total:	15.9s	remaining:	1m 14s
177:	learn:	0.0272564	total:	16s	remaining:	1m 13s
178:	learn:	0.0269426	total:	16.1s	remaining:	1m 13s
179:	learn:	0.0266498	total:	16.2s	remaining:	1m 13s
180:	learn:	0.0264301	total:	16.3s	remaining:	1m 13s
181:	learn:	0.0262303	total:	16.4s	remaining:	1m 13s
182:	learn:	0.0259918	total:	16.4s	remaining:	1m 13s
183:	learn:	0.0257237	total:	16.5s	remaining:	1m 13s
184:	learn:	0.0254217	total:	16.6s	remaining:	1m 13s
185:	learn:	0.0251569	total:	16.7s	remaining:	1m 13s
186:	learn:	0.0248650	total:	16.8s	remaining:	1m 12s
187:	learn:	0.0245975	total:	16.9s	remaining:	1m 12s
188:	learn:	0.0243700	total:	17s	remaining:	1m 12s
189:	learn:	0.0241399	total:	17s	remaining:	1m 12s
190:	learn:	0.0239456	total:	17.1s	remaining:	1m 12s
191:	learn:	0.0237296	total:	17.2s	remaining:	1m 12s
192:	learn:	0.0234593	total:	17.3s	remaining:	1m 12s
193:	learn:	0.0231904	total:	17.4s	remaining:	1m 12s
194:	learn:	0.0229281	total:	17.5s	remaining:	1m 12s
195:	learn:	0.0227019	total:	17.6s	remaining:	1m 12s
196:	learn:	0.0224625	total:	17.7s	remaining:	1m 11s
197:	learn:	0.0222636	total:	17.7s	remaining:	1m 11s
198:	learn:	0.0220648	total:	17.8s	remaining:	1m 11s
199:	learn:	0.0218714	total:	17.9s	remaining:	1m 11s
200:	learn:	0.0215954	total:	18s	remaining:	1m 11s
201:	learn:	0.0214015	total:	18.1s	remaining:	1m 11s
202:	learn:	0.0212099	total:	18.2s	remaining:	1m 11s
203:	learn:	0.0210137	total:	18.3s	remaining:	1m 11s
204:	learn:	0.0208243	total:	18.4s	remaining:	1m 11s
205:	learn:	0.0205551	total:	18.4s	remaining:	1m 11s
206:	learn:	0.0203384	total:	18.5s	remaining:	1m 11s
207:	learn:	0.0202272	total:	18.6s	remaining:	1m 10s
208:	learn:	0.0200951	total:	18.7s	remaining:	1m 10s
209:	learn:	0.0198748	total:	18.8s	remaining:	1m 10s
210:	learn:	0.0197045	total:	18.9s	remaining:	1m 10s
211:	learn:	0.0195176	total:	19s	remaining:	1m 10s
212:	learn:	0.0193795	total:	19.1s	remaining:	1m 10s
213:	learn:	0.0192032	total:	19.2s	remaining:	1m 10s
214:	learn:	0.0190297	total:	19.3s	remaining:	1m 10s
215:	learn:	0.0188329	total:	19.3s	remaining:	1m 10s
216:	learn:	0.0187302	total:	19.4s	remaining:	1m 10s
217:	learn:	0.0185748	total:	19.5s	remaining:	1m 10s
218:	learn:	0.0183820	total:	19.6s	remaining:	1m 9s
219:	learn:	0.0181493	total:	19.7s	remaining:	1m 9s
220:	learn:	0.0180025	total:	19.8s	remaining:	1m 9s
221:	learn:	0.0178509	total:	19.9s	remaining:	1m 9s
222:	learn:	0.0177060	total:	20s	remaining:	1m 9s
223:	learn:	0.0174983	total:	20.1s	remaining:	1m 9s
224:	learn:	0.0173452	total:	20.1s	remaining:	1m 9s
225:	learn:	0.0171867	total:	20.2s	remaining:	1m 9s
226:	learn:	0.0169679	total:	20.3s	remaining:	1m 9s
227:	learn:	0.0167954	total:	20.4s	remaining:	1m 9s
228:	learn:	0.0165651	total:	20.5s	remaining:	1m 8s
229:	learn:	0.0164059	total:	20.6s	remaining:	1m 8s
230:	learn:	0.0162922	total:	20.7s	remaining:	1m 8s
231:	learn:	0.0161450	total:	20.8s	remaining:	1m 8s
232:	learn:	0.0159811	total:	20.8s	remaining:	1m 8s
233:	learn:	0.0158482	total:	20.9s	remaining:	1m 8s
234:	learn:	0.0156980	total:	21s	remaining:	1m 8s
235:	learn:	0.0155717	total:	21.1s	remaining:	1m 8s
236:	learn:	0.0154394	total:	21.2s	remaining:	1m 8s
237:	learn:	0.0153234	total:	21.3s	remaining:	1m 8s
238:	learn:	0.0151606	total:	21.4s	remaining:	1m 8s
239:	learn:	0.0149941	total:	21.5s	remaining:	1m 7s
240:	learn:	0.0148548	total:	21.6s	remaining:	1m 7s

241:	learn:	0.0147540	total:	21.6s	remaining:	1m 7s
242:	learn:	0.0146402	total:	21.7s	remaining:	1m 7s
243:	learn:	0.0145177	total:	21.8s	remaining:	1m 7s
244:	learn:	0.0144201	total:	21.9s	remaining:	1m 7s
245:	learn:	0.0142866	total:	22s	remaining:	1m 7s
246:	learn:	0.0141682	total:	22.1s	remaining:	1m 7s
247:	learn:	0.0140349	total:	22.2s	remaining:	1m 7s
248:	learn:	0.0138885	total:	22.2s	remaining:	1m 7s
249:	learn:	0.0137413	total:	22.3s	remaining:	1m 6s
250:	learn:	0.0136103	total:	22.4s	remaining:	1m 6s
251:	learn:	0.0134533	total:	22.5s	remaining:	1m 6s
252:	learn:	0.0133668	total:	22.6s	remaining:	1m 6s
253:	learn:	0.0132241	total:	22.7s	remaining:	1m 6s
254:	learn:	0.0131111	total:	22.8s	remaining:	1m 6s
255:	learn:	0.0130105	total:	22.9s	remaining:	1m 6s
256:	learn:	0.0128791	total:	22.9s	remaining:	1m 6s
257:	learn:	0.0127822	total:	23s	remaining:	1m 6s
258:	learn:	0.0126790	total:	23.1s	remaining:	1m 6s
259:	learn:	0.0125855	total:	23.2s	remaining:	1m 6s
260:	learn:	0.0124699	total:	23.3s	remaining:	1m 5s
261:	learn:	0.0123810	total:	23.4s	remaining:	1m 5s
262:	learn:	0.0123066	total:	23.5s	remaining:	1m 5s
263:	learn:	0.0121581	total:	23.5s	remaining:	1m 5s
264:	learn:	0.0120654	total:	23.6s	remaining:	1m 5s
265:	learn:	0.0119795	total:	23.7s	remaining:	1m 5s
266:	learn:	0.0118726	total:	23.8s	remaining:	1m 5s
267:	learn:	0.0117180	total:	23.9s	remaining:	1m 5s
268:	learn:	0.0116024	total:	24s	remaining:	1m 5s
269:	learn:	0.0114919	total:	24.1s	remaining:	1m 5s
270:	learn:	0.0113760	total:	24.2s	remaining:	1m 4s
271:	learn:	0.0112969	total:	24.2s	remaining:	1m 4s
272:	learn:	0.0111766	total:	24.3s	remaining:	1m 4s
273:	learn:	0.0110542	total:	24.4s	remaining:	1m 4s
274:	learn:	0.0109691	total:	24.5s	remaining:	1m 4s
275:	learn:	0.0108587	total:	24.6s	remaining:	1m 4s
276:	learn:	0.0107951	total:	24.7s	remaining:	1m 4s
277:	learn:	0.0107141	total:	24.8s	remaining:	1m 4s
278:	learn:	0.0106076	total:	24.9s	remaining:	1m 4s
279:	learn:	0.0105140	total:	24.9s	remaining:	1m 4s
280:	learn:	0.0104459	total:	25s	remaining:	1m 4s
281:	learn:	0.0103318	total:	25.1s	remaining:	1m 3s
282:	learn:	0.0102657	total:	25.2s	remaining:	1m 3s
283:	learn:	0.0101935	total:	25.3s	remaining:	1m 3s
284:	learn:	0.0100845	total:	25.4s	remaining:	1m 3s
285:	learn:	0.0100208	total:	25.5s	remaining:	1m 3s
286:	learn:	0.0099468	total:	25.6s	remaining:	1m 3s
287:	learn:	0.0099009	total:	25.6s	remaining:	1m 3s
288:	learn:	0.0098203	total:	25.7s	remaining:	1m 3s
289:	learn:	0.0097411	total:	25.8s	remaining:	1m 3s
290:	learn:	0.0096824	total:	25.9s	remaining:	1m 3s
291:	learn:	0.0096010	total:	26s	remaining:	1m 3s
292:	learn:	0.0095411	total:	26.1s	remaining:	1m 2s
293:	learn:	0.0094769	total:	26.2s	remaining:	1m 2s
294:	learn:	0.0093758	total:	26.3s	remaining:	1m 2s
295:	learn:	0.0092808	total:	26.3s	remaining:	1m 2s
296:	learn:	0.0091979	total:	26.4s	remaining:	1m 2s
297:	learn:	0.0091332	total:	26.5s	remaining:	1m 2s
298:	learn:	0.0090595	total:	26.6s	remaining:	1m 2s
299:	learn:	0.0089933	total:	26.7s	remaining:	1m 2s
300:	learn:	0.0089115	total:	26.8s	remaining:	1m 2s
301:	learn:	0.0088526	total:	26.9s	remaining:	1m 2s
302:	learn:	0.0087827	total:	27s	remaining:	1m 1s
303:	learn:	0.0087154	total:	27s	remaining:	1m 1s
304:	learn:	0.0086428	total:	27.1s	remaining:	1m 1s
305:	learn:	0.0085690	total:	27.2s	remaining:	1m 1s
306:	learn:	0.0085240	total:	27.3s	remaining:	1m 1s
307:	learn:	0.0084336	total:	27.4s	remaining:	1m 1s
308:	learn:	0.0083501	total:	27.5s	remaining:	1m 1s
309:	learn:	0.0083053	total:	27.6s	remaining:	1m 1s

310:	learn: 0.0082431	total: 27.6s	remaining: 1m 1s
311:	learn: 0.0081777	total: 27.7s	remaining: 1m 1s
312:	learn: 0.0081113	total: 27.8s	remaining: 1m 1s
313:	learn: 0.0080561	total: 27.9s	remaining: 1m
314:	learn: 0.0079937	total: 28s	remaining: 1m
315:	learn: 0.0079436	total: 28.1s	remaining: 1m
316:	learn: 0.0078687	total: 28.2s	remaining: 1m
317:	learn: 0.0078144	total: 28.2s	remaining: 1m
318:	learn: 0.0077437	total: 28.3s	remaining: 1m
319:	learn: 0.0076899	total: 28.4s	remaining: 1m
320:	learn: 0.0076220	total: 28.5s	remaining: 1m
321:	learn: 0.0075612	total: 28.6s	remaining: 1m
322:	learn: 0.0075126	total: 28.7s	remaining: 1m
323:	learn: 0.0074490	total: 28.8s	remaining: 1m
324:	learn: 0.0074017	total: 28.9s	remaining: 60s
325:	learn: 0.0073371	total: 29s	remaining: 59.9s
326:	learn: 0.0072862	total: 29s	remaining: 59.8s
327:	learn: 0.0072299	total: 29.1s	remaining: 59.7s
328:	learn: 0.0071769	total: 29.2s	remaining: 59.6s
329:	learn: 0.0071243	total: 29.3s	remaining: 59.5s
330:	learn: 0.0070713	total: 29.4s	remaining: 59.4s
331:	learn: 0.0070226	total: 29.5s	remaining: 59.3s
332:	learn: 0.0069722	total: 29.6s	remaining: 59.2s
333:	learn: 0.0069129	total: 29.6s	remaining: 59.1s
334:	learn: 0.0068581	total: 29.7s	remaining: 59s
335:	learn: 0.0067986	total: 29.8s	remaining: 58.9s
336:	learn: 0.0067520	total: 29.9s	remaining: 58.9s
337:	learn: 0.0066913	total: 30s	remaining: 58.8s
338:	learn: 0.0066428	total: 30.1s	remaining: 58.7s
339:	learn: 0.0065916	total: 30.2s	remaining: 58.6s
340:	learn: 0.0065471	total: 30.3s	remaining: 58.5s
341:	learn: 0.0065062	total: 30.4s	remaining: 58.4s
342:	learn: 0.0064732	total: 30.4s	remaining: 58.3s
343:	learn: 0.0064287	total: 30.5s	remaining: 58.2s
344:	learn: 0.0063619	total: 30.6s	remaining: 58.1s
345:	learn: 0.0063193	total: 30.7s	remaining: 58s
346:	learn: 0.0062584	total: 30.8s	remaining: 58s
347:	learn: 0.0062042	total: 30.9s	remaining: 57.9s
348:	learn: 0.0061684	total: 31s	remaining: 57.8s
349:	learn: 0.0061318	total: 31.1s	remaining: 57.7s
350:	learn: 0.0060846	total: 31.1s	remaining: 57.6s
351:	learn: 0.0060520	total: 31.2s	remaining: 57.5s
352:	learn: 0.0060110	total: 31.3s	remaining: 57.4s
353:	learn: 0.0059766	total: 31.4s	remaining: 57.3s
354:	learn: 0.0059458	total: 31.5s	remaining: 57.2s
355:	learn: 0.0059055	total: 31.6s	remaining: 57.1s
356:	learn: 0.0058551	total: 31.7s	remaining: 57.1s
357:	learn: 0.0058078	total: 31.8s	remaining: 57s
358:	learn: 0.0057692	total: 31.8s	remaining: 56.9s
359:	learn: 0.0057434	total: 31.9s	remaining: 56.8s
360:	learn: 0.0057007	total: 32s	remaining: 56.7s
361:	learn: 0.0056671	total: 32.1s	remaining: 56.6s
362:	learn: 0.0056257	total: 32.2s	remaining: 56.5s
363:	learn: 0.0056042	total: 32.3s	remaining: 56.4s
364:	learn: 0.0055734	total: 32.4s	remaining: 56.3s
365:	learn: 0.0055437	total: 32.5s	remaining: 56.2s
366:	learn: 0.0055182	total: 32.6s	remaining: 56.2s
367:	learn: 0.0054800	total: 32.6s	remaining: 56.1s
368:	learn: 0.0054400	total: 32.7s	remaining: 56s
369:	learn: 0.0054017	total: 32.8s	remaining: 55.9s
370:	learn: 0.0053661	total: 32.9s	remaining: 55.8s
371:	learn: 0.0053295	total: 33s	remaining: 55.7s
372:	learn: 0.0052902	total: 33.1s	remaining: 55.6s
373:	learn: 0.0052587	total: 33.2s	remaining: 55.5s
374:	learn: 0.0052325	total: 33.3s	remaining: 55.4s
375:	learn: 0.0051985	total: 33.3s	remaining: 55.3s
376:	learn: 0.0051731	total: 33.4s	remaining: 55.2s
377:	learn: 0.0051469	total: 33.5s	remaining: 55.2s
378:	learn: 0.0051166	total: 33.6s	remaining: 55.1s

379:	learn: 0.0050774	total: 33.7s	remaining: 55s
380:	learn: 0.0050475	total: 33.8s	remaining: 54.9s
381:	learn: 0.0050173	total: 33.9s	remaining: 54.8s
382:	learn: 0.0049820	total: 33.9s	remaining: 54.7s
383:	learn: 0.0049483	total: 34s	remaining: 54.6s
384:	learn: 0.0049219	total: 34.1s	remaining: 54.5s
385:	learn: 0.0048934	total: 34.2s	remaining: 54.4s
386:	learn: 0.0048664	total: 34.3s	remaining: 54.3s
387:	learn: 0.0048432	total: 34.4s	remaining: 54.2s
388:	learn: 0.0048110	total: 34.5s	remaining: 54.1s
389:	learn: 0.0047692	total: 34.6s	remaining: 54.1s
390:	learn: 0.0047415	total: 34.7s	remaining: 54s
391:	learn: 0.0047179	total: 34.7s	remaining: 53.9s
392:	learn: 0.0046861	total: 34.8s	remaining: 53.8s
393:	learn: 0.0046629	total: 34.9s	remaining: 53.7s
394:	learn: 0.0046398	total: 35s	remaining: 53.6s
395:	learn: 0.0046106	total: 35.1s	remaining: 53.5s
396:	learn: 0.0045838	total: 35.2s	remaining: 53.4s
397:	learn: 0.0045660	total: 35.3s	remaining: 53.3s
398:	learn: 0.0045284	total: 35.3s	remaining: 53.2s
399:	learn: 0.0044987	total: 35.4s	remaining: 53.1s
400:	learn: 0.0044810	total: 35.5s	remaining: 53.1s
401:	learn: 0.0044557	total: 35.6s	remaining: 53s
402:	learn: 0.0044321	total: 35.7s	remaining: 52.9s
403:	learn: 0.0044162	total: 35.8s	remaining: 52.8s
404:	learn: 0.0043871	total: 35.9s	remaining: 52.7s
405:	learn: 0.0043629	total: 36s	remaining: 52.6s
406:	learn: 0.0043379	total: 36.1s	remaining: 52.5s
407:	learn: 0.0043176	total: 36.1s	remaining: 52.4s
408:	learn: 0.0042963	total: 36.2s	remaining: 52.3s
409:	learn: 0.0042759	total: 36.3s	remaining: 52.3s
410:	learn: 0.0042455	total: 36.4s	remaining: 52.2s
411:	learn: 0.0042181	total: 36.5s	remaining: 52.1s
412:	learn: 0.0041915	total: 36.6s	remaining: 52s
413:	learn: 0.0041728	total: 36.7s	remaining: 51.9s
414:	learn: 0.0041498	total: 36.8s	remaining: 51.8s
415:	learn: 0.0041261	total: 36.9s	remaining: 51.7s
416:	learn: 0.0041006	total: 36.9s	remaining: 51.7s
417:	learn: 0.0040818	total: 37s	remaining: 51.6s
418:	learn: 0.0040665	total: 37.1s	remaining: 51.5s
419:	learn: 0.0040463	total: 37.2s	remaining: 51.4s
420:	learn: 0.0040237	total: 37.3s	remaining: 51.3s
421:	learn: 0.0040004	total: 37.4s	remaining: 51.2s
422:	learn: 0.0039827	total: 37.5s	remaining: 51.1s
423:	learn: 0.0039633	total: 37.6s	remaining: 51s
424:	learn: 0.0039490	total: 37.7s	remaining: 50.9s
425:	learn: 0.0039277	total: 37.7s	remaining: 50.9s
426:	learn: 0.0039049	total: 37.8s	remaining: 50.8s
427:	learn: 0.0038900	total: 37.9s	remaining: 50.7s
428:	learn: 0.0038679	total: 38s	remaining: 50.6s
429:	learn: 0.0038436	total: 38.1s	remaining: 50.5s
430:	learn: 0.0038197	total: 38.2s	remaining: 50.4s
431:	learn: 0.0038026	total: 38.3s	remaining: 50.3s
432:	learn: 0.0037834	total: 38.4s	remaining: 50.2s
433:	learn: 0.0037589	total: 38.5s	remaining: 50.1s
434:	learn: 0.0037412	total: 38.5s	remaining: 50.1s
435:	learn: 0.0037260	total: 38.6s	remaining: 50s
436:	learn: 0.0037066	total: 38.7s	remaining: 49.9s
437:	learn: 0.0036928	total: 38.8s	remaining: 49.8s
438:	learn: 0.0036778	total: 38.9s	remaining: 49.7s
439:	learn: 0.0036778	total: 39s	remaining: 49.6s
440:	learn: 0.0036640	total: 39.1s	remaining: 49.5s
441:	learn: 0.0036467	total: 39.1s	remaining: 49.4s
442:	learn: 0.0036291	total: 39.2s	remaining: 49.3s
443:	learn: 0.0036173	total: 39.3s	remaining: 49.2s
444:	learn: 0.0035991	total: 39.4s	remaining: 49.1s
445:	learn: 0.0035783	total: 39.5s	remaining: 49.1s
446:	learn: 0.0035645	total: 39.6s	remaining: 49s
447:	learn: 0.0035509	total: 39.7s	remaining: 48.9s



448:	learn:	0.0035315	total:	39.8s	remaining:	48.8s
449:	learn:	0.0035133	total:	39.8s	remaining:	48.7s
450:	learn:	0.0034970	total:	39.9s	remaining:	48.6s
451:	learn:	0.0034756	total:	40s	remaining:	48.5s
452:	learn:	0.0034576	total:	40.1s	remaining:	48.4s
453:	learn:	0.0034400	total:	40.2s	remaining:	48.3s
454:	learn:	0.0034241	total:	40.3s	remaining:	48.2s
455:	learn:	0.0034087	total:	40.4s	remaining:	48.1s
456:	learn:	0.0033975	total:	40.4s	remaining:	48s
457:	learn:	0.0033828	total:	40.5s	remaining:	48s
458:	learn:	0.0033644	total:	40.6s	remaining:	47.9s
459:	learn:	0.0033546	total:	40.7s	remaining:	47.8s
460:	learn:	0.0033393	total:	40.8s	remaining:	47.7s
461:	learn:	0.0033210	total:	40.9s	remaining:	47.6s
462:	learn:	0.0033079	total:	41s	remaining:	47.5s
463:	learn:	0.0032926	total:	41.1s	remaining:	47.5s
464:	learn:	0.0032823	total:	41.2s	remaining:	47.4s
465:	learn:	0.0032687	total:	41.3s	remaining:	47.3s
466:	learn:	0.0032557	total:	41.4s	remaining:	47.2s
467:	learn:	0.0032557	total:	41.4s	remaining:	47.1s
468:	learn:	0.0032440	total:	41.5s	remaining:	47s
469:	learn:	0.0032246	total:	41.6s	remaining:	46.9s
470:	learn:	0.0032138	total:	41.7s	remaining:	46.9s
471:	learn:	0.0032007	total:	41.8s	remaining:	46.8s
472:	learn:	0.0031835	total:	41.9s	remaining:	46.7s
473:	learn:	0.0031671	total:	42s	remaining:	46.6s
474:	learn:	0.0031462	total:	42.1s	remaining:	46.5s
475:	learn:	0.0031238	total:	42.2s	remaining:	46.4s
476:	learn:	0.0031115	total:	42.3s	remaining:	46.4s
477:	learn:	0.0030972	total:	42.4s	remaining:	46.3s
478:	learn:	0.0030836	total:	42.5s	remaining:	46.2s
479:	learn:	0.0030724	total:	42.6s	remaining:	46.1s
480:	learn:	0.0030546	total:	42.7s	remaining:	46s
481:	learn:	0.0030416	total:	42.8s	remaining:	46s
482:	learn:	0.0030318	total:	42.8s	remaining:	45.9s
483:	learn:	0.0030223	total:	42.9s	remaining:	45.8s
484:	learn:	0.0030094	total:	43s	remaining:	45.7s
485:	learn:	0.0029972	total:	43.1s	remaining:	45.6s
486:	learn:	0.0029972	total:	43.2s	remaining:	45.5s
487:	learn:	0.0029835	total:	43.3s	remaining:	45.4s
488:	learn:	0.0029705	total:	43.4s	remaining:	45.3s
489:	learn:	0.0029558	total:	43.5s	remaining:	45.2s
490:	learn:	0.0029464	total:	43.5s	remaining:	45.1s
491:	learn:	0.0029338	total:	43.6s	remaining:	45s
492:	learn:	0.0029182	total:	43.7s	remaining:	45s
493:	learn:	0.0029056	total:	43.8s	remaining:	44.9s
494:	learn:	0.0028912	total:	43.9s	remaining:	44.8s
495:	learn:	0.0028813	total:	44s	remaining:	44.7s
496:	learn:	0.0028691	total:	44.1s	remaining:	44.6s
497:	learn:	0.0028691	total:	44.2s	remaining:	44.5s
498:	learn:	0.0028691	total:	44.3s	remaining:	44.4s
499:	learn:	0.0028566	total:	44.3s	remaining:	44.3s
500:	learn:	0.0028419	total:	44.4s	remaining:	44.2s
501:	learn:	0.0028320	total:	44.5s	remaining:	44.2s
502:	learn:	0.0028150	total:	44.6s	remaining:	44.1s
503:	learn:	0.0028035	total:	44.7s	remaining:	44s
504:	learn:	0.0028035	total:	44.8s	remaining:	43.9s
505:	learn:	0.0027896	total:	44.9s	remaining:	43.8s
506:	learn:	0.0027718	total:	44.9s	remaining:	43.7s
507:	learn:	0.0027596	total:	45s	remaining:	43.6s
508:	learn:	0.0027489	total:	45.1s	remaining:	43.5s
509:	learn:	0.0027367	total:	45.2s	remaining:	43.4s
510:	learn:	0.0027231	total:	45.3s	remaining:	43.3s
511:	learn:	0.0027195	total:	45.4s	remaining:	43.3s
512:	learn:	0.0027070	total:	45.5s	remaining:	43.2s
513:	learn:	0.0026957	total:	45.6s	remaining:	43.1s
514:	learn:	0.0026957	total:	45.6s	remaining:	43s
515:	learn:	0.0026865	total:	45.7s	remaining:	42.9s
516:	learn:	0.0026762	total:	45.8s	remaining:	42.8s



517:	learn:	0.0026762	total:	45.9s	remaining:	42.7s
518:	learn:	0.0026606	total:	46s	remaining:	42.6s
519:	learn:	0.0026502	total:	46.1s	remaining:	42.5s
520:	learn:	0.0026383	total:	46.2s	remaining:	42.4s
521:	learn:	0.0026267	total:	46.2s	remaining:	42.3s
522:	learn:	0.0026267	total:	46.3s	remaining:	42.3s
523:	learn:	0.0026267	total:	46.4s	remaining:	42.2s
524:	learn:	0.0026171	total:	46.5s	remaining:	42.1s
525:	learn:	0.0026059	total:	46.6s	remaining:	42s
526:	learn:	0.0026059	total:	46.7s	remaining:	41.9s
527:	learn:	0.0025987	total:	46.8s	remaining:	41.8s
528:	learn:	0.0025987	total:	46.8s	remaining:	41.7s
529:	learn:	0.0025987	total:	46.9s	remaining:	41.6s
530:	learn:	0.0025987	total:	47s	remaining:	41.5s
531:	learn:	0.0025987	total:	47.1s	remaining:	41.4s
532:	learn:	0.0025986	total:	47.2s	remaining:	41.3s
533:	learn:	0.0025874	total:	47.3s	remaining:	41.3s
534:	learn:	0.0025785	total:	47.4s	remaining:	41.2s
535:	learn:	0.0025694	total:	47.4s	remaining:	41.1s
536:	learn:	0.0025602	total:	47.5s	remaining:	41s
537:	learn:	0.0025462	total:	47.6s	remaining:	40.9s
538:	learn:	0.0025461	total:	47.7s	remaining:	40.8s
539:	learn:	0.0025357	total:	47.8s	remaining:	40.7s
540:	learn:	0.0025357	total:	47.9s	remaining:	40.6s
541:	learn:	0.0025287	total:	48s	remaining:	40.5s
542:	learn:	0.0025199	total:	48s	remaining:	40.4s
543:	learn:	0.0025125	total:	48.1s	remaining:	40.3s
544:	learn:	0.0025125	total:	48.2s	remaining:	40.3s
545:	learn:	0.0025022	total:	48.3s	remaining:	40.2s
546:	learn:	0.0025022	total:	48.4s	remaining:	40.1s
547:	learn:	0.0024928	total:	48.5s	remaining:	40s
548:	learn:	0.0024848	total:	48.6s	remaining:	39.9s
549:	learn:	0.0024776	total:	48.7s	remaining:	39.8s
550:	learn:	0.0024683	total:	48.7s	remaining:	39.7s
551:	learn:	0.0024591	total:	48.8s	remaining:	39.6s
552:	learn:	0.0024591	total:	48.9s	remaining:	39.5s
553:	learn:	0.0024480	total:	49s	remaining:	39.4s
554:	learn:	0.0024449	total:	49.1s	remaining:	39.4s
555:	learn:	0.0024449	total:	49.2s	remaining:	39.3s
556:	learn:	0.0024359	total:	49.3s	remaining:	39.2s
557:	learn:	0.0024267	total:	49.4s	remaining:	39.1s
558:	learn:	0.0024267	total:	49.4s	remaining:	39s
559:	learn:	0.0024166	total:	49.5s	remaining:	38.9s
560:	learn:	0.0024092	total:	49.6s	remaining:	38.8s
561:	learn:	0.0023987	total:	49.7s	remaining:	38.7s
562:	learn:	0.0023987	total:	49.8s	remaining:	38.6s
563:	learn:	0.0023987	total:	49.9s	remaining:	38.5s
564:	learn:	0.0023986	total:	50s	remaining:	38.5s
565:	learn:	0.0023986	total:	50s	remaining:	38.4s
566:	learn:	0.0023894	total:	50.1s	remaining:	38.3s
567:	learn:	0.0023894	total:	50.2s	remaining:	38.2s
568:	learn:	0.0023893	total:	50.3s	remaining:	38.1s
569:	learn:	0.0023893	total:	50.4s	remaining:	38s
570:	learn:	0.0023818	total:	50.5s	remaining:	37.9s
571:	learn:	0.0023739	total:	50.6s	remaining:	37.8s
572:	learn:	0.0023739	total:	50.7s	remaining:	37.7s
573:	learn:	0.0023738	total:	50.7s	remaining:	37.7s
574:	learn:	0.0023649	total:	50.8s	remaining:	37.6s
575:	learn:	0.0023571	total:	50.9s	remaining:	37.5s
576:	learn:	0.0023470	total:	51s	remaining:	37.4s
577:	learn:	0.0023470	total:	51.1s	remaining:	37.3s
578:	learn:	0.0023470	total:	51.2s	remaining:	37.2s
579:	learn:	0.0023470	total:	51.3s	remaining:	37.1s
580:	learn:	0.0023470	total:	51.4s	remaining:	37s
581:	learn:	0.0023366	total:	51.4s	remaining:	36.9s
582:	learn:	0.0023366	total:	51.5s	remaining:	36.9s
583:	learn:	0.0023265	total:	51.6s	remaining:	36.8s
584:	learn:	0.0023265	total:	51.7s	remaining:	36.7s
585:	learn:	0.0023265	total:	51.8s	remaining:	36.6s

586:	learn:	0.0023265	total:	51.9s	remaining:	36.5s
587:	learn:	0.0023171	total:	52s	remaining:	36.4s
588:	learn:	0.0023084	total:	52s	remaining:	36.3s
589:	learn:	0.0023084	total:	52.1s	remaining:	36.2s
590:	learn:	0.0023002	total:	52.2s	remaining:	36.1s
591:	learn:	0.0023001	total:	52.3s	remaining:	36s
592:	learn:	0.0023001	total:	52.4s	remaining:	36s
593:	learn:	0.0023001	total:	52.5s	remaining:	35.9s
594:	learn:	0.0023001	total:	52.6s	remaining:	35.8s
595:	learn:	0.0022913	total:	52.7s	remaining:	35.7s
596:	learn:	0.0022913	total:	52.7s	remaining:	35.6s
597:	learn:	0.0022840	total:	52.8s	remaining:	35.5s
598:	learn:	0.0022767	total:	52.9s	remaining:	35.4s
599:	learn:	0.0022682	total:	53s	remaining:	35.3s
600:	learn:	0.0022682	total:	53.1s	remaining:	35.2s
601:	learn:	0.0022569	total:	53.2s	remaining:	35.2s
602:	learn:	0.0022569	total:	53.3s	remaining:	35.1s
603:	learn:	0.0022478	total:	53.3s	remaining:	35s
604:	learn:	0.0022435	total:	53.4s	remaining:	34.9s
605:	learn:	0.0022346	total:	53.5s	remaining:	34.8s
606:	learn:	0.0022267	total:	53.6s	remaining:	34.7s
607:	learn:	0.0022267	total:	53.7s	remaining:	34.6s
608:	learn:	0.0022267	total:	53.8s	remaining:	34.5s
609:	learn:	0.0022266	total:	53.9s	remaining:	34.4s
610:	learn:	0.0022266	total:	53.9s	remaining:	34.3s
611:	learn:	0.0022199	total:	54s	remaining:	34.3s
612:	learn:	0.0022199	total:	54.1s	remaining:	34.2s
613:	learn:	0.0022199	total:	54.2s	remaining:	34.1s
614:	learn:	0.0022132	total:	54.3s	remaining:	34s
615:	learn:	0.0022089	total:	54.4s	remaining:	33.9s
616:	learn:	0.0022023	total:	54.5s	remaining:	33.8s
617:	learn:	0.0021969	total:	54.6s	remaining:	33.7s
618:	learn:	0.0021969	total:	54.6s	remaining:	33.6s
619:	learn:	0.0021969	total:	54.7s	remaining:	33.5s
620:	learn:	0.0021969	total:	54.8s	remaining:	33.4s
621:	learn:	0.0021969	total:	54.9s	remaining:	33.4s
622:	learn:	0.0021968	total:	55s	remaining:	33.3s
623:	learn:	0.0021968	total:	55.1s	remaining:	33.2s
624:	learn:	0.0021968	total:	55.2s	remaining:	33.1s
625:	learn:	0.0021864	total:	55.2s	remaining:	33s
626:	learn:	0.0021863	total:	55.3s	remaining:	32.9s
627:	learn:	0.0021798	total:	55.4s	remaining:	32.8s
628:	learn:	0.0021798	total:	55.5s	remaining:	32.7s
629:	learn:	0.0021798	total:	55.6s	remaining:	32.6s
630:	learn:	0.0021797	total:	55.7s	remaining:	32.6s
631:	learn:	0.0021715	total:	55.8s	remaining:	32.5s
632:	learn:	0.0021647	total:	55.9s	remaining:	32.4s
633:	learn:	0.0021647	total:	55.9s	remaining:	32.3s
634:	learn:	0.0021553	total:	56s	remaining:	32.2s
635:	learn:	0.0021494	total:	56.1s	remaining:	32.1s
636:	learn:	0.0021494	total:	56.2s	remaining:	32s
637:	learn:	0.0021494	total:	56.3s	remaining:	31.9s
638:	learn:	0.0021494	total:	56.4s	remaining:	31.8s
639:	learn:	0.0021494	total:	56.5s	remaining:	31.8s
640:	learn:	0.0021493	total:	56.5s	remaining:	31.7s
641:	learn:	0.0021404	total:	56.6s	remaining:	31.6s
642:	learn:	0.0021320	total:	56.7s	remaining:	31.5s
643:	learn:	0.0021320	total:	56.8s	remaining:	31.4s
644:	learn:	0.0021320	total:	56.9s	remaining:	31.3s
645:	learn:	0.0021320	total:	57s	remaining:	31.2s
646:	learn:	0.0021320	total:	57.1s	remaining:	31.1s
647:	learn:	0.0021319	total:	57.2s	remaining:	31s
648:	learn:	0.0021319	total:	57.2s	remaining:	31s
649:	learn:	0.0021235	total:	57.3s	remaining:	30.9s
650:	learn:	0.0021182	total:	57.4s	remaining:	30.8s
651:	learn:	0.0021182	total:	57.5s	remaining:	30.7s
652:	learn:	0.0021182	total:	57.6s	remaining:	30.6s
653:	learn:	0.0021182	total:	57.7s	remaining:	30.5s
654:	learn:	0.0021182	total:	57.8s	remaining:	30.4s

655:	learn:	0.0021182	total:	57.9s	remaining:	30.3s
656:	learn:	0.0021182	total:	57.9s	remaining:	30.2s
657:	learn:	0.0021182	total:	58s	remaining:	30.2s
658:	learn:	0.0021182	total:	58.1s	remaining:	30.1s
659:	learn:	0.0021124	total:	58.2s	remaining:	30s
660:	learn:	0.0021063	total:	58.3s	remaining:	29.9s
661:	learn:	0.0021063	total:	58.4s	remaining:	29.8s
662:	learn:	0.0021062	total:	58.4s	remaining:	29.7s
663:	learn:	0.0021062	total:	58.5s	remaining:	29.6s
664:	learn:	0.0021036	total:	58.6s	remaining:	29.5s
665:	learn:	0.0021036	total:	58.7s	remaining:	29.4s
666:	learn:	0.0020938	total:	58.8s	remaining:	29.3s
667:	learn:	0.0020938	total:	58.9s	remaining:	29.3s
668:	learn:	0.0020938	total:	59s	remaining:	29.2s
669:	learn:	0.0020852	total:	59s	remaining:	29.1s
670:	learn:	0.0020852	total:	59.1s	remaining:	29s
671:	learn:	0.0020782	total:	59.2s	remaining:	28.9s
672:	learn:	0.0020782	total:	59.3s	remaining:	28.8s
673:	learn:	0.0020782	total:	59.4s	remaining:	28.7s
674:	learn:	0.0020782	total:	59.5s	remaining:	28.6s
675:	learn:	0.0020750	total:	59.6s	remaining:	28.5s
676:	learn:	0.0020750	total:	59.6s	remaining:	28.5s
677:	learn:	0.0020750	total:	59.7s	remaining:	28.4s
678:	learn:	0.0020750	total:	59.8s	remaining:	28.3s
679:	learn:	0.0020750	total:	59.9s	remaining:	28.2s
680:	learn:	0.0020750	total:	60s	remaining:	28.1s
681:	learn:	0.0020750	total:	1m	remaining:	28s
682:	learn:	0.0020750	total:	1m	remaining:	27.9s
683:	learn:	0.0020750	total:	1m	remaining:	27.8s
684:	learn:	0.0020750	total:	1m	remaining:	27.7s
685:	learn:	0.0020750	total:	1m	remaining:	27.7s
686:	learn:	0.0020694	total:	1m	remaining:	27.6s
687:	learn:	0.0020694	total:	1m	remaining:	27.5s
688:	learn:	0.0020694	total:	1m	remaining:	27.4s
689:	learn:	0.0020693	total:	1m	remaining:	27.3s
690:	learn:	0.0020693	total:	1m	remaining:	27.2s
691:	learn:	0.0020693	total:	1m	remaining:	27.1s
692:	learn:	0.0020693	total:	1m 1s	remaining:	27s
693:	learn:	0.0020693	total:	1m 1s	remaining:	26.9s
694:	learn:	0.0020693	total:	1m 1s	remaining:	26.9s
695:	learn:	0.0020693	total:	1m 1s	remaining:	26.8s
696:	learn:	0.0020692	total:	1m 1s	remaining:	26.7s
697:	learn:	0.0020692	total:	1m 1s	remaining:	26.6s
698:	learn:	0.0020692	total:	1m 1s	remaining:	26.5s
699:	learn:	0.0020692	total:	1m 1s	remaining:	26.4s
700:	learn:	0.0020685	total:	1m 1s	remaining:	26.3s
701:	learn:	0.0020685	total:	1m 1s	remaining:	26.2s
702:	learn:	0.0020684	total:	1m 1s	remaining:	26.2s
703:	learn:	0.0020684	total:	1m 1s	remaining:	26.1s
704:	learn:	0.0020684	total:	1m 2s	remaining:	26s
705:	learn:	0.0020619	total:	1m 2s	remaining:	25.9s
706:	learn:	0.0020618	total:	1m 2s	remaining:	25.8s
707:	learn:	0.0020618	total:	1m 2s	remaining:	25.7s
708:	learn:	0.0020618	total:	1m 2s	remaining:	25.6s
709:	learn:	0.0020618	total:	1m 2s	remaining:	25.5s
710:	learn:	0.0020618	total:	1m 2s	remaining:	25.4s
711:	learn:	0.0020618	total:	1m 2s	remaining:	25.4s
712:	learn:	0.0020618	total:	1m 2s	remaining:	25.3s
713:	learn:	0.0020618	total:	1m 2s	remaining:	25.2s
714:	learn:	0.0020618	total:	1m 2s	remaining:	25.1s
715:	learn:	0.0020618	total:	1m 3s	remaining:	25s
716:	learn:	0.0020618	total:	1m 3s	remaining:	24.9s
717:	learn:	0.0020618	total:	1m 3s	remaining:	24.8s
718:	learn:	0.0020618	total:	1m 3s	remaining:	24.7s
719:	learn:	0.0020550	total:	1m 3s	remaining:	24.6s
720:	learn:	0.0020550	total:	1m 3s	remaining:	24.6s
721:	learn:	0.0020550	total:	1m 3s	remaining:	24.5s
722:	learn:	0.0020549	total:	1m 3s	remaining:	24.4s
723:	learn:	0.0020549	total:	1m 3s	remaining:	24.3s

724:	learn: 0.0020549	total: 1m 3s	remaining: 24.2s
725:	learn: 0.0020455	total: 1m 3s	remaining: 24.1s
726:	learn: 0.0020455	total: 1m 3s	remaining: 24s
727:	learn: 0.0020455	total: 1m 4s	remaining: 23.9s
728:	learn: 0.0020455	total: 1m 4s	remaining: 23.8s
729:	learn: 0.0020382	total: 1m 4s	remaining: 23.8s
730:	learn: 0.0020382	total: 1m 4s	remaining: 23.7s
731:	learn: 0.0020382	total: 1m 4s	remaining: 23.6s
732:	learn: 0.0020382	total: 1m 4s	remaining: 23.5s
733:	learn: 0.0020382	total: 1m 4s	remaining: 23.4s
734:	learn: 0.0020382	total: 1m 4s	remaining: 23.3s
735:	learn: 0.0020382	total: 1m 4s	remaining: 23.2s
736:	learn: 0.0020382	total: 1m 4s	remaining: 23.1s
737:	learn: 0.0020382	total: 1m 4s	remaining: 23.1s
738:	learn: 0.0020382	total: 1m 5s	remaining: 23s
739:	learn: 0.0020382	total: 1m 5s	remaining: 22.9s
740:	learn: 0.0020382	total: 1m 5s	remaining: 22.8s
741:	learn: 0.0020381	total: 1m 5s	remaining: 22.7s
742:	learn: 0.0020381	total: 1m 5s	remaining: 22.6s
743:	learn: 0.0020381	total: 1m 5s	remaining: 22.5s
744:	learn: 0.0020381	total: 1m 5s	remaining: 22.4s
745:	learn: 0.0020314	total: 1m 5s	remaining: 22.3s
746:	learn: 0.0020314	total: 1m 5s	remaining: 22.3s
747:	learn: 0.0020314	total: 1m 5s	remaining: 22.2s
748:	learn: 0.0020314	total: 1m 5s	remaining: 22.1s
749:	learn: 0.0020314	total: 1m 5s	remaining: 22s
750:	learn: 0.0020314	total: 1m 6s	remaining: 21.9s
751:	learn: 0.0020314	total: 1m 6s	remaining: 21.8s
752:	learn: 0.0020314	total: 1m 6s	remaining: 21.7s
753:	learn: 0.0020313	total: 1m 6s	remaining: 21.6s
754:	learn: 0.0020313	total: 1m 6s	remaining: 21.5s
755:	learn: 0.0020313	total: 1m 6s	remaining: 21.5s
756:	learn: 0.0020313	total: 1m 6s	remaining: 21.4s
757:	learn: 0.0020234	total: 1m 6s	remaining: 21.3s
758:	learn: 0.0020168	total: 1m 6s	remaining: 21.2s
759:	learn: 0.0020168	total: 1m 6s	remaining: 21.1s
760:	learn: 0.0020168	total: 1m 6s	remaining: 21s
761:	learn: 0.0020086	total: 1m 7s	remaining: 20.9s
762:	learn: 0.0020086	total: 1m 7s	remaining: 20.8s
763:	learn: 0.0020085	total: 1m 7s	remaining: 20.7s
764:	learn: 0.0020085	total: 1m 7s	remaining: 20.7s
765:	learn: 0.0020085	total: 1m 7s	remaining: 20.6s
766:	learn: 0.0020085	total: 1m 7s	remaining: 20.5s
767:	learn: 0.0020085	total: 1m 7s	remaining: 20.4s
768:	learn: 0.0020033	total: 1m 7s	remaining: 20.3s
769:	learn: 0.0020033	total: 1m 7s	remaining: 20.2s
770:	learn: 0.0020033	total: 1m 7s	remaining: 20.1s
771:	learn: 0.0020033	total: 1m 7s	remaining: 20s
772:	learn: 0.0020033	total: 1m 7s	remaining: 20s
773:	learn: 0.0020033	total: 1m 8s	remaining: 19.9s
774:	learn: 0.0020033	total: 1m 8s	remaining: 19.8s
775:	learn: 0.0020032	total: 1m 8s	remaining: 19.7s
776:	learn: 0.0020032	total: 1m 8s	remaining: 19.6s
777:	learn: 0.0020032	total: 1m 8s	remaining: 19.5s
778:	learn: 0.0020032	total: 1m 8s	remaining: 19.4s
779:	learn: 0.0019962	total: 1m 8s	remaining: 19.3s
780:	learn: 0.0019962	total: 1m 8s	remaining: 19.2s
781:	learn: 0.0019962	total: 1m 8s	remaining: 19.2s
782:	learn: 0.0019962	total: 1m 8s	remaining: 19.1s
783:	learn: 0.0019962	total: 1m 8s	remaining: 19s
784:	learn: 0.0019962	total: 1m 8s	remaining: 18.9s
785:	learn: 0.0019962	total: 1m 9s	remaining: 18.8s
786:	learn: 0.0019962	total: 1m 9s	remaining: 18.7s
787:	learn: 0.0019962	total: 1m 9s	remaining: 18.6s
788:	learn: 0.0019962	total: 1m 9s	remaining: 18.5s
789:	learn: 0.0019886	total: 1m 9s	remaining: 18.5s
790:	learn: 0.0019886	total: 1m 9s	remaining: 18.4s
791:	learn: 0.0019886	total: 1m 9s	remaining: 18.3s
792:	learn: 0.0019820	total: 1m 9s	remaining: 18.2s



793:	learn: 0.0019820	total: 1m 9s	remaining: 18.1s
794:	learn: 0.0019819	total: 1m 9s	remaining: 18s
795:	learn: 0.0019819	total: 1m 9s	remaining: 17.9s
796:	learn: 0.0019751	total: 1m 10s	remaining: 17.8s
797:	learn: 0.0019751	total: 1m 10s	remaining: 17.8s
798:	learn: 0.0019751	total: 1m 10s	remaining: 17.7s
799:	learn: 0.0019750	total: 1m 10s	remaining: 17.6s
800:	learn: 0.0019750	total: 1m 10s	remaining: 17.5s
801:	learn: 0.0019678	total: 1m 10s	remaining: 17.4s
802:	learn: 0.0019678	total: 1m 10s	remaining: 17.3s
803:	learn: 0.0019639	total: 1m 10s	remaining: 17.2s
804:	learn: 0.0019639	total: 1m 10s	remaining: 17.1s
805:	learn: 0.0019639	total: 1m 10s	remaining: 17s
806:	learn: 0.0019639	total: 1m 10s	remaining: 17s
807:	learn: 0.0019639	total: 1m 10s	remaining: 16.9s
808:	learn: 0.0019638	total: 1m 11s	remaining: 16.8s
809:	learn: 0.0019638	total: 1m 11s	remaining: 16.7s
810:	learn: 0.0019638	total: 1m 11s	remaining: 16.6s
811:	learn: 0.0019638	total: 1m 11s	remaining: 16.5s
812:	learn: 0.0019568	total: 1m 11s	remaining: 16.4s
813:	learn: 0.0019497	total: 1m 11s	remaining: 16.3s
814:	learn: 0.0019497	total: 1m 11s	remaining: 16.3s
815:	learn: 0.0019423	total: 1m 11s	remaining: 16.2s
816:	learn: 0.0019423	total: 1m 11s	remaining: 16.1s
817:	learn: 0.0019423	total: 1m 11s	remaining: 16s
818:	learn: 0.0019423	total: 1m 11s	remaining: 15.9s
819:	learn: 0.0019423	total: 1m 12s	remaining: 15.8s
820:	learn: 0.0019423	total: 1m 12s	remaining: 15.7s
821:	learn: 0.0019392	total: 1m 12s	remaining: 15.6s
822:	learn: 0.0019392	total: 1m 12s	remaining: 15.6s
823:	learn: 0.0019329	total: 1m 12s	remaining: 15.5s
824:	learn: 0.0019329	total: 1m 12s	remaining: 15.4s
825:	learn: 0.0019329	total: 1m 12s	remaining: 15.3s
826:	learn: 0.0019329	total: 1m 12s	remaining: 15.2s
827:	learn: 0.0019328	total: 1m 12s	remaining: 15.1s
828:	learn: 0.0019328	total: 1m 12s	remaining: 15s
829:	learn: 0.0019328	total: 1m 12s	remaining: 14.9s
830:	learn: 0.0019328	total: 1m 13s	remaining: 14.9s
831:	learn: 0.0019328	total: 1m 13s	remaining: 14.8s
832:	learn: 0.0019328	total: 1m 13s	remaining: 14.7s
833:	learn: 0.0019328	total: 1m 13s	remaining: 14.6s
834:	learn: 0.0019277	total: 1m 13s	remaining: 14.5s
835:	learn: 0.0019277	total: 1m 13s	remaining: 14.4s
836:	learn: 0.0019277	total: 1m 13s	remaining: 14.3s
837:	learn: 0.0019277	total: 1m 13s	remaining: 14.2s
838:	learn: 0.0019277	total: 1m 13s	remaining: 14.1s
839:	learn: 0.0019276	total: 1m 13s	remaining: 14.1s
840:	learn: 0.0019276	total: 1m 13s	remaining: 14s
841:	learn: 0.0019275	total: 1m 14s	remaining: 13.9s
842:	learn: 0.0019275	total: 1m 14s	remaining: 13.8s
843:	learn: 0.0019275	total: 1m 14s	remaining: 13.7s
844:	learn: 0.0019275	total: 1m 14s	remaining: 13.6s
845:	learn: 0.0019275	total: 1m 14s	remaining: 13.5s
846:	learn: 0.0019275	total: 1m 14s	remaining: 13.4s
847:	learn: 0.0019275	total: 1m 14s	remaining: 13.4s
848:	learn: 0.0019275	total: 1m 14s	remaining: 13.3s
849:	learn: 0.0019260	total: 1m 14s	remaining: 13.2s
850:	learn: 0.0019259	total: 1m 14s	remaining: 13.1s
851:	learn: 0.0019250	total: 1m 14s	remaining: 13s
852:	learn: 0.0019250	total: 1m 14s	remaining: 12.9s
853:	learn: 0.0019250	total: 1m 15s	remaining: 12.8s
854:	learn: 0.0019185	total: 1m 15s	remaining: 12.7s
855:	learn: 0.0019185	total: 1m 15s	remaining: 12.7s
856:	learn: 0.0019185	total: 1m 15s	remaining: 12.6s
857:	learn: 0.0019185	total: 1m 15s	remaining: 12.5s
858:	learn: 0.0019185	total: 1m 15s	remaining: 12.4s
859:	learn: 0.0019139	total: 1m 15s	remaining: 12.3s
860:	learn: 0.0019139	total: 1m 15s	remaining: 12.2s
861:	learn: 0.0019139	total: 1m 15s	remaining: 12.1s



862:	learn:	0.0019085	total:	1m 15s	remaining:	12s
863:	learn:	0.0019085	total:	1m 15s	remaining:	12s
864:	learn:	0.0019085	total:	1m 16s	remaining:	11.9s
865:	learn:	0.0019085	total:	1m 16s	remaining:	11.8s
866:	learn:	0.0019085	total:	1m 16s	remaining:	11.7s
867:	learn:	0.0019085	total:	1m 16s	remaining:	11.6s
868:	learn:	0.0019085	total:	1m 16s	remaining:	11.5s
869:	learn:	0.0019085	total:	1m 16s	remaining:	11.4s
870:	learn:	0.0019085	total:	1m 16s	remaining:	11.3s
871:	learn:	0.0019084	total:	1m 16s	remaining:	11.2s
872:	learn:	0.0019034	total:	1m 16s	remaining:	11.2s
873:	learn:	0.0019034	total:	1m 16s	remaining:	11.1s
874:	learn:	0.0019034	total:	1m 16s	remaining:	11s
875:	learn:	0.0019034	total:	1m 16s	remaining:	10.9s
876:	learn:	0.0019034	total:	1m 17s	remaining:	10.8s
877:	learn:	0.0018966	total:	1m 17s	remaining:	10.7s
878:	learn:	0.0018966	total:	1m 17s	remaining:	10.6s
879:	learn:	0.0018966	total:	1m 17s	remaining:	10.5s
880:	learn:	0.0018966	total:	1m 17s	remaining:	10.5s
881:	learn:	0.0018966	total:	1m 17s	remaining:	10.4s
882:	learn:	0.0018937	total:	1m 17s	remaining:	10.3s
883:	learn:	0.0018937	total:	1m 17s	remaining:	10.2s
884:	learn:	0.0018872	total:	1m 17s	remaining:	10.1s
885:	learn:	0.0018872	total:	1m 17s	remaining:	10s
886:	learn:	0.0018872	total:	1m 17s	remaining:	9.93s
887:	learn:	0.0018872	total:	1m 18s	remaining:	9.84s
888:	learn:	0.0018872	total:	1m 18s	remaining:	9.75s
889:	learn:	0.0018871	total:	1m 18s	remaining:	9.66s
890:	learn:	0.0018871	total:	1m 18s	remaining:	9.58s
891:	learn:	0.0018871	total:	1m 18s	remaining:	9.49s
892:	learn:	0.0018871	total:	1m 18s	remaining:	9.4s
893:	learn:	0.0018871	total:	1m 18s	remaining:	9.31s
894:	learn:	0.0018871	total:	1m 18s	remaining:	9.23s
895:	learn:	0.0018871	total:	1m 18s	remaining:	9.14s
896:	learn:	0.0018871	total:	1m 18s	remaining:	9.05s
897:	learn:	0.0018871	total:	1m 18s	remaining:	8.96s
898:	learn:	0.0018871	total:	1m 19s	remaining:	8.88s
899:	learn:	0.0018871	total:	1m 19s	remaining:	8.79s
900:	learn:	0.0018780	total:	1m 19s	remaining:	8.7s
901:	learn:	0.0018780	total:	1m 19s	remaining:	8.61s
902:	learn:	0.0018780	total:	1m 19s	remaining:	8.53s
903:	learn:	0.0018719	total:	1m 19s	remaining:	8.44s
904:	learn:	0.0018719	total:	1m 19s	remaining:	8.35s
905:	learn:	0.0018718	total:	1m 19s	remaining:	8.26s
906:	learn:	0.0018718	total:	1m 19s	remaining:	8.18s
907:	learn:	0.0018718	total:	1m 19s	remaining:	8.09s
908:	learn:	0.0018718	total:	1m 19s	remaining:	8s
909:	learn:	0.0018718	total:	1m 20s	remaining:	7.91s
910:	learn:	0.0018718	total:	1m 20s	remaining:	7.83s
911:	learn:	0.0018718	total:	1m 20s	remaining:	7.74s
912:	learn:	0.0018718	total:	1m 20s	remaining:	7.65s
913:	learn:	0.0018652	total:	1m 20s	remaining:	7.56s
914:	learn:	0.0018652	total:	1m 20s	remaining:	7.48s
915:	learn:	0.0018568	total:	1m 20s	remaining:	7.39s
916:	learn:	0.0018568	total:	1m 20s	remaining:	7.3s
917:	learn:	0.0018568	total:	1m 20s	remaining:	7.21s
918:	learn:	0.0018568	total:	1m 20s	remaining:	7.13s
919:	learn:	0.0018568	total:	1m 20s	remaining:	7.04s
920:	learn:	0.0018568	total:	1m 21s	remaining:	6.95s
921:	learn:	0.0018568	total:	1m 21s	remaining:	6.86s
922:	learn:	0.0018568	total:	1m 21s	remaining:	6.77s
923:	learn:	0.0018568	total:	1m 21s	remaining:	6.69s
924:	learn:	0.0018568	total:	1m 21s	remaining:	6.6s
925:	learn:	0.0018568	total:	1m 21s	remaining:	6.51s
926:	learn:	0.0018568	total:	1m 21s	remaining:	6.42s
927:	learn:	0.0018567	total:	1m 21s	remaining:	6.33s
928:	learn:	0.0018567	total:	1m 21s	remaining:	6.25s
929:	learn:	0.0018567	total:	1m 21s	remaining:	6.16s
930:	learn:	0.0018566	total:	1m 21s	remaining:	6.07s

931:	learn: 0.0018508	total: 1m 21s	remaining: 5.98s
932:	learn: 0.0018508	total: 1m 22s	remaining: 5.89s
933:	learn: 0.0018508	total: 1m 22s	remaining: 5.81s
934:	learn: 0.0018508	total: 1m 22s	remaining: 5.72s
935:	learn: 0.0018507	total: 1m 22s	remaining: 5.63s
936:	learn: 0.0018507	total: 1m 22s	remaining: 5.54s
937:	learn: 0.0018507	total: 1m 22s	remaining: 5.45s
938:	learn: 0.0018507	total: 1m 22s	remaining: 5.37s
939:	learn: 0.0018507	total: 1m 22s	remaining: 5.28s
940:	learn: 0.0018507	total: 1m 22s	remaining: 5.19s
941:	learn: 0.0018507	total: 1m 22s	remaining: 5.1s
942:	learn: 0.0018507	total: 1m 22s	remaining: 5.01s
943:	learn: 0.0018507	total: 1m 23s	remaining: 4.93s
944:	learn: 0.0018507	total: 1m 23s	remaining: 4.84s
945:	learn: 0.0018507	total: 1m 23s	remaining: 4.75s
946:	learn: 0.0018507	total: 1m 23s	remaining: 4.66s
947:	learn: 0.0018507	total: 1m 23s	remaining: 4.57s
948:	learn: 0.0018506	total: 1m 23s	remaining: 4.49s
949:	learn: 0.0018506	total: 1m 23s	remaining: 4.4s
950:	learn: 0.0018506	total: 1m 23s	remaining: 4.31s
951:	learn: 0.0018506	total: 1m 23s	remaining: 4.22s
952:	learn: 0.0018506	total: 1m 23s	remaining: 4.13s
953:	learn: 0.0018435	total: 1m 23s	remaining: 4.04s
954:	learn: 0.0018435	total: 1m 23s	remaining: 3.96s
955:	learn: 0.0018435	total: 1m 24s	remaining: 3.87s
956:	learn: 0.0018435	total: 1m 24s	remaining: 3.78s
957:	learn: 0.0018435	total: 1m 24s	remaining: 3.69s
958:	learn: 0.0018434	total: 1m 24s	remaining: 3.6s
959:	learn: 0.0018420	total: 1m 24s	remaining: 3.52s
960:	learn: 0.0018420	total: 1m 24s	remaining: 3.43s
961:	learn: 0.0018419	total: 1m 24s	remaining: 3.34s
962:	learn: 0.0018419	total: 1m 24s	remaining: 3.25s
963:	learn: 0.0018419	total: 1m 24s	remaining: 3.17s
964:	learn: 0.0018419	total: 1m 24s	remaining: 3.08s
965:	learn: 0.0018419	total: 1m 24s	remaining: 2.99s
966:	learn: 0.0018419	total: 1m 25s	remaining: 2.9s
967:	learn: 0.0018419	total: 1m 25s	remaining: 2.81s
968:	learn: 0.0018419	total: 1m 25s	remaining: 2.73s
969:	learn: 0.0018419	total: 1m 25s	remaining: 2.64s
970:	learn: 0.0018419	total: 1m 25s	remaining: 2.55s
971:	learn: 0.0018419	total: 1m 25s	remaining: 2.46s
972:	learn: 0.0018419	total: 1m 25s	remaining: 2.37s
973:	learn: 0.0018419	total: 1m 25s	remaining: 2.29s
974:	learn: 0.0018419	total: 1m 25s	remaining: 2.2s
975:	learn: 0.0018419	total: 1m 25s	remaining: 2.11s
976:	learn: 0.0018419	total: 1m 25s	remaining: 2.02s
977:	learn: 0.0018419	total: 1m 25s	remaining: 1.93s
978:	learn: 0.0018419	total: 1m 26s	remaining: 1.85s
979:	learn: 0.0018419	total: 1m 26s	remaining: 1.76s
980:	learn: 0.0018419	total: 1m 26s	remaining: 1.67s
981:	learn: 0.0018419	total: 1m 26s	remaining: 1.58s
982:	learn: 0.0018406	total: 1m 26s	remaining: 1.49s
983:	learn: 0.0018406	total: 1m 26s	remaining: 1.41s
984:	learn: 0.0018406	total: 1m 26s	remaining: 1.32s
985:	learn: 0.0018406	total: 1m 26s	remaining: 1.23s
986:	learn: 0.0018406	total: 1m 26s	remaining: 1.14s
987:	learn: 0.0018406	total: 1m 26s	remaining: 1.05s
988:	learn: 0.0018406	total: 1m 26s	remaining: 967ms
989:	learn: 0.0018406	total: 1m 27s	remaining: 879ms
990:	learn: 0.0018406	total: 1m 27s	remaining: 791ms
991:	learn: 0.0018406	total: 1m 27s	remaining: 703ms
992:	learn: 0.0018405	total: 1m 27s	remaining: 615ms
993:	learn: 0.0018405	total: 1m 27s	remaining: 527ms
994:	learn: 0.0018405	total: 1m 27s	remaining: 440ms
995:	learn: 0.0018405	total: 1m 27s	remaining: 352ms
996:	learn: 0.0018405	total: 1m 27s	remaining: 264ms
997:	learn: 0.0018405	total: 1m 27s	remaining: 176ms
998:	learn: 0.0018405	total: 1m 27s	remaining: 87.9ms
999:	learn: 0.0018405	total: 1m 27s	remaining: 0us

Out[136]:

```
StackingCVClassifier(classifiers=[ExtraTreesClassifier(max_features=0.5,
                                                         n_estimators=200,
                                                         n_jobs=-1,
                                                         random_state=33),
                                SVC(C=100.0, gamma=0.001, max_iter=10000,
                                     random_state=33),
                                XGBClassifier(base_score=0.5,
                                              booster='gbtree',
                                              colsample_bylevel=1,
                                              colsample_bynode=1,
                                              colsample_bytree=0.6, gamma=0,
                                              gpu_id=-1,
                                              importance_type='gain',
                                              interaction_constraints='',
                                              learner='gb',
                                              min_split_gain=0.5,
                                              n_estimators=400,
                                              num_leaves=26,
                                              objective='binary',
                                              random_state=33,
                                              scale_pos_weight=1,
                                              subsample=0.8012317612884369,
                                              subsample_for_bin=200,
                                              subsample_freq=1),
                                KNeighborsClassifier(n_jobs=-1, n_neighbors=1,
                                                      p=1)],
                    cv=5,
                    meta_classifier=LogisticRegression(C=0.1, max_iter=10000,
                                                       penalty='l1',
                                                       random_state=33,
                                                       solver='liblinear'),
                    n_jobs=-1, random_state=33)
```

## 12.3 Assess the performance

In [158]:

```
estimators.append(sclf)
print('Training accuracy: {:.2f}%'.format(
    sclf.score(X_important_train, y_train) * 100))
print('Test accuracy: {:.2f}%'.format(
    sclf.score(X_important_test, y_test) * 100))
```

```
Training accuracy: 100.00%
Test accuracy: 97.37%
```

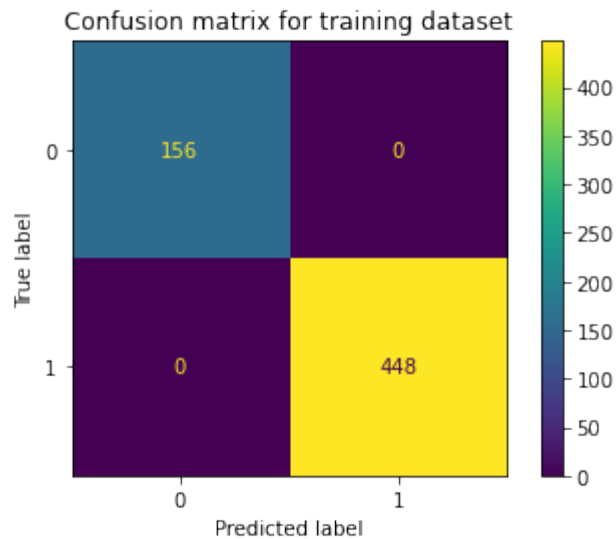
In [139]:

```
print('Performance on training data:\n')
print(
    classification_report(y_train,
                          sclf.predict(X_important_train),
                          labels=[0, 1],
                          digits=4))
plot_confusion_matrix(sclf, X_important_train, y_train, labels=[0, 1])
plt.title('Confusion matrix for training dataset')
plt.show()
```

Performance on training data:

	precision	recall	f1-score	support
0	1.0000	1.0000	1.0000	156
1	1.0000	1.0000	1.0000	448
accuracy			1.0000	604
macro avg	1.0000	1.0000	1.0000	604

weighted avg      1.0000      1.0000      1.0000      604

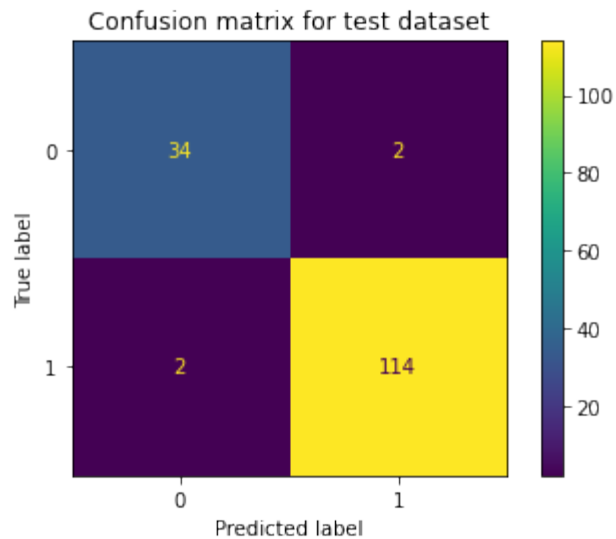


In [140]:

```
print('Performance on test data:\n')
print(
    classification_report(y_test,
                           sclf.predict(X_important_test),
                           labels=[0, 1],
                           digits=4))
plot_confusion_matrix(sclf, X_important_test, y_test, labels=[0, 1])
plt.title('Confusion matrix for test dataset')
plt.show()
```

Performance on test data:

	precision	recall	f1-score	support
0	0.9444	0.9444	0.9444	36
1	0.9828	0.9828	0.9828	116
accuracy			0.9737	152
macro avg	0.9636	0.9636	0.9636	152
weighted avg	0.9737	0.9737	0.9737	152



## 13 Summary

### 13.1 Plot the training accuracy for each estimator

In [159]:

```
estimators.append(best_lr)
est_names.append('Logistic Regression')
```

In [211]:

```
def plot_bar_acc(acc, title):
    plt.figure(figsize=(16, 8))
    ax = sns.barplot(x=est_names, y=acc)
    plt.ylim((0, 1.3))
    ax.yaxis.set_major_formatter(FuncFormatter('{0:.0%}'.format))
    totals = []
    for i in ax.patches:
        totals.append(i.get_height())

    # set individual bar lables using above list
    total = sum(totals)

    # set individual bar lables using above list
    for i, n in zip(ax.patches, range(len(train_acc))):
        # get_x pulls left or right; get_height pushes up or down
        ax.text(i.get_x() + 0.1,
                i.get_height() + .05,
                str(np.round(acc[n], 2) * 100) + '%',
                fontsize=15,
                color='black')
    plt.xlabel('estimators')
    plt.ylabel('accuracy')
    plt.title(title)
    plt.show()
```

In [212]:

```
def print_acc(train_acc, test_acc):
    for trn_acc, tst_acc, name in zip(train_acc, test_acc, est_names):
        print(name + ' training accuracy: {:.2f}'.format(trn_acc*100))
        print(name + ' test accuracy: {:.2f}\n'.format(tst_acc*100))
```

In [213]:

```
train_acc = [estimator.score(X_important_train, y_train) for estimator in estimators]
test_acc = [estimator.score(X_important_test, y_test) for estimator in estimators]
```

In [214]:

```
print_acc(train_acc, test_acc)
```

```
Extra Trees training accuracy: 100.00
Extra Trees test accuracy: 91.45
```

```
SVM training accuracy: 99.83
SVM test accuracy: 92.76
```

```
XGBoost Classifier training accuracy: 100.00
XGBoost Classifier test accuracy: 93.42
```

```
CatBoost Classifier training accuracy: 100.00
CatBoost Classifier test accuracy: 93.42
```

```
LightGBM Classifier training accuracy: 100.00
LightGBM Classifier test accuracy: 93.42
```



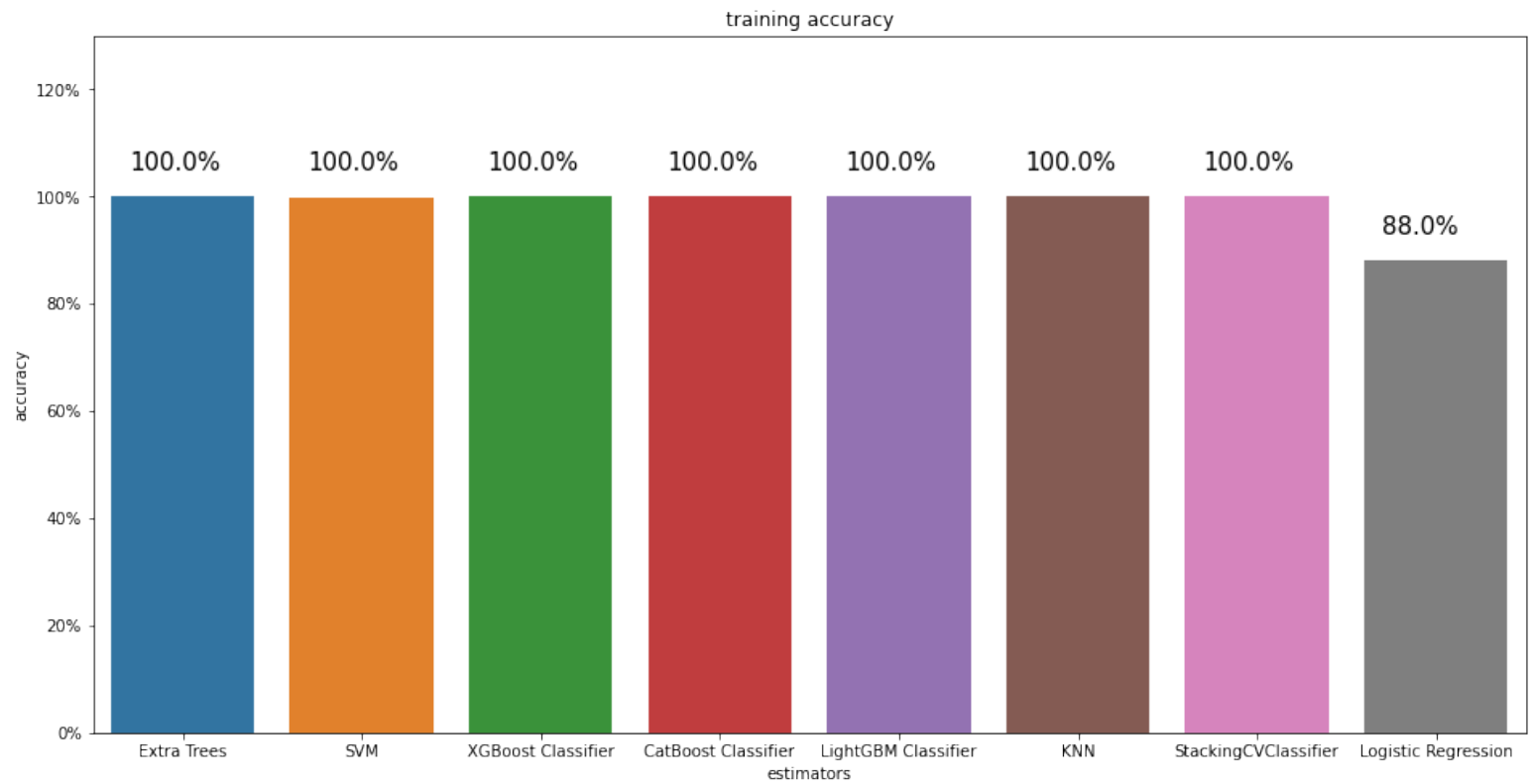
```
KNN training accuracy: 100.00
KNN test accuracy: 97.37

StackingCVClassifier training accuracy: 100.00
StackingCVClassifier test accuracy: 97.37

Logistic Regression training accuracy: 88.08
Logistic Regression test accuracy: 91.45
```

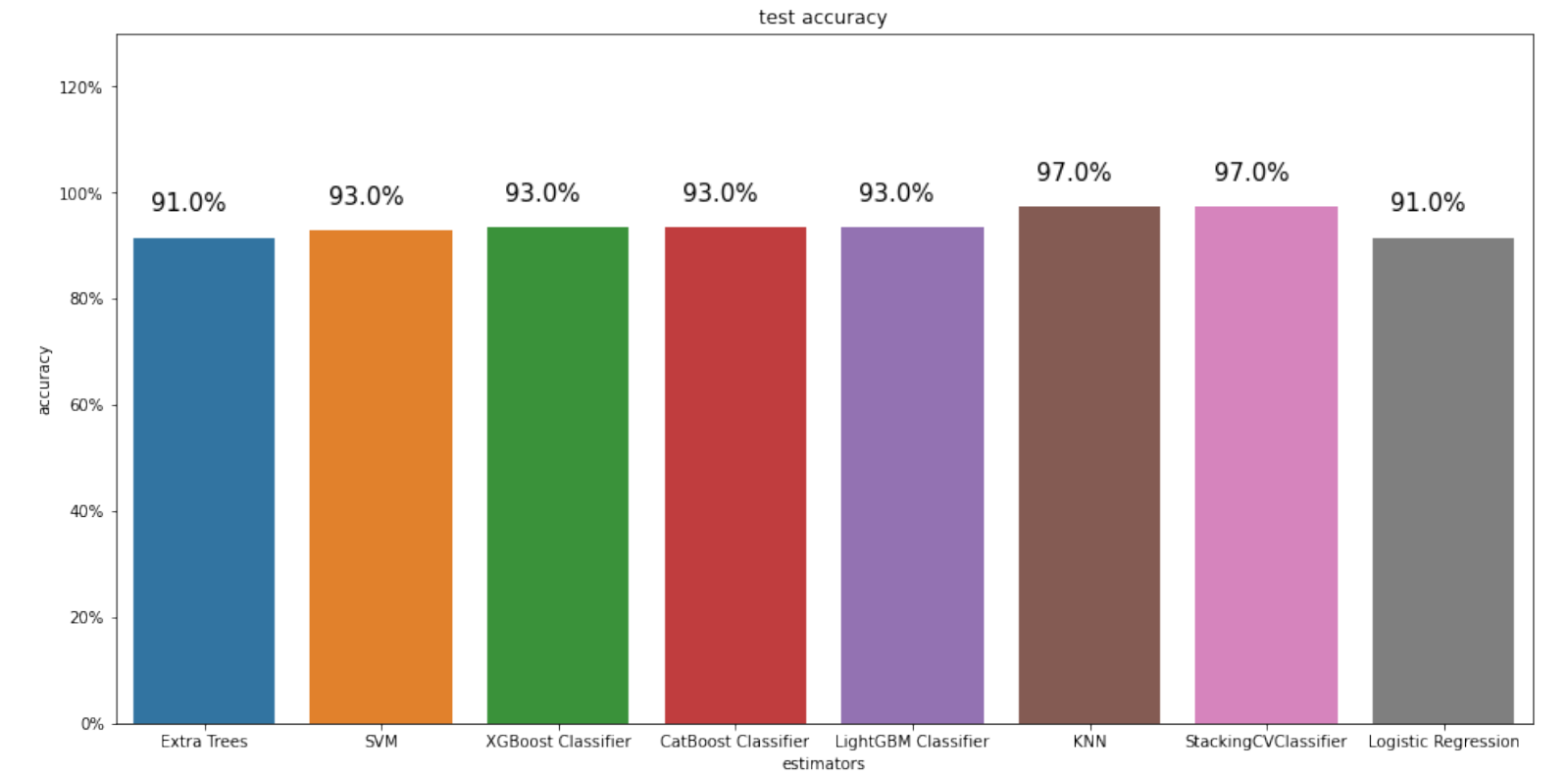
In [215]:

```
plot_bar_acc(train_acc, 'training accuracy')
```



In [216]:

```
plot_bar_acc(test_acc, 'test accuracy')
```



In [ ]: