Case Study 5: Cybersecurity & Firewall Traffic

Matt Farrow July 3, 2022

1 Introduction

In this study, the goal to take a <u>data set</u> of internet traffic records from Firat University in Turkey, and through the use of Support Vector Classifier (SVC) and Stochastic Gradient Descent (SGD) models, make decisions about whether to allow, deny, drop, and reset-both the traffic.

2 Methods

2.1 Data Examination

An initial examination of the data revealed 65,532 observations and 12 features, including 'Action', the multiclass response variable of interest (Table 1).

Feature
Source
Port
Destination Port
NAT Source Port
NAT Destination Port
Action
Bytes
Bytes Sent
Bytes Received
Packets
Elapsed Time (sec)
pkts_sent
pkts received

Table 1: Feature List

In taking a closer look at the response variable, 'Action', I noticed that the data is heavily imbalanced, with allow being the primary action, and reset-both being almost non-existent in the data. As a result, that value was dropped from the data set in order to make a small correction to the imbalanced data.

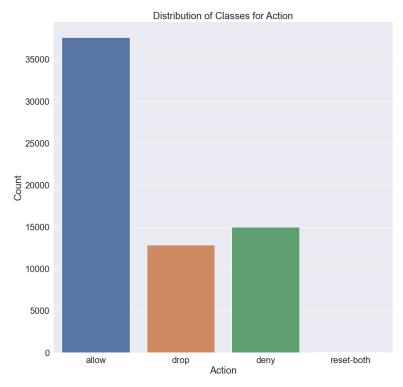


Figure 1: Distribution of Response Variable

The one piece of feature engineering that I did was to create an attribute to note, through either a 0 or 1, whether the source or destination port required network address translation (NAT). I then examined the pairwise correlations of the numeric features in the data. Unsurprisingly, bytes is closey related to packets and pkts_sent, but I ended up deciding to leave those features in the data.

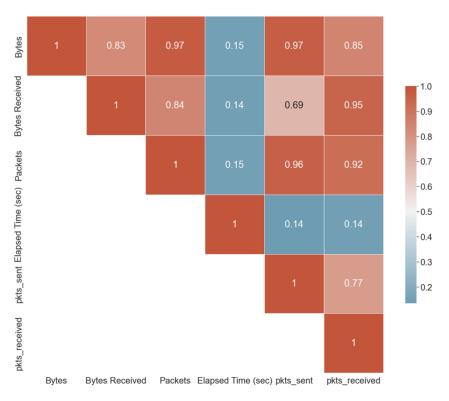


Figure 2: Pairwise Correlation of Numeric Variables

2.2 Model Preparation & Execution

The data was split into test and training data sets using an 80/20 split with a stratified shuffle and columns that needed to be one-hot encoded were identified (Table 2).

Feature
Source Port
Destination Port
NAT Source Port
NAT Destination Port
Source Need NAT
Destination Need NAT

Table 2: Features to One-Hot Encode

A pipeline was built to define the model and handle the one-hot encoding and scaling of the numeric features before running everything through a randomized cross-validation with a hyperparameter search for the SVC (Table 3) and SGD (Table 4) models.

Attribute	Values to Search
С	0.001, 0.01, 0.1, 1, 10, 100
Gamma	'scale', 'auto', 1, 0.1, 0.01, 0.001, 0.0001
Kernel	'linear', 'poly', 'rbf'

Table 3: SVC Hyperparameter Search

Attribute	Values to Search
Alpha	0.000001, 0.00001, 0.0001, 0.001, 0.01
Eta0 (learning rate 'optimal')	0
Eta0 (learning rate 'constant')	0.01, 0.1, 1.0, 5, 10
Class Weight	'balanced

Table 4: SGD Hyperparameter Search

Additionally, in order to try and visualize the work that the SVC model was doing, the response variable was collapsed to only two potential actions — allow and deny. That allowed me to plot the results and see the impact of the hyperparameter choices, rather than simply visualize them through confusion matrices. In that model, the hyperparameter list was also simplified (Table 5).

Attribute	Values to Search
Kernel	'linear', 'rbf'
C	0.001, 0.01, 0.1, 1, 10, 100

Table 5: SVC Model (Allow/Deny) Hyperparameter Search

3 Results

3.1 Support Vector Classifier Model

The SVC model performed with an extremely high score across all measures (Table 6). Additionally, the confusion matrix (Figure 1) details the small number of error made in the model with deny being the most frequent incorrect classification (as allow).

	Precision	Recall	F1-Score	Support	
allow	0.999334	0.997078	0.998205	7528	
deny	0.992027	0.995997	0.994008	2998	
drop	0.997287	0.999222	0.998251	2570	
accuracy			0.997251	13096	
macro avg	0.996214	0.997432	0.996821	13096	
weighted avg	0.997259	0.997251	0.997253	13096	

Table 6: SVC Model Performance

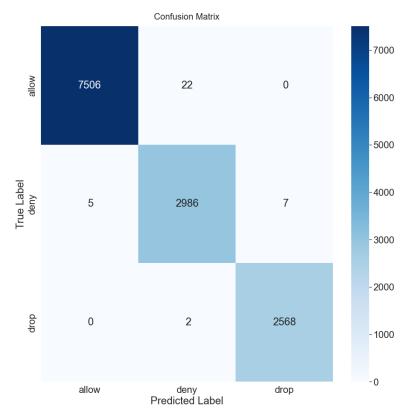


Figure 3: SVC Confusion Matrix

The linear kernel shows an extremely high accuracy score, followed by rbf, and finally poly has a significant drop off in accuracy (Figure 4). Examining the accuracy performance of the 'C' parameter, I note that the performance of all values below 1 is 57.5% accuracy, but at C = 1, the accuracy jumps to 91.9% (Figure 5).

Kernel	Accuracy Score
linear	0.9928
poly	0.5748
rbf	0.9187

Figure 4: SVC Kernel Accuracy

С	Accuracy Score
0.001	0.5748
0.01	0.5748
0.1	0.5748
1	0.9187
10	0.9193
100	0.9497

Figure 5: SVC 'C' Accuracy

3.2 Stochastic Gradient Descent Model

For all intents and purposes, the model performance of the SGD model is almost identical to that of the SVC model (Table 7).

	Danaisian	D 11	E1 C	Comment	
	Precision	Recall	F1-Score	Support	
allow	0.999334	0.996148	0.997738	7528	
deny	0.990381	0.995997	0.993181	2998	
drop	0.997284	1.000000	0.998640	2570	
accuracy			0.996869	13096	
macro avg	0.995666	0.997382	0.996520	13096	
weighted avg	0.996882	0.996869	0.996872	13096	

 Table 7: SGD Model Performance

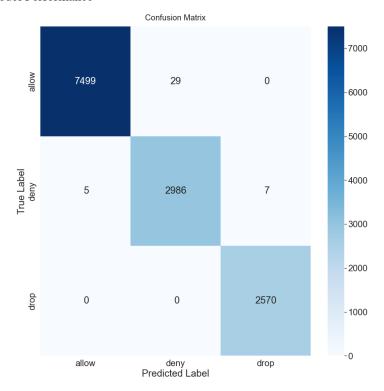


Figure 6: SGD Confusion Matrix

3.3 Two-Class SVC Model

For the two-class SVC model, the training and testing accuracy were both 99.8% and the confusion matrix shows the small number of misclassifications that the model made (Figure 7).

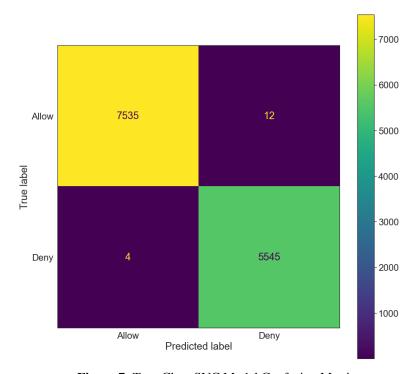


Figure 7: Two-Class SVC Model Confusion Matrix

In terms of kernel performance, the 'linear' kernel returned a 99.9% accuracy whereas the 'rbf' kernel was only able to return 77.9% accuracy. This is able to be visualized in the plots which show the distribution of points and the fitted hyperplane (Figure 8).

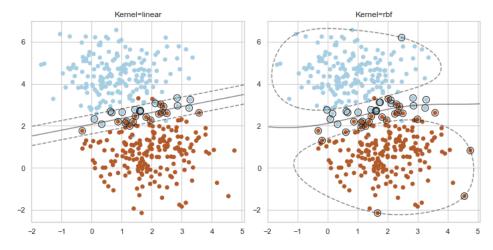


Figure 8: Kernel Hyperplane Performance and Margin

Visualizing the performance of the linear (Figure 9) and rbf (Figure 10) kernels, I was able to see how the different levels of 'C' affected the accuracy results in the original models. When C < 1, it is possible to see the points that the model did not correctly capture through the margin and hyperplane.

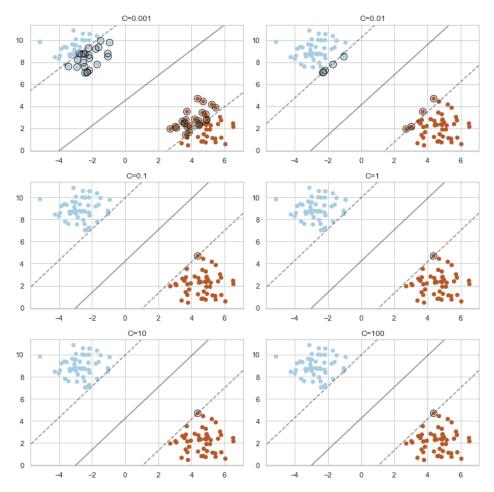


Figure 9: Linear Kernel Performance at Different Levels of 'C'

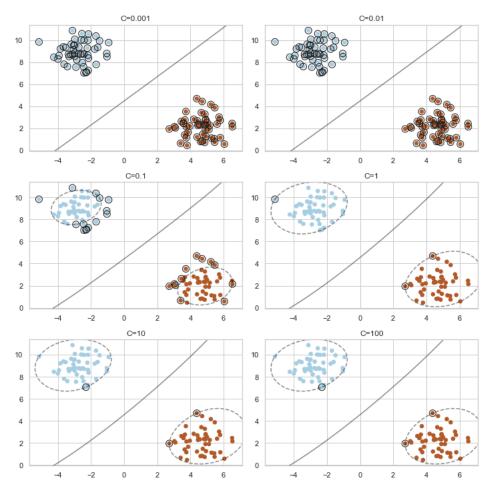


Figure 10: RBF Kernel Performance at Different Levels of 'C'

4 Conclusion

In conclusion, all three models were able to produce incredibly accurate accuracy scores which does give me pause to think that more work should have been done to take multicollinearity into account. In terms of model preference, I'm partial to the two-class SVC model due to the ability to visualize the hyperplane, margin, and model performance.

Appendix

Code

Code begins on the following page.

Case Study 5

Description

Build a classifier to determine whether or not firewall requests should be automated. source

Attributes

- Source Port
- Destination Port
- NAT Source Port
- NAT Destination Port
- Action
- Bytes
- Bytes Sent,
- Bytes Received
- Packets
- Elapsed Time (sec)
- pkts_sent
- pkts_received

Response (Action)

- allow
- action
- drop
- reset-both

Class Notes

- Try to do all four classes but they are "horribly" imbalanced (one has very few samples)
- bytes, packets are ordinal (whole integers)
- Ports are in a sense categorical (needs one hot encoded)
 - Can try to group if desired
- Make sure to do random shuffle when splitting/cross validation
- Column E (allow, not allow, etc) is target variable
- Do test/train split instead of cross validation (takes too long)
- Can make a new features out of pkts_sent and pkts_received (maybe the delta?)

Setup

Load Libraries

```
In [1]: # Import standard libraries
        import matplotlib.pyplot as plt
        import numpy as np
        import pandas as pd
        import seaborn as sns
        # Import sklearn libraries
        from sklearn.compose import ColumnTransformer
        from sklearn.datasets import make blobs
        from sklearn.linear model import SGDClassifier
        from sklearn.metrics import accuracy score
        from sklearn.metrics import auc
        from sklearn.metrics import classification report
        from sklearn.metrics import confusion_matrix
        from sklearn.metrics import ConfusionMatrixDisplay
        from sklearn.metrics import f1 score
        from sklearn.metrics import plot confusion matrix
        from sklearn.metrics import plot roc curve
        from sklearn.metrics import precision score
        from sklearn.metrics import recall_score
        from sklearn.metrics import roc curve
        from sklearn.model selection import GridSearchCV
        from sklearn.model selection import RandomizedSearchCV
        from sklearn.model selection import StratifiedKFold
        from sklearn.model selection import train test split
        from sklearn.multiclass import OneVsRestClassifier
        from sklearn.pipeline import Pipeline
        from sklearn.preprocessing import MinMaxScaler
        from sklearn.preprocessing import OneHotEncoder
        from sklearn.svm import LinearSVC
        from sklearn.svm import SVC
        # Import other libraries
        import os
        import warnings
        import joblib
        from mlxtend.plotting import plot decision regions
        # Set options
        pd.set_option("display.max_columns", None)
        pd.options.display.max_rows = 10000
        pd.options.display.max_columns = 10000
        warnings.filterwarnings("ignore")
        get ipython().run line magic("matplotlib", "inline")
```

Load Data

In [2]: df = pd.read_csv('log2.csv')

Exploratory Data Analysis

In [3]: df.head()

Out[3]:

	Source Port	Destination Port	NAT Source Port	NAT Destination Port	Action	Bytes	Bytes Sent	Bytes Received	Packets	Elap: Ti (s
0	57222	53	54587	53	allow	177	94	83	2	
1	56258	3389	56258	3389	allow	4768	1600	3168	19	
2	6881	50321	43265	50321	allow	238	118	120	2	1
3	50553	3389	50553	3389	allow	3327	1438	1889	15	
4	50002	443	45848	443	allow	25358	6778	18580	31	

In [4]: df.tail()

Out[4]:

	Source Port	Destination Port	NAT Source Port	NAT Destination Port	Action	Bytes	Bytes Sent	Bytes Received	Packet
65527	63691	80	13237	80	allow	314	192	122	
65528	50964	80	13485	80	allow	4680740	67312	4613428	467
65529	54871	445	0	0	drop	70	70	0	
65530	54870	445	0	0	drop	70	70	0	
65531	54867	445	0	0	drop	70	70	0	

In [5]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 65532 entries, 0 to 65531
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	Source Port	65532 non-null	int64
1	Destination Port	65532 non-null	int64
2	NAT Source Port	65532 non-null	int64
3	NAT Destination Port	65532 non-null	int64
4	Action	65532 non-null	object
5	Bytes	65532 non-null	int64
6	Bytes Sent	65532 non-null	int64
7	Bytes Received	65532 non-null	int64
8	Packets	65532 non-null	int64
9	Elapsed Time (sec)	65532 non-null	int64
10	pkts_sent	65532 non-null	int64
11	pkts_received	65532 non-null	int64
41		1 \	

dtypes: int64(11), object(1)

memory usage: 6.0+ MB

In [6]: df.shape

Out[6]: (65532, 12)

In [7]: df.describe()

Out[7]:

	Source Port	Destination Port	NAT Source Port	NAT Destination Port	Bytes	Bytes
count	65532.000000	65532.000000	65532.000000	65532.000000	6.553200e+04	6.553200€
mean	49391.969343	10577.385812	19282.972761	2671.049930	9.712395e+04	2.238580
std	15255.712537	18466.027039	21970.689669	9739.162278	5.618439e+06	3.8281396
min	0.000000	0.000000	0.000000	0.000000	6.000000e+01	6.000000
25%	49183.000000	80.000000	0.000000	0.000000	6.600000e+01	6.600000
50%	53776.500000	445.000000	8820.500000	53.000000	1.680000e+02	9.000000
75%	58638.000000	15000.000000	38366.250000	443.000000	7.522500e+02	2.1000000
max	65534.000000	65535.000000	65535.000000	65535.000000	1.269359e+09	9.4847726

In [8]: # Missing values?
 df.isnull().sum()

```
Out[8]: Source Port
                                  0
         Destination Port
                                  0
         NAT Source Port
                                  0
         NAT Destination Port
                                  0
         Action
         Bytes
                                  0
                                  0
         Bytes Sent
                                  0
         Bytes Received
         Packets
                                  0
         Elapsed Time (sec)
         pkts_sent
                                  0
         pkts_received
         dtype: int64
 In [9]: df.nunique()
                                  22724
         Source Port
 Out[9]:
         Destination Port
                                   3273
         NAT Source Port
                                  29152
         NAT Destination Port
                                   2533
         Action
                                       4
         Bytes
                                  10724
         Bytes Sent
                                   6683
         Bytes Received
                                   8814
         Packets
                                   1116
         Elapsed Time (sec)
                                    915
         pkts_sent
                                    749
         pkts_received
                                    922
         dtype: int64
In [10]: df['Action'].value_counts()
         allow
                        37640
Out[10]:
         deny
                        14987
         drop
                        12851
         reset-both
                           54
         Name: Action, dtype: int64
In [11]: # Pairwise Correlation
          df.corr()
```

_		Γa.	a 7	
	ut	1.1	1 1	
U	u L	1 4		

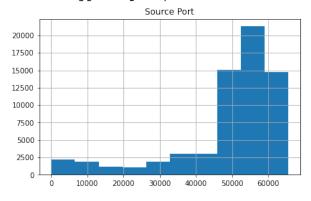
		Source Port	Destination Port	NAT Source Port	NAT Destination Port	Bytes	Bytes Sent	Byt Receiv
	Source Port	1.000000	-0.332246	0.145391	-0.024843	0.000221	-0.000931	0.0019
	Destination Port	-0.332246	1.000000	-0.281676	0.410042	-0.005297	0.001675	-0.0146
	NAT Source Port	0.145391	-0.281676	1.000000	0.178435	0.010659	0.002242	0.0208
	NAT Destination Port	-0.024843	0.410042	0.178435	1.000000	0.003975	0.007904	-0.0032
	Bytes	0.000221	-0.005297	0.010659	0.003975	1.000000	0.933462	0.8302
	Bytes Sent	-0.000931	0.001675	0.002242	0.007904	0.933462	1.000000	0.5750
	Bytes Received	0.001950	-0.014684	0.020827	-0.003216	0.830225	0.575047	1.0000
	Packets	-0.001742	-0.006063	0.012633	0.004605	0.974379	0.887596	0.8430
	Elapsed Time (sec)	-0.046515	0.023537	0.141485	0.219776	0.148834	0.126039	0.1436
	pkts_sent	-0.001422	-0.002134	0.007180	0.006136	0.966548	0.973976	0.6909
	pkts_received	-0.001962	-0.010909	0.018772	0.001747	0.850209	0.639098	0.9460

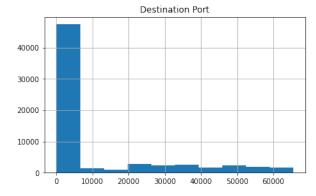
NAT

NAT

```
In [12]: df.hist(figsize = (15, 30), layout = (6, 2))
```

```
Out[12]:
```







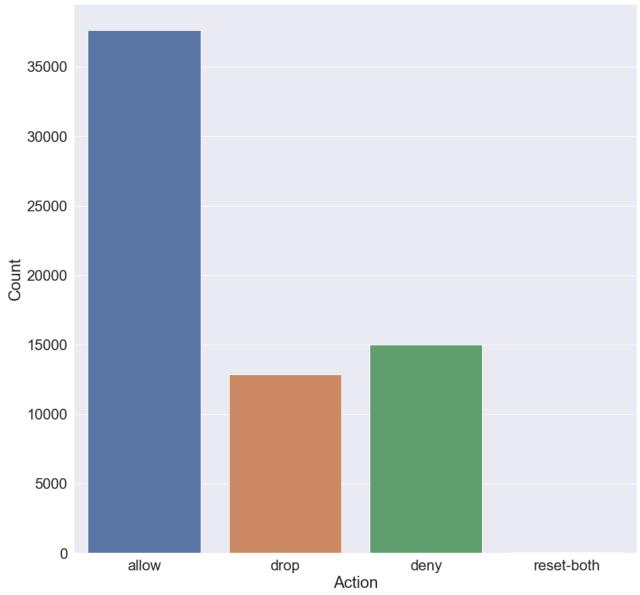




```
In [13]: sns.set(rc = {'figure.figsize':(15,15)})
sns.set(font_scale = 2)
p = sns.countplot(df['Action'])
p.set_title('Distribution of Classes for Action')
p.set_ylabel('Count')
```

Out[13]: Text(0, 0.5, 'Count')

Distribution of Classes for Action



```
In [14]: # Drop 'reset-both' since it doesn't even register on the bar graph
df = df.loc[df['Action'] != 'reset-both']
```

In [15]: print(f"Source Port Unique Levels: {len(df['Source Port'].unique())}")
 print(f"Destination Port Unique Levels: {len(df['Destination Port'].unique())}
 print(f"NAT Source Port Unique Levels: {len(df['NAT Source Port'].unique())}
 print(f"NAT Destination Port Unique Levels: {len(df['NAT Destination Port'].

Source Port Unique Levels: 22692 Destination Port Unique Levels: 3264 NAT Source Port Unique Levels: 29143 NAT Destination Port Unique Levels: 2533

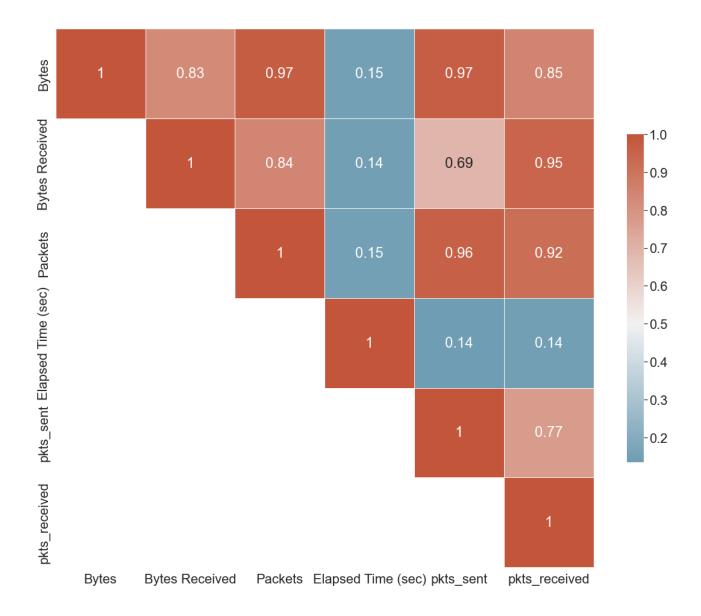
```
In [16]:
          # Does the source or destination port need NAT?
          def nat_needed(num: int) -> int:
              if num == 0:
                   return num
              else:
                   return 1
          df['Source Need NAT'] = df['NAT Source Port'].apply(lambda x: nat_needed(x))
          df['Destination Need NAT'] = df['NAT Destination Port'].apply(lambda x: nat
          df.head()
Out[16]:
                                  NAT
                                             NAT
                                                                                       Elap:
             Source Destination
                                                                Bytes
                                                                         Bytes
                               Source Destination Action
                                                         Bytes
                                                                               Packets
                                                                                          Ti
               Port
                          Port
                                                                 Sent Received
                                  Port
                                             Port
                                                                                          (s
                                                                                     2
          0
             57222
                            53
                                54587
                                              53
                                                           177
                                                                  94
                                                                           83
                                                   allow
             56258
                          3389
                                56258
                                            3389
                                                   allow
                                                          4768
                                                                 1600
                                                                          3168
                                                                                    19
          2
               6881
                         50321
                                43265
                                           50321
                                                   allow
                                                           238
                                                                  118
                                                                           120
                                                                                     2
          3
             50553
                          3389
                                50553
                                            3389
                                                   allow
                                                          3327
                                                                1438
                                                                          1889
                                                                                    15
          4
             50002
                          443
                                45848
                                             443
                                                   allow 25358
                                                                6778
                                                                         18580
                                                                                    31
In [17]: # Define numeric variables
          df numeric = df[['Bytes','Bytes Received','Packets','Elapsed Time (sec)','pk
In [18]: # Check correlation of numeric variables
          corr = df numeric.corr().abs()
          upper tri = corr.where(np.triu(np.ones(corr.shape),k = 0).astype(np.bool))
          sns.set(font scale = 2)
          sns.set_style(style = "white")
          plt.subplots(figsize = (20,20))
          cmap = sns.diverging_palette(230, 20, as_cmap=True)
```

```
Out[18]: <AxesSubplot:>
```

sns.heatmap(upper_tri,

cmap = cmap, vmax = 1, center = 0.5, annot = True, square = True, linewidths = .5,

cbar kws = {"shrink": .5})



Test/Train Splits

```
In [19]: X, y = df.drop(['Action'], inplace = False, axis = 1), df['Action']
         X train, X test, y train, y test = train test split(X, y,
                                                               test_size = 0.20,
                                                               shuffle = True,
                                                               random_state = 1,
                                                               stratify = y
                                                              )
In [20]:
         # Define columns to one-hot encode
         df_ohe = ['Source Port',
                    'Destination Port',
                    'NAT Source Port',
                    'NAT Destination Port',
                    'Source Need NAT',
                    'Destination Need NAT']
         # Define remaining numeric columns
         df_numeric = [names for names in X.columns if names not in df_ohe]
```

Helper Functions

```
In [21]: # Define cross-validation
         cv = StratifiedKFold(n splits = 10, shuffle = True, random state = 1)
         # Define cross-validation summary
         def cv summary(estimator, columns):
             df = pd.DataFrame(estimator.cv results )
             cv result summary = df[columns]
             cv_result_summary.index = np.arange(1,len(cv_result_summary)+1)
             cv_result_summary = cv_result_summary.reset_index()
             cv_result_summary = cv_result_summary.rename(columns = {
                  "index": "param combination", "mean test score": "mean validation so
                 "rank_test_score": "rank_validation_score"
             })
             return cv result summary
         # Define get accuracy score
         def get acc score(model, x, y):
             Calculates socre from estimator (gridsearch) inherit scorer
             return model.score(x, y)
         # Define get confusion matrix
         def get_confusion_matrix(y, yhat,mat_title="Confusion Matrix"):
             x_axis_labels = ['allow','deny','drop']
             y axis labels = ['allow','deny','drop']
             cm_n = confusion_matrix(y, yhat)
             ax = sns.heatmap(cm_n,cmap='Blues',annot=True,fmt='2d',xticklabels=x axi
             ax.set(xlabel='Predicted Label', ylabel='True Label')
             ax.set_title(mat_title,fontdict={'fontsize':20},pad=15)
         # Define get classification report
         def get classification report(x train, y train, x test, y test, pred, model)
             This function is used to get comprehensive classification report for bin
             Training Accuracy, Test Accuracy, print classification_report, plot conf
             print(f"Training Score: {get acc score(model, x train, y train):,.6f}")
             print("Test result:")
             print(classification_report(y_test,pred,digits=6))
             get confusion_matrix(y_test,pred)
```

Support Vector Classifier Model

```
In [22]: params_svc = {
    "model__C": [0.001, 0.01, 0.1, 1, 10, 100],
    "model__gamma": ["scale", "auto", 1, 0.1, 0.01, 0.001, 0.0001],
    "model__kernel":["linear", "poly", "rbf"],
    "model__class_weight":["balanced"]
}
```

```
In [23]: | %time
          # Define preprocessing for columns
          t = [('cat', OneHotEncoder(handle unknown='ignore'), df ohe),
               ('scale', MinMaxScaler(), df numeric)]
          col transform = ColumnTransformer(t)
          # Define model
          model_svc = SVC()
          # Define pipeline
          pipeline_svc = Pipeline(steps=[('preprocessing',col_transform), ('model', model', model')
          # Define model
          clf_svc = RandomizedSearchCV(pipeline_svc,
                                        params svc,
                                        n iter = 10,
                                        scoring = 'f1_weighted',
                                        random state = 1,
                                        verbose = 15,
                                        cv = cv
                                        n jobs = -1,
                                        return_train_score = True)
         CPU times: user 2 \mus, sys: 1e+03 ns, total: 3 \mus
         Wall time: 5.01 \mu s
In [24]: | %time
          # Run model
          search_svc = clf_svc.fit(X_train, y_train)
         CPU times: user 1 \mus, sys: 1e+03 ns, total: 2 \mus
         Wall time: 4.77 \mus
         Fitting 10 folds for each of 10 candidates, totalling 100 fits
In [25]: df_search_svc_cv_results_columns = ['param_model_kernel',
                                                'param_model_gamma',
                                                'param model class weight',
                                                'param model C',
                                                'mean_train_score',
                                                'mean test score',
                                                'rank test score']
          df search svc cv results = cv summary(search svc, df search svc cv results d
          df_search_svc_cv_results
```

Out[25]:		param_combination	param_modelkernel	param_modelgamma	param_modelclass_v
	0	1	linear	0.001	ba
	1	2	poly	auto	ba
	2	3	poly	0.1	ba
	3	4	linear	1	ba
	4	5	linear	0.0001	ba
	5	6	rbf	0.1	ba
	6	7	rbf	0.01	ba
	7	8	poly	0.0001	ba
	8	9	rbf	0.1	ba
	9	10	linear	scale	ba

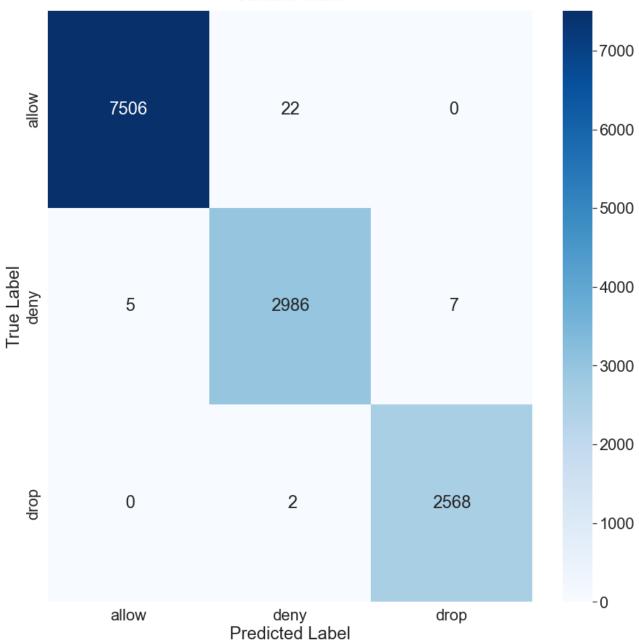
In [26]: predict = search_svc.predict(X_test)
get_classification_report(X_train, y_train, X_test, y_test, predict, search_

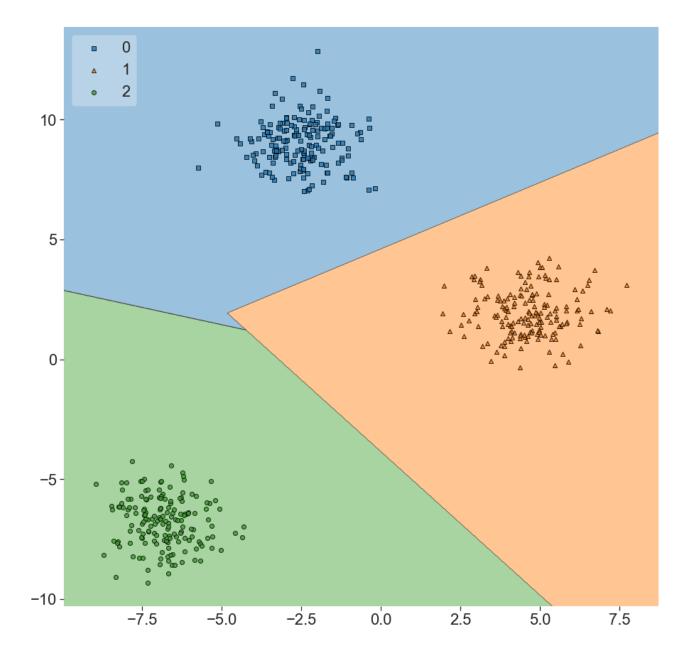
Training Score: 0.998417

Test result:

	precision	recall	f1-score	support
allow	0.999334	0.997078	0.998205	7528
deny	0.992027	0.995997	0.994008	2998
drop	0.997282	0.999222	0.998251	2570
accuracy			0.997251	13096
macro avg	0.996214	0.997432	0.996821	13096
weighted avg	0.997259	0.997251	0.997253	13096







Evaluate Kernels

ma=0.1, model kernel=rbf

```
In [28]: p = ['linear', 'poly', 'rbf']

for i in p:
    model = SVC(kernel = i)
    model.fit(X_train, y_train)
    y_hat_rf_test = model.predict(X_test)
    print(i,":", accuracy_score(y_hat_rf_test, y_test))

[CV 3/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=auto, model__kernel=poly
[CV 3/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamma =auto, model__kernel=poly;, score=(train=0.064, test=0.065) total time=23.7m in
[CV 7/10; 6/10] START model C=0.1, model_class_weight=balanced, model_gam
```

[CV 7/10; 6/10] END model__C=0.1, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.996, test=0.997) total time= 1.7min

```
[CV 7/10; 7/10] START model C=100, model class weight=balanced, model gam
ma=0.01, model kernel=rbf
[CV 7/10; 7/10] END model C=100, model class weight=balanced, model gamma
=0.01, model_kernel=rbf;, score=(train=0.998, test=0.998) total time= 3.1mi
[CV 3/10; 9/10] START model C=100, model class weight=balanced, model gam
ma=0.1, model kernel=rbf
[CV 3/10; 9/10] END model__C=100, model__class_weight=balanced, model__gamma
=0.1, model kernel=rbf;, score=(train=0.998, test=0.997) total time= 3.1min
[CV 7/10; 9/10] START model C=100, model class weight=balanced, model gam
ma=0.1, model__kernel=rbf
[CV 7/10; 9/10] END model C=100, model class weight=balanced, model gamma
=0.1, model kernel=rbf;, score=(train=0.998, test=0.998) total time= 3.2min
[CV 8/10; 10/10] START model C=0.1, model class weight=balanced, model ga
mma=scale, model kernel=linear
[CV 8/10; 10/10] END model C=0.1, model class weight=balanced, model gamm
a=scale, model__kernel=linear;, score=(train=0.996, test=0.996) total time=
1.4min
[CV 1/10; 1/10] START model C=100, model class weight=balanced, model gam
ma=0.001, model__kernel=linear
[CV 1/10; 1/10] END model C=100, model class weight=balanced, model gamma
=0.001, model_kernel=linear;, score=(train=0.998, test=0.998) total time= 2
.6min
[CV 6/10; 3/10] START model C=1, model class weight=balanced, model gamma
=0.1, model_kernel=poly
[CV 6/10; 3/10] END model C=1, model class weight=balanced, model gamma=0
.1, model kernel=poly;, score=(train=0.996, test=0.996) total time= 3.4min
[CV 2/10; 4/10] START model C=10, model class weight=balanced, model gamm
a=1, model kernel=linear
[CV 2/10; 4/10] END model C=10, model class weight=balanced, model gamma=
1, model kernel=linear;, score=(train=0.998, test=0.997) total time= 1.4min
[CV 4/10; 4/10] START model__C=10, model__class_weight=balanced, model__gamm
a=1, model kernel=linear
[CV 4/10; 4/10] END model C=10, model class weight=balanced, model gamma=
1, model_kernel=linear;, score=(train=0.998, test=0.998) total time= 1.6min
[CV 9/10; 4/10] START model C=10, model class weight=balanced, model gamm
a=1, model__kernel=linear
[CV 9/10; 4/10] END model C=10, model class weight=balanced, model gamma=
1, model_kernel=linear;, score=(train=0.998, test=0.996) total time= 7.0min
[CV 5/10; 5/10] START model C=10, model class weight=balanced, model gamm
a=0.0001, model kernel=linear
[CV 5/10; 5/10] END model C=10, model class weight=balanced, model gamma=
0.0001, model kernel=linear;, score=(train=0.998, test=0.997) total time= 2
.3min
[CV 1/10; 6/10] START model__C=0.1, model__class_weight=balanced, model__gam
ma=0.1, model kernel=rbf
[CV 1/10; 6/10] END model C=0.1, model class weight=balanced, model gamma
=0.1, model_kernel=rbf;, score=(train=0.996, test=0.997) total time= 2.2min
[CV 8/10; 7/10] START model__C=100, model__class_weight=balanced, model__gam
ma=0.01, model__kernel=rbf
[CV 8/10; 7/10] END model C=100, model class weight=balanced, model gamma
=0.01, model kernel=rbf;, score=(train=0.998, test=0.997) total time= 3.2mi
[CV 6/10; 9/10] START model C=100, model class weight=balanced, model gam
ma=0.1, model kernel=rbf
[CV 6/10; 9/10] END model C=100, model class weight=balanced, model gamma
```

=0.1, model__kernel=rbf;, score=(train=0.999, test=0.997) total time= 3.6min

```
[CV 2/10; 10/10] START model__C=0.1, model__class_weight=balanced, model__ga mma=scale, model kernel=linear
```

- [CV 2/10; 10/10] END model__C=0.1, model__class_weight=balanced, model__gamm a=scale, model__kernel=linear;, score=(train=0.996, test=0.996) total time= 1.1min
- [CV 3/10; 10/10] START model__C=0.1, model__class_weight=balanced, model__ga mma=scale, model kernel=linear
- [CV 3/10; 10/10] END model__C=0.1, model__class_weight=balanced, model__gamm a=scale, model__kernel=linear;, score=(train=0.996, test=0.995) total time= 1.4min
- [CV 5/10; 10/10] START model__C=0.1, model__class_weight=balanced, model__ga mma=scale, model kernel=linear
- [CV 5/10; 10/10] END model__C=0.1, model__class_weight=balanced, model__gamm a=scale, model__kernel=linear;, score=(train=0.996, test=0.996) total time= 1.9min
- [CV 3/10; 1/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.001, model__kernel=linear
- [CV 3/10; 1/10] END model__C=100, model__class_weight=balanced, model__gamma =0.001, model__kernel=linear;, score=(train=0.998, test=0.997) total time= 2 .3min
- [CV 8/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__gam
 ma=auto, model__kernel=poly
- [CV 8/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamma =auto, model__kernel=poly;, score=(train=0.064, test=0.064) total time=23.6m in
- [CV 6/10; 7/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.01, model kernel=rbf
- [CV 6/10; 7/10] END model__C=100, model__class_weight=balanced, model__gamma =0.01, model__kernel=rbf;, score=(train=0.998, test=0.997) total time= 3.2mi n
- [CV 4/10; 9/10] START model__C=100, model__class_weight=balanced, model__gam
 ma=0.1, model__kernel=rbf
- [CV 4/10; 9/10] END model__C=100, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.998, test=0.998) total time= 3.3min [CV 9/10; 9/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.1, model__kernel=rbf
- [CV 9/10; 9/10] END model__C=100, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.998, test=0.996) total time= 3.0min [CV 7/10; 10/10] START model__C=0.1, model__class_weight=balanced, model__ga mma=scale, model kernel=linear
- [CV 7/10; 10/10] END model__C=0.1, model__class_weight=balanced, model__gamm a=scale, model__kernel=linear;, score=(train=0.996, test=0.997) total time= 1.6min
- [CV 2/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=auto, model kernel=poly
- [CV 2/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamma =auto, model__kernel=poly;, score=(train=0.085, test=0.085) total time=23.6m in
- [CV 4/10; 6/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=0.1, model__kernel=rbf
- [CV 4/10; 6/10] END model__C=0.1, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.996, test=0.997) total time= 1.7min [CV 9/10; 7/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.01, model__kernel=rbf
- [CV 9/10; 7/10] END model__C=100, model__class_weight=balanced, model__gamma =0.01, model__kernel=rbf;, score=(train=0.998, test=0.996) total time= 3.1mi

```
[CV 5/10; 9/10] START model__C=100, model__class_weight=balanced, model__gam
ma=0.1, model__kernel=rbf
```

- [CV 5/10; 9/10] END model__C=100, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.998, test=0.997) total time= 3.6min [CV 1/10; 10/10] START model__C=0.1, model__class_weight=balanced, model__ga mma=scale, model__kernel=linear
- [CV 1/10; 10/10] END model__C=0.1, model__class_weight=balanced, model__gamm a=scale, model__kernel=linear;, score=(train=0.996, test=0.997) total time= 1.2min
- [CV 4/10; 10/10] START model__C=0.1, model__class_weight=balanced, model__ga mma=scale, model__kernel=linear
- [CV 4/10; 10/10] END model__C=0.1, model__class_weight=balanced, model__gamm a=scale, model__kernel=linear;, score=(train=0.996, test=0.997) total time= 1.5min
- [CV 6/10; 10/10] START model__C=0.1, model__class_weight=balanced, model__ga mma=scale, model kernel=linear
- [CV 6/10; 10/10] END model__C=0.1, model__class_weight=balanced, model__gamm a=scale, model__kernel=linear;, score=(train=0.996, test=0.996) total time= 2.3min
- linear : 0.9927458766035431
- [CV 6/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=auto, model__kernel=poly
- [CV 6/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamma =auto, model__kernel=poly;, score=(train=0.064, test=0.064) total time=23.6m in
- [CV 3/10; 6/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=0.1, model__kernel=rbf
- [CV 3/10; 6/10] END model__C=0.1, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.996, test=0.995) total time= 1.6min [CV 5/10; 7/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.01, model kernel=rbf
- [CV 5/10; 7/10] END model__C=100, model__class_weight=balanced, model__gamma =0.01, model__kernel=rbf;, score=(train=0.998, test=0.997) total time= 3.1mi n
- [CV 1/10; 9/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.1, model__kernel=rbf
- [CV 1/10; 9/10] END model__C=100, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.998, test=0.997) total time= 3.5min [CV 8/10; 9/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.1, model kernel=rbf
- [CV 8/10; 9/10] END model__C=100, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.998, test=0.997) total time= 3.9min [CV 9/10; 10/10] START model__C=0.1, model__class_weight=balanced, model__ga mma=scale, model__kernel=linear
- [CV 9/10; 10/10] END model__C=0.1, model__class_weight=balanced, model__gamm a=scale, model__kernel=linear;, score=(train=0.996, test=0.995) total time= 3.1min
- [CV 4/10; 1/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.001, model__kernel=linear
- [CV 4/10; 1/10] END model__C=100, model__class_weight=balanced, model__gamma =0.001, model__kernel=linear;, score=(train=0.998, test=0.998) total time= 2 .4min
- [CV 10/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__ga mma=auto, model kernel=poly
- [CV 10/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamm a=auto, model__kernel=poly;, score=(train=0.064, test=0.064) total time=23.6 min

```
[CV 10/10; 7/10] START model__C=100, model__class_weight=balanced, model__ga mma=0.01, model kernel=rbf
```

- [CV 10/10; 7/10] END model__C=100, model__class_weight=balanced, model__gamm a=0.01, model__kernel=rbf;, score=(train=0.998, test=0.997) total time= 2.9m in
- [CV 2/10; 9/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.1, model kernel=rbf
- [CV 2/10; 9/10] END model__C=100, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.998, test=0.997) total time= 3.7min [CV 10/10; 9/10] START model__C=100, model__class_weight=balanced, model__ga mma=0.1, model__kernel=rbf
- [CV 10/10; 9/10] END model__C=100, model__class_weight=balanced, model__gamm a=0.1, model__kernel=rbf;, score=(train=0.998, test=0.998) total time= 3.7mi n
- [CV 10/10; 10/10] START model__C=0.1, model__class_weight=balanced, model__g amma=scale, model kernel=linear
- [CV 10/10; 10/10] END model__C=0.1, model__class_weight=balanced, model__gam ma=scale, model__kernel=linear;, score=(train=0.996, test=0.996) total time= 3.0min
- poly: 0.5748320097739767
- [CV 2/10; 1/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.001, model__kernel=linear
- [CV 2/10; 1/10] END model__C=100, model__class_weight=balanced, model__gamma =0.001, model__kernel=linear;, score=(train=0.998, test=0.997) total time= 2 .3min
- [CV 9/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=auto, model__kernel=poly
- [CV 9/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamma =auto, model__kernel=poly;, score=(train=0.064, test=0.064) total time=23.7m in
- [CV 1/10; 8/10] START model__C=100, model__class_weight=balanced, model__gam
 ma=0.0001, model kernel=poly
- [CV 1/10; 8/10] END model__C=100, model__class_weight=balanced, model__gamma =0.0001, model__kernel=poly;, score=(train=0.064, test=0.064) total time=13.
- [CV 5/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=auto, model__kernel=poly
- [CV 5/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamma =auto, model__kernel=poly;, score=(train=0.064, test=0.064) total time=23.6m in
- [CV 5/10; 6/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=0.1, model kernel=rbf
- [CV 5/10; 6/10] END model__C=0.1, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.996, test=0.996) total time= 2.0min [CV 2/10; 8/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.0001, model kernel=poly
- [CV 2/10; 8/10] END model__C=100, model__class_weight=balanced, model__gamma =0.0001, model__kernel=poly;, score=(train=0.064, test=0.064) total time=13.1min
- [CV 7/10; 1/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.001, model__kernel=linear
- [CV 7/10; 1/10] END model__C=100, model__class_weight=balanced, model__gamma =0.001, model__kernel=linear;, score=(train=0.998, test=0.998) total time= 2 .3min
- [CV 1/10; 3/10] START model__C=1, model__class_weight=balanced, model__gamma =0.1, model__kernel=poly
- [CV 1/10; 3/10] END model__C=1, model__class_weight=balanced, model__gamma=0

```
.1, model__kernel=poly;, score=(train=0.996, test=0.997) total time= 2.9min
[CV 7/10; 3/10] START model__C=1, model__class_weight=balanced, model__gamma
=0.1, model__kernel=poly
```

- [CV 7/10; 3/10] END model__C=1, model__class_weight=balanced, model__gamma=0.1, model__kernel=poly;, score=(train=0.996, test=0.997) total time= 2.7min
- [CV 5/10; 4/10] START model__C=10, model__class_weight=balanced, model__gamm a=1, model kernel=linear
- [CV 5/10; 4/10] END model__C=10, model__class_weight=balanced, model__gamma=1, model__kernel=linear;, score=(train=0.998, test=0.997) total time=12.6min [CV 1/10; 5/10] START model__C=10, model__class_weight=balanced, model__gamm a=0.0001, model__kernel=linear
- [CV 1/10; 5/10] END model__C=10, model__class_weight=balanced, model__gamma= 0.0001, model__kernel=linear;, score=(train=0.998, test=0.998) total time= 1.9min
- [CV 7/10; 5/10] START model__C=10, model__class_weight=balanced, model__gamm a=0.0001, model kernel=linear
- [CV 7/10; 5/10] END model__C=10, model__class_weight=balanced, model__gamma= 0.0001, model__kernel=linear;, score=(train=0.998, test=0.998) total time= 1 .3min
- [CV 9/10; 6/10] START model__C=0.1, model__class_weight=balanced, model__gam
 ma=0.1, model__kernel=rbf
- [CV 9/10; 6/10] END model__C=0.1, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.996, test=0.995) total time= 1.7min [CV 3/10; 8/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.0001, model__kernel=poly
- [CV 3/10; 8/10] END model__C=100, model__class_weight=balanced, model__gamma =0.0001, model__kernel=poly;, score=(train=0.064, test=0.065) total time=13. 2min
- [CV 1/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=auto, model kernel=poly
- [CV 1/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamma =auto, model__kernel=poly;, score=(train=0.085, test=0.085) total time=23.8m in
- [CV 8/10; 6/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=0.1, model__kernel=rbf
- [CV 8/10; 6/10] END model__C=0.1, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.996, test=0.996) total time= 2.1min [CV 4/10; 8/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.0001, model__kernel=poly
- [CV 4/10; 8/10] END model__C=100, model__class_weight=balanced, model__gamma =0.0001, model__kernel=poly;, score=(train=0.420, test=0.420) total time=13.2min
- [CV 4/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=auto, model__kernel=poly
- [CV 4/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamma =auto, model__kernel=poly;, score=(train=0.064, test=0.064) total time=23.6m in
- [CV 6/10; 6/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=0.1, model__kernel=rbf
- [CV 6/10; 6/10] END model__C=0.1, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.996, test=0.996) total time= 2.1min [CV 5/10; 8/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.0001, model_ kernel=poly
- [CV 5/10; 8/10] END model__C=100, model__class_weight=balanced, model__gamma =0.0001, model__kernel=poly;, score=(train=0.420, test=0.420) total time=13.2min
- [CV 5/10; 1/10] START model__C=100, model__class_weight=balanced, model__gam

```
ma=0.001, model__kernel=linear
[CV 5/10; 1/10] END model C=100, model class weight=balanced, model gamma
=0.001, model kernel=linear;, score=(train=0.998, test=0.997) total time= 2
.6min
[CV 5/10; 3/10] START model C=1, model class_weight=balanced, model gamma
=0.1, model kernel=poly
[CV 5/10; 3/10] END model C=1, model class weight=balanced, model gamma=0
.1, model _kernel=poly;, score=(train=0.996, test=0.996) total time= 3.2min
[CV 1/10; 4/10] START model C=10, model class weight=balanced, model gamm
a=1, model__kernel=linear
[CV 1/10; 4/10] END model__C=10, model__class_weight=balanced, model__gamma=
1, model kernel=linear;, score=(train=0.998, test=0.998) total time= 1.5min
[CV 3/10; 4/10] START model__C=10, model__class_weight=balanced, model__gamm
a=1, model kernel=linear
[CV 3/10; 4/10] END model C=10, model class weight=balanced, model gamma=
1, model kernel=linear;, score=(train=0.998, test=0.997) total time= 1.6min
[CV 6/10; 4/10] START model C=10, model class weight=balanced, model gamm
a=1, model kernel=linear
[CV 6/10; 4/10] END model C=10, model class weight=balanced, model gamma=
1, model kernel=linear;, score=(train=0.999, test=0.997) total time=12.4min
[CV 2/10; 5/10] START model C=10, model class weight=balanced, model gamm
a=0.0001, model kernel=linear
[CV 2/10; 5/10] END model C=10, model class weight=balanced, model gamma=
0.0001, model_kernel=linear;, score=(train=0.998, test=0.997) total time= 2
[CV 8/10; 5/10] START model C=10, model class weight=balanced, model gamm
a=0.0001, model kernel=linear
[CV 8/10; 5/10] END model C=10, model class weight=balanced, model gamma=
0.0001, model kernel=linear;, score=(train=0.998, test=0.997) total time= 1
[CV 10/10; 6/10] START model C=0.1, model class weight=balanced, model ga
mma=0.1, model__kernel=rbf
[CV 10/10; 6/10] END model C=0.1, model class weight=balanced, model gamm
a=0.1, model kernel=rbf;, score=(train=0.996, test=0.996) total time= 1.7mi
[CV 6/10; 8/10] START model C=100, model class weight=balanced, model gam
ma=0.0001, model__kernel=poly
[CV 6/10; 8/10] END model C=100, model class weight=balanced, model gamma
```

=0.0001, model_kernel=poly;, score=(train=0.420, test=0.420) total time=13.

[CV 9/10; 1/10] START model C=100, model class weight=balanced, model gam

[CV 9/10; 1/10] END model__C=100, model__class_weight=balanced, model__gamma =0.001, model kernel=linear;, score=(train=0.998, test=0.996) total time= 2

[CV 4/10; 3/10] START model C=1, model class weight=balanced, model gamma

[CV 4/10; 3/10] END model__C=1, model__class_weight=balanced, model__gamma=0
.1, model__kernel=poly;, score=(train=0.996, test=0.997) total time= 3.0min
[CV 10/10; 3/10] START model__C=1, model__class_weight=balanced, model__gamm

[CV 10/10; 3/10] END model__C=1, model__class_weight=balanced, model__gamma= 0.1, model__kernel=poly;, score=(train=0.996, test=0.996) total time= 3.3min [CV 8/10; 4/10] START model__C=10, model__class_weight=balanced, model__gamm

[CV 8/10; 4/10] END model__C=10, model__class_weight=balanced, model__gamma= 1, model__kernel=linear;, score=(train=0.998, test=0.997) total time= 7.0min

ma=0.001, model kernel=linear

=0.1, model kernel=poly

a=0.1, model_kernel=poly

a=1, model kernel=linear

.5min

```
[CV 6/10; 5/10] START model C=10, model class weight=balanced, model gamm
a=0.0001, model kernel=linear
[CV 6/10; 5/10] END model C=10, model class weight=balanced, model gamma=
0.0001, model_kernel=linear;, score=(train=0.999, test=0.997) total time= 2
[CV 10/10; 5/10] START model C=10, model class weight=balanced, model gam
ma=0.0001, model kernel=linear
[CV 10/10; 5/10] END model__C=10, model__class_weight=balanced, model__gamma
=0.0001, model kernel=linear;, score=(train=0.998, test=0.998) total time=
1.5min
[CV 1/10; 7/10] START model__C=100, model__class_weight=balanced, model__gam
ma=0.01, model kernel=rbf
[CV 1/10; 7/10] END model__C=100, model__class_weight=balanced, model__gamma
=0.01, model kernel=rbf;, score=(train=0.998, test=0.998) total time= 2.0mi
[CV 7/10; 8/10] START model C=100, model class weight=balanced, model gam
ma=0.0001, model__kernel=poly
[CV 7/10; 8/10] END model C=100, model class weight=balanced, model gamma
=0.0001, model kernel=poly;, score=(train=0.420, test=0.420) total time=14.
[CV 6/10; 1/10] START model C=100, model class weight=balanced, model gam
ma=0.001, model kernel=linear
[CV 6/10; 1/10] END model C=100, model class weight=balanced, model gamma
=0.001, model kernel=linear;, score=(train=0.999, test=0.997) total time= 2
[CV 2/10; 3/10] START model C=1, model class weight=balanced, model gamma
=0.1, model kernel=poly
[CV 2/10; 3/10] END model C=1, model__class_weight=balanced, model__gamma=0
```

.1, model__kernel=poly;, score=(train=0.996, test=0.996) total time= 2.9min [CV 8/10; 3/10] START model C=1, model class weight=balanced, model gamma

[CV 8/10; 3/10] END model__C=1, model__class_weight=balanced, model__gamma=0.1, model__kernel=poly;, score=(train=0.996, test=0.996) total time= 3.3min [CV 7/10; 4/10] START model C=10, model class weight=balanced, model gamm

[CV 7/10; 4/10] END model__C=10, model__class_weight=balanced, model__gamma=
1, model__kernel=linear;, score=(train=0.998, test=0.998) total time=12.4min
[CV 3/10; 5/10] START model__C=10, model__class_weight=balanced, model__gamm

[CV 3/10; 5/10] END model__C=10, model__class_weight=balanced, model__gamma= 0.0001, model kernel=linear;, score=(train=0.998, test=0.997) total time= 2

[CV 9/10; 5/10] START model C=10, model class weight=balanced, model gamm

[CV 9/10; 5/10] END model__C=10, model__class_weight=balanced, model__gamma= 0.0001, model kernel=linear;, score=(train=0.998, test=0.996) total time= 1

[CV 2/10; 7/10] START model C=100, model class weight=balanced, model gam

[CV 2/10; 7/10] END model__C=100, model__class_weight=balanced, model__gamma =0.01, model__kernel=rbf;, score=(train=0.998, test=0.997) total time= 2.0mi

[CV 8/10; 8/10] START model C=100, model class weight=balanced, model gam

[CV 8/10; 8/10] END model__C=100, model__class_weight=balanced, model__gamma =0.0001, model kernel=poly;, score=(train=0.420, test=0.420) total time=14.

=0.1, model kernel=poly

a=1, model kernel=linear

.1min

.7min

7min

a=0.0001, model__kernel=linear

a=0.0001, model__kernel=linear

ma=0.01, model kernel=rbf

ma=0.0001, model__kernel=poly

```
[CV 10/10; 1/10] START model__C=100, model__class_weight=balanced, model__ga
mma=0.001, model__kernel=linear
```

- [CV 10/10; 1/10] END model__C=100, model__class_weight=balanced, model__gamm a=0.001, model__kernel=linear;, score=(train=0.998, test=0.998) total time= 2.1min
- [CV 7/10; 2/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=auto, model kernel=poly
- [CV 7/10; 2/10] END model__C=0.1, model__class_weight=balanced, model__gamma =auto, model__kernel=poly;, score=(train=0.064, test=0.064) total time=23.3m in
- [CV 4/10; 7/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.01, model kernel=rbf
- [CV 4/10; 7/10] END model__C=100, model__class_weight=balanced, model__gamma =0.01, model__kernel=rbf;, score=(train=0.998, test=0.998) total time= 2.5mi n
- [CV 9/10; 8/10] START model__C=100, model__class_weight=balanced, model__gam
 ma=0.0001, model__kernel=poly
- [CV 9/10; 8/10] END model__C=100, model__class_weight=balanced, model__gamma =0.0001, model__kernel=poly;, score=(train=0.420, test=0.420) total time=14.
- [CV 8/10; 1/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.001, model__kernel=linear
- [CV 8/10; 1/10] END model__C=100, model__class_weight=balanced, model__gamma =0.001, model__kernel=linear;, score=(train=0.998, test=0.997) total time= 2 .5min
- [CV 3/10; 3/10] START model__C=1, model__class_weight=balanced, model__gamma =0.1, model__kernel=poly
- [CV 3/10; 3/10] END model__C=1, model__class_weight=balanced, model__gamma=0
- .1, model__kernel=poly;, score=(train=0.996, test=0.995) total time= 3.0min [CV 9/10; 3/10] START model__C=1, model__class_weight=balanced, model__gamma
- [CV 9/10; 3/10] START model__C=1, model__class_weight=balanced, model__gamma =0.1, model__kernel=poly
- [CV 9/10; 3/10] END model__C=1, model__class_weight=balanced, model__gamma=0 .1, model kernel=poly;, score=(train=0.996, test=0.995) total time= 3.4min
- [CV 10/10; 4/10] START model__C=10, model__class_weight=balanced, model__gam
 ma=1, model__kernel=linear
- [CV 10/10; 4/10] END model__C=10, model__class_weight=balanced, model__gamma =1, model__kernel=linear;, score=(train=0.998, test=0.998) total time= 6.7mi n
- [CV 4/10; 5/10] START model__C=10, model__class_weight=balanced, model__gamm a=0.0001, model kernel=linear
- [CV 4/10; 5/10] END model__C=10, model__class_weight=balanced, model__gamma= 0.0001, model__kernel=linear;, score=(train=0.998, test=0.998) total time= 2 .3min
- [CV 2/10; 6/10] START model__C=0.1, model__class_weight=balanced, model__gam ma=0.1, model kernel=rbf
- [CV 2/10; 6/10] END model__C=0.1, model__class_weight=balanced, model__gamma =0.1, model__kernel=rbf;, score=(train=0.996, test=0.996) total time= 1.8min [CV 3/10; 7/10] START model__C=100, model__class_weight=balanced, model__gam ma=0.01, model kernel=rbf
- [CV 3/10; 7/10] END model__C=100, model__class_weight=balanced, model__gamma =0.01, model__kernel=rbf;, score=(train=0.998, test=0.997) total time= 2.6mi n
- [CV 10/10; 8/10] START model__C=100, model__class_weight=balanced, model__ga mma=0.0001, model__kernel=poly
- [CV 10/10; 8/10] END model__C=100, model__class_weight=balanced, model__gamm a=0.0001, model__kernel=poly;, score=(train=0.420, test=0.420) total time=13.9min

```
rbf: 0.9186774587660355
```

SGD Classifier

```
In [30]: def plot_svc_decision_function(model, ax=None, plot_support=True):
              """Plot the decision function for a 2D SVC"""
             if ax is None:
                  ax = plt.gca()
             xlim = ax.get xlim()
             ylim = ax.get ylim()
             # create grid to evaluate model
             x = np.linspace(xlim[0], xlim[1], 30)
             y = np.linspace(ylim[0], ylim[1], 30)
             Y, X = np.meshgrid(y, x)
             xy = np.vstack([X.ravel(), Y.ravel()]).T
             P = model.decision_function(xy).reshape(X.shape)
             # plot decision boundary and margins
             ax.contour(X, Y, P, colors='k',
                         levels=[-1, 0, 1], alpha=0.5,
                         linestyles=['--', '-', '--'])
             # plot support vectors
             if plot support:
                  ax.scatter(model.support vectors [:, 0],
                             model.support_vectors_[:, 1],
                             s=300, linewidth=1, facecolors='none');
             ax.set_xlim(xlim)
             ax.set_ylim(ylim)
```

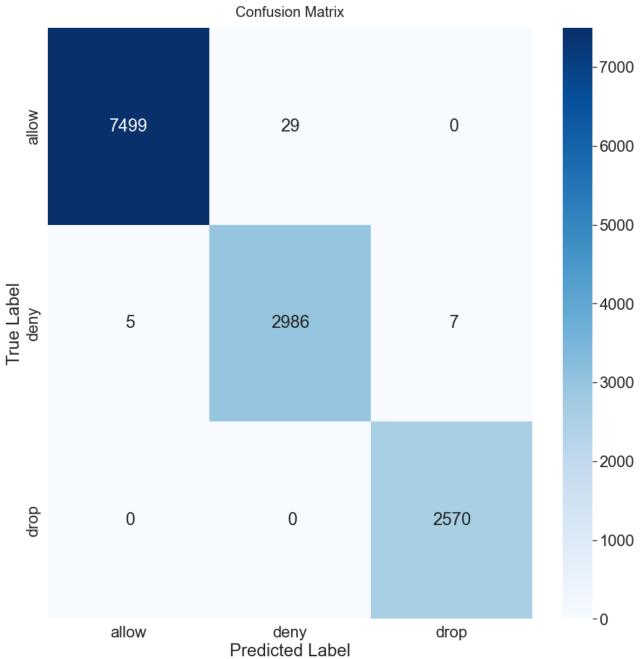
```
In [31]: params sgd = [
              # optimal learning rate
                  "model learning rate": ["optimal"],
                  "model alpha": [0.000001, 0.00001, 0.0001, 0.001, 0.01],
                  "model _eta0": [0],
                  "model__class_weight":["balanced"]
              },
              # constant learning rate (eta0 cannot be 0 for this)
              "model__learning_rate": ["constant"],
              "model__alpha": [0.000001, 0.00001, 0.0001, 0.001, 0.01],
              "model__eta0": [0.01, 0.1, 1.0, 5, 10],
              "model__class_weight":["balanced"]
              }
          ]
In [32]: | %%time
          # define preprocessing for columns
          t = [('cat', OneHotEncoder(handle_unknown='ignore'), df_ohe),
               ('scale', MinMaxScaler(), df numeric)]
          col_transform = ColumnTransformer(t)
          model_sgd = SGDClassifier(loss = 'log_loss')
          pipeline_sgd = Pipeline(steps=[('preprocessing',col_transform), ('model', model', model')
          #Fit
          clf sgd = GridSearchCV(pipeline sgd,
                                 params sgd,
                                 scoring = 'f1 weighted',
                                 verbose = 15,
                                 cv = cv
                                 n jobs = -1,
                                 return_train_score = True)
          search_sgd = clf_sgd.fit(X_train, y_train)
         Fitting 10 folds for each of 30 candidates, totalling 300 fits
         CPU times: user 3.13 s, sys: 1.49 s, total: 4.62 s
         Wall time: 26.4 s
In [33]: df_search_sgd_cv_results_columns = ['param_model__alpha',
                                               'param model eta0',
                                               'param model learning rate',
                                               'mean train score',
                                               'mean test score',
                                               'rank_test_score']
          df search sqd cv results = cv summary(search sqd, df search sqd cv results d
          df search sqd cv results
```

:	param_combination	param_modelalpha	param_modeleta0	param_modellearning_
	1	0.000001	0	ob.
•	2	0.00001	0	op.
2	2 3	0.0001	0	ob.
3	4	0.001	0	ob.
4	5	0.01	0	ob.
5	6	0.000001	0.01	cons
6	7	0.000001	0.1	cons
7	8	0.000001	1.0	cons
8	9	0.000001	5	cons
ę	10	0.000001	10	cons
10	11	0.00001	0.01	cons
1′	12	0.00001	0.1	cons
12	2 13	0.00001	1.0	cons
13	14	0.00001	5	cons
14	15	0.00001	10	cons
15	16	0.0001	0.01	cons
16	5 17	0.0001	0.1	cons
17	18	0.0001	1.0	cons
18	19	0.0001	5	cons
19	20	0.0001	10	cons
20	21	0.001	0.01	cons
2	22	0.001	0.1	cons
22	2 23	0.001	1.0	cons
23	24	0.001	5	cons
24	25	0.001	10	cons
25	5 26	0.01	0.01	cons
26	3 27	0.01	0.1	cons
27	28	0.01	1.0	cons
28	29	0.01	5	cons
29	30	0.01	10	cons

Out[33]:

In [34]: predict_sgd = search_sgd.predict(X_test)
 get_classification_report(X_train, y_train, X_test, y_test, predict_sgd, sea

Training Score: 0.997826 Test result: precision recall f1-score support allow 0.999334 0.996148 0.997738 7528 deny 0.990381 0.995997 0.993181 2998 drop 0.997284 1.000000 0.998640 2570 accuracy 0.996869 13096 macro avg 0.995666 0.997382 0.996520 13096 weighted avg 0.996882 0.996869 0.996872 13096 Confusion Matrix



Collapsing to Two Classes

```
In [35]: | df['target'] = df['Action'].map({'allow': "Allow",
                                             'deny': "Deny",
                                             'drop': "Deny",
                                             'reset-both': "Deny"})
          df["target"].value counts(normalize = False)
         Allow
                   37640
Out[35]:
         Deny
                   27838
         Name: target, dtype: int64
          df.drop(columns = "Action", inplace = True)
In [36]:
          df.head()
Out[36]:
                                 NAT
                                            NAT
                                                                               Elapsed
            Source Destination
                                                       Bytes
                                                                Bytes
                               Source Destination
                                                 Bytes
                                                                       Packets
                                                                                 Time pkt
               Port
                          Port
                                                        Sent Received
                                 Port
                                            Port
                                                                                 (sec)
          0
             57222
                           53
                               54587
                                             53
                                                   177
                                                          94
                                                                   83
                                                                            2
                                                                                   30
             56258
                         3389
                               56258
                                           3389
                                                  4768
                                                        1600
                                                                 3168
                                                                           19
                                                                                   17
          2
              6881
                        50321
                               43265
                                          50321
                                                   238
                                                         118
                                                                  120
                                                                            2
                                                                                 1199
          3
             50553
                         3389
                               50553
                                           3389
                                                  3327
                                                        1438
                                                                 1889
                                                                           15
                                                                                   17
          4
             50002
                          443
                               45848
                                            443 25358
                                                        6778
                                                                18580
                                                                           31
                                                                                   16
In [37]: X = df.loc[:, df.columns != 'target'].values
          y = df['target'].values
          X train, X test, y train, y test = train test split(
              X, y, test size=0.2, random state = 42)
In [38]:
         param grid= {'kernel': ('linear', 'rbf'),
                       'C': [0.001, 0.01, 0.1, 1, 10, 100]}
          base estimator = SVC(gamma = 'scale')
          grid_search = GridSearchCV(base_estimator,
                                      param_grid = param_grid,
                                      n \text{ jobs} = -1)
          grid_search.fit(X_train, y_train)
          [CV 2/10; 1/30] START model_alpha=1e-06, model_class_weight=balanced, mode
          l__eta0=0, model__learning_rate=optimal
          [CV 2/10; 1/30] END model_alpha=1e-06, model_class_weight=balanced, model_
          _eta0=0, model__learning_rate=optimal;, score=(train=0.998, test=0.997) tota
         1 time=
          [CV 7/10; 3/30] START model alpha=0.0001, model class weight=balanced, mod
         el eta0=0, model learning rate=optimal
         [CV 7/10; 3/30] END model alpha=0.0001, model class weight=balanced, model
          eta0=0, model learning rate=optimal;, score=(train=0.996, test=0.997) tot
         al time=
          [CV 3/10; 5/30] START model alpha=0.01, model class weight=balanced, model
           eta0=0, model learning rate=optimal
         [CV 3/10; 5/30] END model__alpha=0.01, model__class_weight=balanced, model_
         eta0=0, model_learning_rate=optimal;, score=(train=0.991, test=0.990) total
```

```
time=
       0.7s
[CV 10/10; 6/30] START model alpha=1e-06, model class weight=balanced, mod
el eta0=0.01, model learning rate=constant
[CV 10/10; 6/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.996)
total time=
             1.0s
[CV 4/10; 8/30] START model_alpha=1e-06, model_class_weight=balanced, mode
l__eta0=1.0, model__learning_rate=constant
[CV 4/10; 8/30] END model alpha=1e-06, model class weight=balanced, model
_eta0=1.0, model__learning_rate=constant;, score=(train=0.998, test=0.998) t
otal time=
             0.8s
[CV 2/10; 10/30] START model alpha=1e-06, model class weight=balanced, mod
   _eta0=10, model__learning_rate=constant
[CV 2/10; 10/30] END model alpha=1e-06, model class weight=balanced, model
eta0=10, model learning rate=constant;, score=(train=0.997, test=0.997) t
otal time=
             1.1s
[CV 10/10; 11/30] START model__alpha=1e-05, model__class_weight=balanced, mo
del__eta0=0.01, model__learning_rate=constant
[CV 10/10; 11/30] END model_alpha=1e-05, model_class_weight=balanced, mode
1__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996
) total time=
               1.0s
[CV 8/10; 13/30] START model_alpha=1e-05, model_class_weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 8/10; 13/30] END model alpha=1e-05, model class weight=balanced, model
eta0=1.0, model_learning_rate=constant;, score=(train=0.996, test=0.997)
total time=
             0.7s
[CV 2/10; 15/30] START model_alpha=1e-05, model class weight=balanced, mod
el eta0=10, model__learning_rate=constant
[CV 2/10; 15/30] END model alpha=1e-05, model class weight=balanced, model
eta0=10, model learning rate=constant;, score=(train=0.995, test=0.995) t
otal time=
            0.8s
[CV 8/10; 16/30] START model__alpha=0.0001, model__class_weight=balanced, mo
del eta0=0.01, model learning rate=constant
[CV 8/10; 16/30] END model alpha=0.0001, model class weight=balanced, mode
1__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996
) total time=
               0.9s
[CV 5/10; 18/30] START model_alpha=0.0001, model_class_weight=balanced, mo
del__eta0=1.0, model__learning_rate=constant
[CV 5/10; 18/30] END model_alpha=0.0001, model_class_weight=balanced, mode
1 eta0=1.0, model__learning_rate=constant;, score=(train=0.992, test=0.990)
             0.7s
total time=
[CV 1/10; 20/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=10, model learning rate=constant
[CV 1/10; 20/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1__eta0=10, model__learning_rate=constant;, score=(train=0.996, test=0.997)
total time=
             0.9s
[CV 7/10; 21/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=0.01, model__learning_rate=constant
[CV 7/10; 21/30] END model_alpha=0.001, model_class_weight=balanced, model
 eta0=0.01, model__learning_rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.8s
[CV 3/10; 23/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 3/10; 23/30] END model _alpha=0.001, model _class_weight=balanced, model
eta0=1.0, model learning rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.7s
```

[CV 9/10; 24/30] START model__alpha=0.001, model__class_weight=balanced, mod

```
el eta0=5, model learning rate=constant
[CV 9/10; 24/30] END model alpha=0.001, model class weight=balanced, model
 eta0=5, model learning rate=constant;, score=(train=0.991, test=0.987) to
tal time=
           1.0s
[CV 7/10; 26/30] START model_alpha=0.01, model_class_weight=balanced, mode
l__eta0=0.01, model__learning_rate=constant
[CV 7/10; 26/30] END model alpha=0.01, model class weight=balanced, model
eta0=0.01, model_learning_rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.8s
[CV 2/10; 28/30] START model_alpha=0.01, model_class_weight=balanced, mode
l__eta0=1.0, model__learning_rate=constant
[CV 2/10; 28/30] END model alpha=0.01, model class weight=balanced, model
eta0=1.0, model learning rate=constant;, score=(train=0.990, test=0.992) t
otal time=
            0.8s
[CV 7/10; 29/30] START model alpha=0.01, model class weight=balanced, mode
1 eta0=5, model learning rate=constant
[CV 7/10; 29/30] END model__alpha=0.01, model__class_weight=balanced, model_
eta0=5, model learning rate=constant;, score=(train=0.728, test=0.730) tot
al time=
          1.0s
[CV 10/10; 30/30] START model_alpha=0.01, model_class_weight=balanced, mod
el eta0=10, model learning rate=constant
[CV 10/10; 30/30] END model alpha=0.01, model class weight=balanced, model
_eta0=10, model__learning_rate=constant;, score=(train=0.728, test=0.728) t
otal time=
            0.6s
[CV 2/10; 2/30] START model_alpha=1e-05, model_class_weight=balanced, mode
l eta0=0, model learning rate=optimal
[CV 2/10; 2/30] END model alpha=1e-05, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.997, test=0.997) tota
l time=
         1.0s
[CV 2/10; 3/30] START model alpha=0.0001, model class weight=balanced, mod
el eta0=0, model learning rate=optimal
[CV 2/10; 3/30] END model alpha=0.0001, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.996, test=0.996) tot
al time=
          0.7s
[CV 6/10; 4/30] START model_alpha=0.001, model_class_weight=balanced, mode
l__eta0=0, model__learning_rate=optimal
[CV 6/10; 4/30] END model__alpha=0.001, model__class_weight=balanced, model__
eta0=0, model learning rate=optimal;, score=(train=0.991, test=0.990) tota
l time=
         0.7s
[CV 3/10; 6/30] START model alpha=1e-06, model class weight=balanced, mode
```

- [CV 3/10; 6/30] START model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant
- [CV 3/10; 6/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.995) total time= 0.9s
- [CV 9/10; 7/30] START model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant
- [CV 9/10; 7/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant;, score=(train=0.997, test=0.995) t otal time= 0.7s
- [CV 6/10; 9/30] START model__alpha=1e-06, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant
- [CV 6/10; 9/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.998, test=0.996) tot al time= 1.1s
- [CV 2/10; 11/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=0.01, model__learning_rate=constant
- [CV 2/10; 11/30] END model__alpha=1e-05, model__class_weight=balanced, model

```
eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996)
             0.9s
total time=
[CV 8/10; 12/30] START model alpha=1e-05, model class weight=balanced, mod
el__eta0=0.1, model__learning_rate=constant
[CV 8/10; 12/30] END model_alpha=1e-05, model_class_weight=balanced, model
eta0=0.1, model__learning_rate=constant;, score=(train=0.997, test=0.996)
[CV 4/10; 14/30] START model_alpha=1e-05, model_class_weight=balanced, mod
el eta0=5, model learning rate=constant
[CV 4/10; 14/30] END model alpha=1e-05, model class weight=balanced, model
__eta0=5, model__learning_rate=constant;, score=(train=0.997, test=0.998) to
tal time=
           0.7s
[CV 10/10; 15/30] START model alpha=1e-05, model class weight=balanced, mo
del eta0=10, model learning rate=constant
[CV 10/10; 15/30] END model alpha=1e-05, model class weight=balanced, mode
1 eta0=10, model learning rate=constant;, score=(train=0.992, test=0.992)
total time=
             0.7s
[CV 6/10; 17/30] START model_alpha=0.0001, model_class_weight=balanced, mo
del eta0=0.1, model learning rate=constant
[CV 6/10; 17/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=0.1, model learning rate=constant;, score=(train=0.996, test=0.996)
total time=
             0.5s
[CV 2/10; 19/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=5, model learning rate=constant
[CV 2/10; 19/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=5, model learning rate=constant;, score=(train=0.995, test=0.995) t
otal time=
            0.5s
[CV 8/10; 20/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=10, model learning rate=constant
[CV 8/10; 20/30] END model alpha=0.0001, model class weight=balanced, mode
1 eta0=10, model learning rate=constant;, score=(train=0.996, test=0.994)
total time=
             0.9s
[CV 3/10; 22/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=0.1, model__learning_rate=constant
[CV 3/10; 22/30] END model alpha=0.001, model class weight=balanced, model
eta0=0.1, model learning rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.6s
[CV 10/10; 23/30] START model__alpha=0.001, model__class_weight=balanced, mo
del__eta0=1.0, model__learning_rate=constant
[CV 10/10; 23/30] END model alpha=0.001, model class weight=balanced, mode
1 eta0=1.0, model learning rate=constant;, score=(train=0.991, test=0.990)
total time=
              0.6s
[CV 6/10; 25/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=10, model__learning_rate=constant
[CV 6/10; 25/30] END model alpha=0.001, model class weight=balanced, model
 eta0=10, model learning rate=constant;, score=(train=0.728, test=0.728) t
otal time=
            1.0s
[CV 2/10; 27/30] START model alpha=0.01, model class weight=balanced, mode
l eta0=0.1, model__learning_rate=constant
[CV 2/10; 27/30] END model alpha=0.01, model class weight=balanced, model
_eta0=0.1, model__learning_rate=constant;, score=(train=0.990, test=0.992) t
otal time=
            0.7s
[CV 8/10; 28/30] START model_alpha=0.01, model_class_weight=balanced, mode
l__eta0=1.0, model__learning_rate=constant
[CV 8/10; 28/30] END model_alpha=0.01, model_class_weight=balanced, model_
eta0=1.0, model learning rate=constant;, score=(train=0.990, test=0.992) t
otal time=
            0.7s
```

```
[CV 4/10; 30/30] START model__alpha=0.01, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant
```

- [CV 4/10; 30/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.420, test=0.420) to tal time= 1.2s
- [CV 4/10; 2/30] START model__alpha=1e-05, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal
- [CV 4/10; 2/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.997, test=0.998) tota l time= 0.9s
- [CV 7/10; 2/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.997, test=0.998) tota l time= 0.7s
- [CV 4/10; 4/30] START model__alpha=0.001, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal
- [CV 4/10; 4/30] END model__alpha=0.001, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.991, test=0.991) tota l time= 0.6s
- [CV 9/10; 5/30] START model__alpha=0.01, model__class_weight=balanced, model __eta0=0, model__learning_rate=optimal
- [CV 9/10; 5/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.991, test=0.987) total time= 0.6s
- [CV 6/10; 7/30] START model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant
- [CV 6/10; 7/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant;, score=(train=0.997, test=0.996) t otal time= 0.5s
- [CV 1/10; 9/30] START model__alpha=1e-06, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant
- [CV 1/10; 9/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.998, test=0.997) tot al time= 0.8s
- [CV 7/10; 10/30] START model__alpha=1e-06, model__class_weight=balanced, mod el__eta0=10, model__learning_rate=constant
- [CV 7/10; 10/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.997, test=0.997) t otal time= 0.9s
- [CV 3/10; 12/30] START model__alpha=1e-05, model__class_weight=balanced, mod el eta0=0.1, model learning rate=constant
- [CV 3/10; 12/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant;, score=(train=0.997, test=0.995) total time= 0.7s
- [CV 9/10; 13/30] START model__alpha=1e-05, model__class_weight=balanced, mod el eta0=1.0, model learning rate=constant
- [CV 9/10; 13/30] END model__alpha=1e-05, model__class_weight=balanced, model __eta0=1.0, model__learning_rate=constant;, score=(train=0.997, test=0.997) total time= 0.7s
- [CV 5/10; 15/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=10, model__learning_rate=constant
- [CV 5/10; 15/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.996, test=0.997) t otal time= 0.8s
- [CV 1/10; 17/30] START model__alpha=0.0001, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant

```
[CV 1/10; 17/30] END model alpha=0.0001, model class weight=balanced, mode
1__eta0=0.1, model__learning_rate=constant;, score=(train=0.996, test=0.997)
total time=
             0.8s
[CV 3/10; 18/30] START model_alpha=0.0001, model_class_weight=balanced, mo
del__eta0=1.0, model__learning_rate=constant
[CV 3/10; 18/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1__eta0=1.0, model__learning_rate=constant;, score=(train=0.996, test=0.995)
total time=
             0.7s
[CV 9/10; 19/30] START model alpha=0.0001, model class weight=balanced, mo
del__eta0=5, model__learning_rate=constant
[CV 9/10; 19/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=5, model learning rate=constant;, score=(train=0.996, test=0.995) t
otal time=
           0.7s
[CV 6/10; 21/30] START model alpha=0.001, model class weight=balanced, mod
el eta0=0.01, model__learning_rate=constant
[CV 6/10; 21/30] END model alpha=0.001, model class weight=balanced, model
 _eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.8s
[CV 2/10; 23/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 2/10; 23/30] END model alpha=0.001, model class weight=balanced, model
 eta0=1.0, model_learning_rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.6s
[CV 6/10; 24/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=5, model__learning_rate=constant
[CV 6/10; 24/30] END model alpha=0.001, model class weight=balanced, model
 eta0=5, model learning rate=constant;, score=(train=0.991, test=0.990) to
tal time=
           0.8s
[CV 9/10; 25/30] START model alpha=0.001, model class weight=balanced, mod
el eta0=10, model learning rate=constant
[CV 9/10; 25/30] END model alpha=0.001, model class weight=balanced, model
__eta0=10, model__learning_rate=constant;, score=(train=0.991, test=0.987) t
otal time= 0.8s
[CV 5/10; 27/30] START model__alpha=0.01, model__class_weight=balanced, mode
l__eta0=0.1, model__learning_rate=constant
[CV 5/10; 27/30] END model_alpha=0.01, model_class_weight=balanced, model_
_eta0=0.1, model__learning_rate=constant;, score=(train=0.991, test=0.989) t
otal time= 0.8s
[CV 1/10; 29/30] START model alpha=0.01, model class weight=balanced, mode
1 eta0=5, model learning rate=constant
[CV 1/10; 29/30] END model alpha=0.01, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.740, test=0.742) tot
```

- al time= 0.9s
- [CV 7/10; 30/30] START model__alpha=0.01, model__class_weight=balanced, mode l__eta0=10, model__learning_rate=constant
- [CV 7/10; 30/30] END model alpha=0.01, model class weight=balanced, model eta0=10, model learning rate=constant;, score=(train=0.693, test=0.694) to tal time=
- [CV 1/10; 2/30] START model_alpha=1e-05, model_class_weight=balanced, mode l__eta0=0, model__learning_rate=optimal
- [CV 1/10; 2/30] END model__alpha=1e-05, model__class_weight=balanced, model_ eta0=0, model_learning_rate=optimal;, score=(train=0.998, test=0.997) tota l time=
- [CV 9/10; 2/30] START model_alpha=1e-05, model_class_weight=balanced, mode l__eta0=0, model__learning_rate=optimal
- [CV 9/10; 2/30] END model alpha=1e-05, model class weight=balanced, model _eta0=0, model__learning_rate=optimal;, score=(train=0.997, test=0.995) tota

```
l time=
         0.7s
[CV 7/10; 4/30] START model alpha=0.001, model class weight=balanced, mode
l eta0=0, model learning rate=optimal
[CV 7/10; 4/30] END model__alpha=0.001, model__class_weight=balanced, model__
eta0=0, model learning rate=optimal;, score=(train=0.990, test=0.992) tota
1 time=
         0.7s
[CV 4/10; 6/30] START model alpha=1e-06, model class weight=balanced, mode
l__eta0=0.01, model__learning_rate=constant
[CV 4/10; 6/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.997)
total time=
              0.9s
[CV 10/10; 7/30] START model alpha=1e-06, model class weight=balanced, mod
   _eta0=0.1, model__learning_rate=constant
[CV 10/10; 7/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0.1, model learning rate=constant;, score=(train=0.997, test=0.996)
              0.7s
total time=
[CV 5/10; 9/30] START model__alpha=1e-06, model__class_weight=balanced, mode
1 eta0=5, model__learning_rate=constant
[CV 5/10; 9/30] END model alpha=1e-06, model class weight=balanced, model
_eta0=5, model__learning_rate=constant;, score=(train=0.998, test=0.998) tot
al time=
          0.9s
[CV 1/10; 11/30] START model alpha=1e-05, model class weight=balanced, mod
el__eta0=0.01, model__learning_rate=constant
[CV 1/10; 11/30] END model alpha=1e-05, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.997)
total time=
             0.8s
[CV 7/10; 12/30] START model alpha=1e-05, model class weight=balanced, mod
el eta0=0.1, model learning rate=constant
[CV 7/10; 12/30] END model alpha=1e-05, model class weight=balanced, model
eta0=0.1, model learning rate=constant;, score=(train=0.996, test=0.997)
total time=
             0.6s
[CV 3/10; 14/30] START model__alpha=1e-05, model__class_weight=balanced, mod
el eta0=5, model learning rate=constant
[CV 3/10; 14/30] END model alpha=1e-05, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.992, test=0.991) to
tal time=
           0.6s
[CV 9/10; 15/30] START model_alpha=1e-05, model_class_weight=balanced, mod
el__eta0=10, model__learning_rate=constant
[CV 9/10; 15/30] END model_alpha=1e-05, model_class_weight=balanced, model
 eta0=10, model learning rate=constant;, score=(train=0.997, test=0.995) t
            0.6s
otal time=
[CV 2/10; 17/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=0.1, model learning rate=constant
[CV 2/10; 17/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1__eta0=0.1, model__learning_rate=constant;, score=(train=0.996, test=0.996)
total time=
             0.7s
```

- [CV 8/10; 18/30] START model__alpha=0.0001, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant
- [CV 8/10; 18/30] END model__alpha=0.0001, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant;, score=(train=0.996, test=0.996) total time= 0.7s
- [CV 5/10; 20/30] START model__alpha=0.0001, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant
- [CV 5/10; 20/30] END model__alpha=0.0001, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.992, test=0.990) total time= 0.9s
- [CV 1/10; 22/30] START model__alpha=0.001, model__class_weight=balanced, mod

```
el eta0=0.1, model learning rate=constant
[CV 1/10; 22/30] END model alpha=0.001, model class weight=balanced, model
 eta0=0.1, model learning rate=constant;, score=(train=0.990, test=0.992)
total time=
              0.7s
[CV 6/10; 23/30] START model alpha=0.001, model class_weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 6/10; 23/30] END model alpha=0.001, model class weight=balanced, model
_eta0=1.0, model__learning_rate=constant;, score=(train=0.991, test=0.990)
total time=
              0.7s
[CV 3/10; 25/30] START model_alpha=0.001, model_class_weight=balanced, mod
el__eta0=10, model__learning_rate=constant
[CV 3/10; 25/30] END model alpha=0.001, model class weight=balanced, model
_eta0=10, model__learning_rate=constant;, score=(train=0.991, test=0.990) t
otal time=
            0.9s
[CV 8/10; 26/30] START model _alpha=0.01, model__class_weight=balanced, mode
l eta0=0.01, model learning rate=constant
[CV 8/10; 26/30] END model__alpha=0.01, model__class_weight=balanced, model_
eta0=0.01, model learning rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.8s
[CV 3/10; 28/30] START model_alpha=0.01, model_class_weight=balanced, mode
l eta0=1.0, model learning rate=constant
[CV 3/10; 28/30] END model alpha=0.01, model class weight=balanced, model
eta0=1.0, model_learning_rate=constant;, score=(train=0.991, test=0.990) t
            0.8s
otal time=
[CV 9/10; 29/30] START model_alpha=0.01, model_class_weight=balanced, mode
1 eta0=5, model learning rate=constant
[CV 9/10; 29/30] END model alpha=0.01, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.693, test=0.691) tot
al time=
          1.0s
[CV 10/10; 1/30] START model alpha=1e-06, model class weight=balanced, mod
el eta0=0, model learning rate=optimal
[CV 10/10; 1/30] END model__alpha=1e-06, model__class_weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.998, test=0.997) tot
al time=
          1.3s
[CV 6/10; 3/30] START model alpha=0.0001, model class_weight=balanced, mod
el__eta0=0, model__learning_rate=optimal
[CV 6/10; 3/30] END model__alpha=0.0001, model__class_weight=balanced, model
__eta0=0, model__learning_rate=optimal;, score=(train=0.996, test=0.996) tot
al time=
          0.7s
[CV 1/10; 5/30] START model alpha=0.01, model class weight=balanced, model
 eta0=0, model learning rate=optimal
[CV 1/10; 5/30] END model alpha=0.01, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.990, test=0.992) total
time=
       0.7s
[CV 7/10; 6/30] START model__alpha=1e-06, model__class_weight=balanced, mode
  eta0=0.01, model learning rate=constant
[CV 7/10; 6/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.997)
total time=
              1.0s
[CV 8/10; 8/30] START model_alpha=1e-06, model_class_weight=balanced, mode
l__eta0=1.0, model__learning_rate=constant
[CV 8/10; 8/30] END model alpha=1e-06, model class weight=balanced, model
_eta0=1.0, model__learning_rate=constant;, score=(train=0.998, test=0.997) t
otal time=
            0.8s
[CV 1/10; 10/30] START model alpha=1e-06, model class weight=balanced, mod
el eta0=10, model learning rate=constant
[CV 1/10; 10/30] END model__alpha=1e-06, model__class_weight=balanced, model
```

```
eta0=10, model learning rate=constant;, score=(train=0.997, test=0.998) t
            1.2s
otal time=
[CV 2/10; 12/30] START model alpha=1e-05, model class weight=balanced, mod
el__eta0=0.1, model__learning_rate=constant
[CV 2/10; 12/30] END model _alpha=1e-05, model _class_weight=balanced, model
eta0=0.1, model learning rate=constant;, score=(train=0.997, test=0.996)
[CV 5/10; 13/30] START model_alpha=1e-05, model_class_weight=balanced, mod
el eta0=1.0, model learning rate=constant
[CV 5/10; 13/30] END model alpha=1e-05, model class weight=balanced, model
eta0=1.0, model learning rate=constant;, score=(train=0.997, test=0.997)
total time=
[CV 4/10; 15/30] START model alpha=1e-05, model class weight=balanced, mod
el eta0=10, model learning rate=constant
[CV 4/10; 15/30] END model alpha=1e-05, model class weight=balanced, model
eta0=10, model learning rate=constant;, score=(train=0.997, test=0.997) t
            0.8s
otal time=
[CV 9/10; 16/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=0.01, model learning rate=constant
[CV 9/10; 16/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.995
) total time=
               0.9s
[CV 6/10; 18/30] START model alpha=0.0001, model class weight=balanced, mo
del__eta0=1.0, model__learning_rate=constant
[CV 6/10; 18/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=1.0, model learning rate=constant;, score=(train=0.996, test=0.996)
total time=
             0.7s
[CV 3/10; 20/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=10, model learning rate=constant
[CV 3/10; 20/30] END model alpha=0.0001, model class weight=balanced, mode
1 eta0=10, model learning rate=constant;, score=(train=0.995, test=0.995)
total time=
              1.2s
[CV 4/10; 22/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=0.1, model__learning_rate=constant
[CV 4/10; 22/30] END model alpha=0.001, model class weight=balanced, model
 eta0=0.1, model learning rate=constant;, score=(train=0.991, test=0.991)
total time=
              0.6s
[CV 9/10; 23/30] START model_alpha=0.001, model_class_weight=balanced, mod
   _eta0=1.0, model__learning_rate=constant
[CV 9/10; 23/30] END model alpha=0.001, model class weight=balanced, model
 eta0=1.0, model learning rate=constant;, score=(train=0.991, test=0.987)
total time=
              0.6s
[CV 5/10; 25/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=10, model__learning_rate=constant
[CV 5/10; 25/30] END model alpha=0.001, model class weight=balanced, model
 eta0=10, model learning rate=constant;, score=(train=0.990, test=0.988) t
otal time=
            1.0s
[CV 1/10; 27/30] START model_alpha=0.01, model_class_weight=balanced, mode
l eta0=0.1, model__learning_rate=constant
[CV 1/10; 27/30] END model _alpha=0.01, model _class_weight=balanced, model_
_eta0=0.1, model__learning_rate=constant;, score=(train=0.990, test=0.992) t
otal time=
            0.8s
[CV 7/10; 28/30] START model_alpha=0.01, model_class_weight=balanced, mode
l__eta0=1.0, model__learning_rate=constant
[CV 7/10; 28/30] END model_alpha=0.01, model_class_weight=balanced, model_
eta0=1.0, model learning rate=constant;, score=(train=0.990, test=0.992) t
otal time=
            0.8s
```

```
[CV 3/10; 30/30] START model__alpha=0.01, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant
[CV 3/10; 30/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.728, test=0.728) to
```

- tal time= 1.2s
 [CV 6/10; 1/30] START model__alpha=1e-06, model__class_weight=balanced, mode
 l eta0=0, model learning rate=optimal
- [CV 6/10; 1/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.998, test=0.997) tota l time= 1.3s
- [CV 3/10; 3/30] START model__alpha=0.0001, model__class_weight=balanced, mod el eta0=0, model learning rate=optimal
- [CV 3/10; 3/30] END model__alpha=0.0001, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.996, test=0.995) tot al time= 0.7s
- [CV 9/10; 4/30] START model__alpha=0.001, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal
- [CV 9/10; 4/30] END model__alpha=0.001, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.991, test=0.987) tota l time= 0.7s
- [CV 5/10; 6/30] START model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant
- [CV 5/10; 6/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996) total time= 1.0s
- [CV 2/10; 8/30] START model__alpha=1e-06, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant
- [CV 2/10; 8/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant;, score=(train=0.998, test=0.998) t otal time= 0.8s
- [CV 7/10; 9/30] START model__alpha=1e-06, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant
- [CV 7/10; 9/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.997, test=0.997) tot al time= 1.0s
- [CV 3/10; 11/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=0.01, model__learning_rate=constant
- [CV 3/10; 11/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.995) total time= 0.9s
- [CV 9/10; 12/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=0.1, model__learning_rate=constant
- [CV 9/10; 12/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant;, score=(train=0.997, test=0.995) total time= 0.7s
- [CV 5/10; 14/30] START model__alpha=1e-05, model__class_weight=balanced, mod el eta0=5, model learning rate=constant
- [CV 5/10; 14/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.996, test=0.996) to tal time= 0.7s
- [CV 1/10; 16/30] START model__alpha=0.0001, model__class_weight=balanced, model eta0=0.01, model learning rate=constant
- [CV 1/10; 16/30] END model__alpha=0.0001, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.997) total time= 0.8s
- [CV 7/10; 17/30] START model__alpha=0.0001, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant

```
[CV 7/10; 17/30] END model alpha=0.0001, model class weight=balanced, mode
1__eta0=0.1, model__learning_rate=constant;, score=(train=0.996, test=0.997)
total time=
             0.6s
[CV 3/10; 19/30] START model_alpha=0.0001, model_class_weight=balanced, mo
del__eta0=5, model__learning_rate=constant
[CV 3/10; 19/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1__eta0=5, model__learning_rate=constant;, score=(train=0.996, test=0.997) t
otal time=
            0.6s
[CV 9/10; 20/30] START model alpha=0.0001, model class weight=balanced, mo
del__eta0=10, model__learning_rate=constant
[CV 9/10; 20/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=10, model learning rate=constant;, score=(train=0.990, test=0.987)
total time=
             0.7s
[CV 5/10; 22/30] START model alpha=0.001, model class weight=balanced, mod
el eta0=0.1, model__learning_rate=constant
[CV 5/10; 22/30] END model alpha=0.001, model class weight=balanced, model
 _eta0=0.1, model__learning_rate=constant;, score=(train=0.991, test=0.989)
total time=
             0.5s
[CV 1/10; 24/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=5, model__learning_rate=constant
[CV 1/10; 24/30] END model alpha=0.001, model class weight=balanced, model
 eta0=5, model learning rate=constant;, score=(train=0.990, test=0.991) to
tal time= 0.8s
[CV 7/10; 25/30] START model__alpha=0.001, model__class_weight=balanced, mod
el__eta0=10, model__learning_rate=constant
[CV 7/10; 25/30] END model alpha=0.001, model class weight=balanced, model
 eta0=10, model learning rate=constant;, score=(train=0.994, test=0.996) t
otal time=
            0.7s
[CV 3/10; 27/30] START model alpha=0.01, model class weight=balanced, mode
l eta0=0.1, model learning rate=constant
[CV 3/10; 27/30] END model alpha=0.01, model class weight=balanced, model
_eta0=0.1, model__learning_rate=constant;, score=(train=0.991, test=0.990) t
otal time= 0.6s
[CV 9/10; 28/30] START model__alpha=0.01, model__class_weight=balanced, mode
l__eta0=1.0, model__learning_rate=constant
[CV 9/10; 28/30] END model_alpha=0.01, model_class_weight=balanced, model_
_eta0=1.0, model__learning_rate=constant;, score=(train=0.991, test=0.987) t
otal time= 0.6s
[CV 5/10; 30/30] START model_alpha=0.01, model_class_weight=balanced, mode
1 eta0=10, model learning rate=constant
[CV 5/10; 30/30] END model alpha=0.01, model class weight=balanced, model
eta0=10, model learning rate=constant;, score=(train=0.693, test=0.691) to
tal time=
          1.1s
[CV 7/10; 1/30] START model__alpha=1e-06, model__class_weight=balanced, mode
```

- l__eta0=0, model__learning_rate=optimal
- [CV 7/10; 1/30] END model alpha=1e-06, model class weight=balanced, model eta0=0, model learning rate=optimal;, score=(train=0.998, test=0.998) tota
- [CV 9/10; 3/30] START model alpha=0.0001, model class weight=balanced, mod el__eta0=0, model__learning_rate=optimal
- [CV 9/10; 3/30] END model__alpha=0.0001, model__class_weight=balanced, model _eta0=0, model__learning_rate=optimal;, score=(train=0.996, test=0.995) tot al time=
- [CV 6/10; 5/30] START model alpha=0.01, model class weight=balanced, model eta0=0, model__learning_rate=optimal
- [CV 6/10; 5/30] END model alpha=0.01, model class weight=balanced, model eta0=0, model__learning_rate=optimal;, score=(train=0.991, test=0.990) total

```
time=
        0.7s
[CV 2/10; 7/30] START model alpha=1e-06, model class weight=balanced, mode
l eta0=0.1, model learning rate=constant
[CV 2/10; 7/30] END model__alpha=1e-06, model__class_weight=balanced, model_
eta0=0.1, model_learning_rate=constant;, score=(train=0.997, test=0.997) t
otal time=
             0.9s
[CV 3/10; 8/30] START model alpha=1e-06, model class weight=balanced, mode
1__eta0=1.0, model__learning_rate=constant
[CV 3/10; 8/30] END model alpha=1e-06, model class weight=balanced, model
_eta0=1.0, model__learning_rate=constant;, score=(train=0.998, test=0.998) t
otal time=
             0.9s
[CV 10/10; 9/30] START model alpha=1e-06, model class weight=balanced, mod
   _eta0=5, model__learning_rate=constant
[CV 10/10; 9/30] END model alpha=1e-06, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.998, test=0.997) to
tal time=
           1.0s
[CV 5/10; 11/30] START model__alpha=1e-05, model__class_weight=balanced, mod
el__eta0=0.01, model__learning_rate=constant
[CV 5/10; 11/30] END model alpha=1e-05, model class weight=balanced, model
__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996)
             0.9s
total time=
[CV 1/10; 13/30] START model_alpha=1e-05, model_class_weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 1/10; 13/30] END model alpha=1e-05, model class weight=balanced, model
eta0=1.0, model__learning_rate=constant;, score=(train=0.997, test=0.997)
total time=
             0.7s
[CV 7/10; 14/30] START model _alpha=1e-05, model__class_weight=balanced, mod
el eta0=5, model__learning_rate=constant
[CV 7/10; 14/30] END model alpha=1e-05, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.996, test=0.996) to
tal time=
           0.8s
[CV 4/10; 16/30] START model__alpha=0.0001, model__class_weight=balanced, mo
del eta0=0.01, model learning rate=constant
[CV 4/10; 16/30] END model alpha=0.0001, model class weight=balanced, mode
1__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.997
) total time=
               0.9s
[CV 10/10; 17/30] START model alpha=0.0001, model class weight=balanced, m
odel__eta0=0.1, model__learning_rate=constant
[CV 10/10; 17/30] END model_alpha=0.0001, model_class_weight=balanced, mod
el eta0=0.1, model_learning_rate=constant;, score=(train=0.996, test=0.996
) total time=
               0.7s
[CV 7/10; 19/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=5, model learning rate=constant
[CV 7/10; 19/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1__eta0=5, model__learning_rate=constant;, score=(train=0.720, test=0.720) t
otal time=
            0.8s
[CV 3/10; 21/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=0.01, model__learning_rate=constant
[CV 3/10; 21/30] END model_alpha=0.001, model_class_weight=balanced, model
 eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.8s
[CV 9/10; 22/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=0.1, model__learning_rate=constant
[CV 9/10; 22/30] END model alpha=0.001, model class weight=balanced, model
 eta0=0.1, model learning rate=constant;, score=(train=0.991, test=0.987)
total time=
             0.7s
[CV 7/10; 24/30] START model__alpha=0.001, model__class_weight=balanced, mod
```

```
el eta0=5, model learning rate=constant
[CV 7/10; 24/30] END model alpha=0.001, model class weight=balanced, model
 eta0=5, model learning rate=constant;, score=(train=0.990, test=0.992) to
tal time=
           0.8s
[CV 1/10; 26/30] START model_alpha=0.01, model_class_weight=balanced, mode
l__eta0=0.01, model__learning_rate=constant
[CV 1/10; 26/30] END model alpha=0.01, model class weight=balanced, model
eta0=0.01, model_learning_rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.7s
[CV 4/10; 27/30] START model_alpha=0.01, model_class_weight=balanced, mode
l__eta0=0.1, model__learning_rate=constant
[CV 4/10; 27/30] END model alpha=0.01, model class weight=balanced, model
eta0=0.1, model learning rate=constant;, score=(train=0.991, test=0.991) t
otal time=
            0.7s
[CV 10/10; 28/30] START model alpha=0.01, model class weight=balanced, mod
el eta0=1.0, model learning rate=constant
[CV 10/10; 28/30] END model__alpha=0.01, model__class_weight=balanced, model
 eta0=1.0, model learning rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.6s
[CV 6/10; 30/30] START model_alpha=0.01, model_class_weight=balanced, mode
l eta0=10, model learning rate=constant
[CV 6/10; 30/30] END model alpha=0.01, model class weight=balanced, model
eta0=10, model_learning_rate=constant;, score=(train=0.676, test=0.675) to
           1.0s
tal time=
[CV 5/10; 2/30] START model_alpha=1e-05, model_class_weight=balanced, mode
l eta0=0, model learning rate=optimal
[CV 5/10; 2/30] END model alpha=1e-05, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.997, test=0.997) tota
l time=
         1.0s
[CV 10/10; 2/30] START model alpha=1e-05, model class weight=balanced, mod
el eta0=0, model learning rate=optimal
[CV 10/10; 2/30] END model__alpha=1e-05, model__class_weight=balanced, model
 eta0=0, model learning rate=optimal;, score=(train=0.997, test=0.997) tot
al time=
          0.8s
[CV 8/10; 4/30] START model_alpha=0.001, model_class_weight=balanced, mode
l__eta0=0, model__learning_rate=optimal
[CV 8/10; 4/30] END model__alpha=0.001, model__class_weight=balanced, model__
eta0=0, model learning rate=optimal;, score=(train=0.990, test=0.992) tota
l time=
         0.6s
[CV 2/10; 6/30] START model alpha=1e-06, model class weight=balanced, mode
  eta0=0.01, model learning rate=constant
[CV 2/10; 6/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.996)
total time=
             0.9s
[CV 8/10; 7/30] START model alpha=1e-06, model class weight=balanced, mode
  eta0=0.1, model learning rate=constant
[CV 8/10; 7/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0.1, model_learning_rate=constant;, score=(train=0.997, test=0.996) t
otal time=
            0.7s
[CV 4/10; 9/30] START model_alpha=1e-06, model_class_weight=balanced, mode
```

- 1__eta0=5, model__learning_rate=constant
 [CV 4/10; 9/30] END model__alpha=1e-06, model__class_weight=balanced, model__
- _eta0=5, model__learning_rate=constant;, score=(train=0.998, test=0.997) tot al time= 0.9s
- [CV 9/10; 10/30] START model__alpha=1e-06, model__class_weight=balanced, mod el__eta0=10, model__learning_rate=constant
- [CV 9/10; 10/30] END model__alpha=1e-06, model__class_weight=balanced, model

```
eta0=10, model learning rate=constant;, score=(train=0.998, test=0.997) t
            0.8s
otal time=
[CV 6/10; 12/30] START model alpha=1e-05, model class weight=balanced, mod
el__eta0=0.1, model__learning_rate=constant
[CV 6/10; 12/30] END model _alpha=1e-05, model _class_weight=balanced, model
eta0=0.1, model learning rate=constant;, score=(train=0.996, test=0.996)
[CV 2/10; 14/30] START model_alpha=1e-05, model_class_weight=balanced, mod
el eta0=5, model learning rate=constant
[CV 2/10; 14/30] END model alpha=1e-05, model class weight=balanced, model
__eta0=5, model__learning_rate=constant;, score=(train=0.997, test=0.997) to
tal time=
           0.6s
[CV 7/10; 15/30] START model alpha=1e-05, model class weight=balanced, mod
el eta0=10, model learning rate=constant
[CV 7/10; 15/30] END model alpha=1e-05, model class weight=balanced, model
eta0=10, model learning rate=constant;, score=(train=0.995, test=0.996) t
otal time=
            0.9s
[CV 5/10; 17/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=0.1, model learning rate=constant
[CV 5/10; 17/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=0.1, model learning rate=constant;, score=(train=0.996, test=0.996)
total time=
             0.6s
[CV 1/10; 19/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=5, model learning rate=constant
[CV 1/10; 19/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=5, model learning rate=constant;, score=(train=0.995, test=0.995) t
otal time=
            0.7s
[CV 7/10; 20/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=10, model learning rate=constant
[CV 7/10; 20/30] END model alpha=0.0001, model class weight=balanced, mode
1 eta0=10, model learning rate=constant;, score=(train=0.996, test=0.997)
total time=
             0.8s
[CV 2/10; 22/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=0.1, model__learning_rate=constant
[CV 2/10; 22/30] END model alpha=0.001, model class weight=balanced, model
 eta0=0.1, model learning rate=constant;, score=(train=0.990, test=0.992)
total time=
              0.7s
[CV 7/10; 23/30] START model_alpha=0.001, model_class_weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 7/10; 23/30] END model alpha=0.001, model class weight=balanced, model
 eta0=1.0, model learning rate=constant;, score=(train=0.990, test=0.992)
total time=
              0.7s
[CV 2/10; 25/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=10, model__learning_rate=constant
[CV 2/10; 25/30] END model alpha=0.001, model class weight=balanced, model
 eta0=10, model learning rate=constant;, score=(train=0.990, test=0.991) t
otal time=
            1.1s
[CV 9/10; 26/30] START model alpha=0.01, model class weight=balanced, mode
l eta0=0.01, model__learning_rate=constant
[CV 9/10; 26/30] END model _alpha=0.01, model _class weight=balanced, model_
eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.987)
total time=
              0.7s
[CV 5/10; 28/30] START model alpha=0.01, model class weight=balanced, mode
l__eta0=1.0, model__learning_rate=constant
[CV 5/10; 28/30] END model_alpha=0.01, model_class_weight=balanced, model_
eta0=1.0, model learning rate=constant;, score=(train=0.991, test=0.989) t
otal time=
            0.8s
```

```
[CV 2/10; 30/30] START model__alpha=0.01, model__class_weight=balanced, mode l__eta0=10, model__learning_rate=constant
[CV 2/10; 30/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.990, test=0.992) to tal time= 1.2s
[CV 3/10; 2/30] START model__alpha=1e-05, model__class_weight=balanced, mode l__eta0=0, model__learning_rate=optimal
[CV 3/10; 2/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.997, test=0.996) tota l time= 1.0s
[CV 8/10; 2/30] START model__alpha=1e-05, model__class_weight=balanced, mode
```

[CV 8/10; 2/30] END model__alpha=1e-05, model__class_weight=balanced, model_ eta0=0, model learning rate=optimal;, score=(train=0.997, test=0.997) tota

[CV 5/10; 4/30] START model alpha=0.001, model class weight=balanced, mode

[CV 5/10; 4/30] END model__alpha=0.001, model__class_weight=balanced, model_ eta0=0, model learning_rate=optimal;, score=(train=0.991, test=0.989) tota

[CV 10/10; 5/30] START model alpha=0.01, model class weight=balanced, mode

[CV 10/10; 5/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.991, test=0.990) tota

[CV 5/10; 7/30] START model alpha=1e-06, model class weight=balanced, mode

[CV 5/10; 7/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant;, score=(train=0.997, test=0.997) t

[CV 2/10; 9/30] START model alpha=1e-06, model class weight=balanced, mode

[CV 2/10; 9/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.998, test=0.997) tot

[CV 8/10; 10/30] START model alpha=1e-06, model class weight=balanced, mod

[CV 8/10; 10/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.997, test=0.997) t

[CV 4/10; 12/30] START model alpha=1e-05, model class weight=balanced, mod

[CV 4/10; 12/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant;, score=(train=0.997, test=0.997)

[CV 10/10; 13/30] START model alpha=1e-05, model class weight=balanced, mo

[CV 6/10; 15/30] START model alpha=1e-05, model class_weight=balanced, mod

[CV 6/10; 15/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.995, test=0.995) t

[CV 3/10; 17/30] START model alpha=0.0001, model class weight=balanced, mo

l eta0=0, model__learning_rate=optimal

l__eta0=0, model__learning_rate=optimal

1 eta0=0, model__learning_rate=optimal

eta0=0.1, model learning rate=constant

_eta0=5, model__learning_rate=constant

el__eta0=10, model__learning_rate=constant

el eta0=0.1, model learning rate=constant

del eta0=1.0, model learning rate=constant

el eta0=10, model learning rate=constant

del__eta0=0.1, model__learning_rate=constant

l time=

l time=

l time=

otal time=

al time=

otal time=

total time=

total time=

otal time=

0.7s

0.6s

0.5s

1.0s

0.9s

0.7s

0.7s

0.8s

```
[CV 3/10; 17/30] END model alpha=0.0001, model class weight=balanced, mode
1__eta0=0.1, model__learning_rate=constant;, score=(train=0.996, test=0.995)
total time=
             0.7s
[CV 9/10; 18/30] START model_alpha=0.0001, model_class_weight=balanced, mo
del__eta0=1.0, model__learning_rate=constant
[CV 9/10; 18/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1__eta0=1.0, model__learning_rate=constant;, score=(train=0.996, test=0.995)
total time=
             0.7s
[CV 4/10; 20/30] START model alpha=0.0001, model class weight=balanced, mo
del__eta0=10, model__learning_rate=constant
[CV 4/10; 20/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=10, model learning rate=constant;, score=(train=0.990, test=0.991)
total time=
             0.9s
[CV 10/10; 21/30] START model alpha=0.001, model class weight=balanced, mo
del eta0=0.01, model learning rate=constant
[CV 10/10; 21/30] END model alpha=0.001, model class weight=balanced, mode
1__eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.990
) total time=
              0.8s
[CV 8/10; 23/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 8/10; 23/30] END model alpha=0.001, model class weight=balanced, model
 eta0=1.0, model_learning_rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.7s
[CV 4/10; 25/30] START model__alpha=0.001, model__class_weight=balanced, mod
el__eta0=10, model__learning_rate=constant
[CV 4/10; 25/30] END model alpha=0.001, model class weight=balanced, model
 _eta0=10, model__learning_rate=constant;, score=(train=0.729, test=0.729) t
otal time=
            0.9s
[CV 6/10; 26/30] START model alpha=0.01, model class weight=balanced, mode
l eta0=0.01, model learning rate=constant
[CV 6/10; 26/30] END model alpha=0.01, model class weight=balanced, model
_eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.990)
total time=
[CV 1/10; 28/30] START model__alpha=0.01, model__class_weight=balanced, mode
l__eta0=1.0, model__learning_rate=constant
[CV 1/10; 28/30] END model_alpha=0.01, model_class_weight=balanced, model_
_eta0=1.0, model__learning_rate=constant;, score=(train=0.990, test=0.992) t
otal time= 0.8s
[CV 5/10; 29/30] START model alpha=0.01, model class weight=balanced, mode
1 eta0=5, model learning rate=constant
[CV 5/10; 29/30] END model alpha=0.01, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.728, test=0.727) tot
al time=
          1.2s
[CV 4/10; 1/30] START model__alpha=1e-06, model__class_weight=balanced, mode
l__eta0=0, model__learning_rate=optimal
[CV 4/10; 1/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.998, test=0.998) tota
l time=
```

[CV 4/10; 3/30] START model alpha=0.0001, model class weight=balanced, mod

[CV 4/10; 3/30] END model__alpha=0.0001, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.996, test=0.997) tot

[CV 2/10; 5/30] START model alpha=0.01, model class weight=balanced, model

[CV 2/10; 5/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.990, test=0.992) total

el__eta0=0, model__learning_rate=optimal

_eta0=0, model__learning_rate=optimal

al time=

```
time=
        0.7s
[CV 6/10; 6/30] START model alpha=1e-06, model class weight=balanced, mode
l eta0=0.01, model learning rate=constant
[CV 6/10; 6/30] END model__alpha=1e-06, model__class_weight=balanced, model__
eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996)
total time=
              1.0s
[CV 1/10; 8/30] START model_alpha=1e-06, model_class_weight=balanced, mode
l__eta0=1.0, model__learning_rate=constant
[CV 1/10; 8/30] END model alpha=1e-06, model class weight=balanced, model
_eta0=1.0, model__learning_rate=constant;, score=(train=0.998, test=0.997) t
otal time=
             0.8s
[CV 9/10; 9/30] START model alpha=1e-06, model class weight=balanced, mode
  eta0=5, model learning rate=constant
[CV 9/10; 9/30] END model alpha=1e-06, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.998, test=0.997) tot
al time=
          1.1s
[CV 8/10; 11/30] START model__alpha=1e-05, model__class_weight=balanced, mod
el__eta0=0.01, model__learning_rate=constant
[CV 8/10; 11/30] END model alpha=1e-05, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.996)
             1.0s
total time=
[CV 7/10; 13/30] START model__alpha=1e-05, model__class_weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 7/10; 13/30] END model alpha=1e-05, model class weight=balanced, model
__eta0=1.0, model__learning_rate=constant;, score=(train=0.997, test=0.997)
total time=
              0.7s
[CV 3/10; 15/30] START model alpha=1e-05, model class weight=balanced, mod
el eta0=10, model__learning_rate=constant
[CV 3/10; 15/30] END model alpha=1e-05, model class weight=balanced, model
eta0=10, model learning rate=constant;, score=(train=0.996, test=0.995) t
otal time=
            0.9s
[CV 10/10; 16/30] START model__alpha=0.0001, model__class_weight=balanced, m
odel eta0=0.01, model learning rate=constant
[CV 10/10; 16/30] END model alpha=0.0001, model class weight=balanced, mod
el__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.99
6) total time=
                0.9s
[CV 7/10; 18/30] START model_alpha=0.0001, model_class_weight=balanced, mo
del__eta0=1.0, model__learning_rate=constant
[CV 7/10; 18/30] END model_alpha=0.0001, model_class_weight=balanced, mode
1 eta0=1.0, model__learning_rate=constant;, score=(train=0.995, test=0.996)
             0.7s
total time=
[CV 2/10; 20/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=10, model learning rate=constant
[CV 2/10; 20/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1__eta0=10, model__learning_rate=constant;, score=(train=0.996, test=0.997)
total time=
             0.9s
[CV 8/10; 21/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=0.01, model__learning_rate=constant
[CV 8/10; 21/30] END model alpha=0.001, model class weight=balanced, model
 eta0=0.01, model__learning_rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.8s
[CV 4/10; 23/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 4/10; 23/30] END model alpha=0.001, model class weight=balanced, model
```

[CV 10/10; 24/30] START model__alpha=0.001, model__class_weight=balanced, mo

eta0=1.0, model learning rate=constant;, score=(train=0.991, test=0.991)

total time=

0.7s

```
del eta0=5, model learning rate=constant
[CV 10/10; 24/30] END model alpha=0.001, model class weight=balanced, mode
1 eta0=5, model learning rate=constant;, score=(train=0.991, test=0.990) t
otal time=
            0.9s
[CV 5/10; 26/30] START model alpha=0.01, model class weight=balanced, mode
l__eta0=0.01, model__learning_rate=constant
[CV 5/10; 26/30] END model alpha=0.01, model class weight=balanced, model
eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.989)
total time=
              0.8s
[CV 10/10; 27/30] START model_alpha=0.01, model_class_weight=balanced, mod
el__eta0=0.1, model__learning_rate=constant
[CV 10/10; 27/30] END model alpha=0.01, model class weight=balanced, model
eta0=0.1, model learning rate=constant;, score=(train=0.991, test=0.990)
total time=
              0.8s
[CV 8/10; 29/30] START model alpha=0.01, model class weight=balanced, mode
1 eta0=5, model learning rate=constant
[CV 8/10; 29/30] END model__alpha=0.01, model__class_weight=balanced, model_
eta0=5, model learning rate=constant;, score=(train=0.693, test=0.694) tot
al time=
          1.2s
[CV 3/10; 1/30] START model_alpha=1e-06, model_class_weight=balanced, mode
l eta0=0, model learning rate=optimal
[CV 3/10; 1/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.998, test=0.997) tota
l time=
         1.4s
[CV 2/10; 4/30] START model__alpha=0.001, model__class_weight=balanced, mode
l eta0=0, model learning rate=optimal
[CV 2/10; 4/30] END model alpha=0.001, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.990, test=0.992) tota
l time=
         0.7s
[CV 8/10; 5/30] START model alpha=0.01, model class weight=balanced, model
 eta0=0, model__learning_rate=optimal
[CV 8/10; 5/30] END model__alpha=0.01, model__class_weight=balanced, model_
eta0=0, model learning rate=optimal;, score=(train=0.990, test=0.992) total
time=
        0.6s
[CV 3/10; 7/30] START model_alpha=1e-06, model_class_weight=balanced, mode
l__eta0=0.1, model__learning_rate=constant
[CV 3/10; 7/30] END model__alpha=1e-06, model__class_weight=balanced, model_
_eta0=0.1, model__learning_rate=constant;, score=(train=0.997, test=0.998) t
otal time=
            0.9s
[CV 6/10; 8/30] START model alpha=1e-06, model class weight=balanced, mode
  eta0=1.0, model learning rate=constant
[CV 6/10; 8/30] END model alpha=1e-06, model class weight=balanced, model
eta0=1.0, model learning rate=constant;, score=(train=0.999, test=0.997) t
otal time=
            0.9s
```

- [CV 4/10; 10/30] START model__alpha=1e-06, model__class_weight=balanced, mod el__eta0=10, model__learning_rate=constant
- [CV 4/10; 10/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.997, test=0.998) t otal time= 1.0s
- [CV 7/10; 11/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=0.01, model__learning_rate=constant
- [CV 7/10; 11/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.997) total time= 0.9s
- [CV 4/10; 13/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=1.0, model__learning_rate=constant
- [CV 4/10; 13/30] END model__alpha=1e-05, model__class_weight=balanced, model

```
eta0=1.0, model learning rate=constant;, score=(train=0.997, test=0.998)
             0.7s
total time=
[CV 9/10; 14/30] START model alpha=1e-05, model class weight=balanced, mod
el__eta0=5, model__learning_rate=constant
[CV 9/10; 14/30] END model _alpha=1e-05, model _class_weight=balanced, model
__eta0=5, model__learning_rate=constant;, score=(train=0.997, test=0.995) to
[CV 5/10; 16/30] START model_alpha=0.0001, model_class_weight=balanced, mo
del eta0=0.01, model learning rate=constant
[CV 5/10; 16/30] END model_alpha=0.0001, model_class_weight=balanced, mode
1__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996
              0.9s
) total time=
[CV 1/10; 18/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=1.0, model learning rate=constant
[CV 1/10; 18/30] END model alpha=0.0001, model class weight=balanced, mode
1 eta0=1.0, model learning rate=constant;, score=(train=0.996, test=0.997)
             0.7s
total time=
[CV 6/10; 19/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=5, model learning rate=constant
[CV 6/10; 19/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=5, model learning rate=constant;, score=(train=0.995, test=0.994) t
otal time=
            0.8s
[CV 4/10; 21/30] START model alpha=0.001, model class_weight=balanced, mod
el__eta0=0.01, model__learning_rate=constant
[CV 4/10; 21/30] END model_alpha=0.001, model_class_weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.991, test=0.991)
total time=
             0.8s
[CV 10/10; 22/30] START model alpha=0.001, model class weight=balanced, mo
del eta0=0.1, model learning rate=constant
[CV 10/10; 22/30] END model alpha=0.001, model class weight=balanced, mode
1 eta0=0.1, model learning rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.7s
[CV 5/10; 24/30] START model alpha=0.001, model class weight=balanced, mod
el eta0=5, model learning rate=constant
[CV 5/10; 24/30] END model alpha=0.001, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.991, test=0.989) to
tal time=
           0.9s
[CV 3/10; 26/30] START model_alpha=0.01, model_class_weight=balanced, mode
l__eta0=0.01, model__learning_rate=constant
[CV 3/10; 26/30] END model alpha=0.01, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.7s
[CV 7/10; 27/30] START model alpha=0.01, model class weight=balanced, mode
l__eta0=0.1, model__learning_rate=constant
[CV 7/10; 27/30] END model alpha=0.01, model class weight=balanced, model
eta0=0.1, model learning rate=constant;, score=(train=0.990, test=0.992) t
otal time=
            0.8s
[CV 4/10; 29/30] START model alpha=0.01, model class weight=balanced, mode
1 eta0=5, model learning rate=constant
[CV 4/10; 29/30] END model alpha=0.01, model class weight=balanced, model
_eta0=5, model__learning_rate=constant;, score=(train=0.712, test=0.715) tot
al time=
           1.1s
[CV 1/10; 1/30] START model_alpha=1e-06, model_class_weight=balanced, mode
l__eta0=0, model__learning_rate=optimal
[CV 1/10; 1/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.999, test=0.997) tota
l time=
         1.4s
```

```
[CV 1/10; 4/30] START model__alpha=0.001, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal
[CV 1/10; 4/30] END model__alpha=0.001, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.990, test=0.992) tota
```

[CV 5/10; 5/30] START model alpha=0.01, model class weight=balanced, model

[CV 5/10; 5/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=0, model learning rate=optimal;, score=(train=0.991, test=0.989) total

[CV 1/10; 7/30] START model__alpha=1e-06, model__class_weight=balanced, mode

[CV 1/10; 7/30] END model__alpha=1e-06, model__class_weight=balanced, model_ eta0=0.1, model learning rate=constant;, score=(train=0.997, test=0.997) t

[CV 10/10; 8/30] START model alpha=1e-06, model class weight=balanced, mod

[CV 10/10; 8/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant;, score=(train=0.998, test=0.998)

[CV 6/10; 10/30] START model alpha=1e-06, model class weight=balanced, mod

[CV 6/10; 10/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.998, test=0.997) t

[CV 1/10; 12/30] START model alpha=1e-05, model class weight=balanced, mod

[CV 1/10; 12/30] END model__alpha=1e-05, model__class_weight=balanced, model eta0=0.1, model learning rate=constant;, score=(train=0.996, test=0.997)

[CV 2/10; 13/30] START model alpha=1e-05, model class weight=balanced, mod

[CV 2/10; 13/30] END model__alpha=1e-05, model__class_weight=balanced, model eta0=1.0, model__learning_rate=constant;, score=(train=0.997, test=0.997)

[CV 10/10; 14/30] START model__alpha=1e-05, model__class_weight=balanced, mo

[CV 10/10; 14/30] END model__alpha=1e-05, model__class_weight=balanced, mode 1__eta0=5, model__learning_rate=constant;, score=(train=0.996, test=0.996) t

[CV 7/10; 16/30] START model alpha=0.0001, model class weight=balanced, mo

[CV 7/10; 16/30] END model__alpha=0.0001, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.997

[CV 4/10; 18/30] START model_alpha=0.0001, model class weight=balanced, mo

[CV 4/10; 18/30] END model__alpha=0.0001, model__class_weight=balanced, model _ eta0=1.0, model__learning_rate=constant;, score=(train=0.991, test=0.991)

[CV 10/10; 19/30] START model alpha=0.0001, model class weight=balanced, m

[CV 10/10; 19/30] END model__alpha=0.0001, model__class_weight=balanced, mod el__eta0=5, model__learning_rate=constant;, score=(train=0.996, test=0.996)

[CV 5/10; 21/30] START model alpha=0.001, model class weight=balanced, mod

1 time = 0.7s

otal time=

total time=

otal time=

total time=

total time=

otal time=

) total time=

total time=

total time=

eta0=0, model learning rate=optimal

l eta0=0.1, model learning rate=constant

el__eta0=1.0, model__learning_rate=constant

el__eta0=10, model__learning_rate=constant

eta0=0.1, model learning rate=constant

el__eta0=1.0, model__learning_rate=constant

del__eta0=5, model__learning_rate=constant

del__eta0=0.01, model__learning_rate=constant

del eta0=1.0, model learning rate=constant

odel eta0=5, model learning rate=constant

el__eta0=0.01, model__learning_rate=constant

0.9s

0.9s

1.0s

0.8s

0.8s

0.8s

0.9s

0.7s

0.8s

```
[CV 5/10; 21/30] END model alpha=0.001, model class weight=balanced, model
 _eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.989)
total time=
             0.8s
[CV 1/10; 23/30] START model_alpha=0.001, model_class_weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant
[CV 1/10; 23/30] END model alpha=0.001, model class weight=balanced, model
 eta0=1.0, model learning_rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.6s
[CV 4/10; 24/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=5, model__learning_rate=constant
[CV 4/10; 24/30] END model__alpha=0.001, model__class_weight=balanced, model
 eta0=5, model learning rate=constant;, score=(train=0.991, test=0.991) to
tal time=
           1.0s
[CV 4/10; 26/30] START model alpha=0.01, model class weight=balanced, mode
l eta0=0.01, model__learning_rate=constant
[CV 4/10; 26/30] END model alpha=0.01, model class weight=balanced, model
_eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.991)
total time=
             0.8s
[CV 9/10; 27/30] START model_alpha=0.01, model_class_weight=balanced, mode
  _eta0=0.1, model__learning_rate=constant
[CV 9/10; 27/30] END model alpha=0.01, model class weight=balanced, model
eta0=0.1, model__learning_rate=constant;, score=(train=0.991, test=0.987) t
otal time=
           0.8s
[CV 6/10; 29/30] START model alpha=0.01, model class weight=balanced, mode
1__eta0=5, model__learning_rate=constant
[CV 6/10; 29/30] END model alpha=0.01, model class weight=balanced, model
_eta0=5, model__learning_rate=constant;, score=(train=0.728, test=0.727) tot
          1.0s
al time=
[CV 8/10; 1/30] START model alpha=1e-06, model class weight=balanced, mode
l eta0=0, model learning rate=optimal
[CV 8/10; 1/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.998, test=0.997) tota
[CV 8/10; 3/30] START model alpha=0.0001, model class weight=balanced, mod
el__eta0=0, model__learning_rate=optimal
[CV 8/10; 3/30] END model_alpha=0.0001, model_class_weight=balanced, model
 eta0=0, model learning rate=optimal;, score=(train=0.996, test=0.996) tot
          0.7s
[CV 4/10; 5/30] START model alpha=0.01, model class weight=balanced, model
 eta0=0, model learning rate=optimal
[CV 4/10; 5/30] END model alpha=0.01, model class weight=balanced, model
```

eta0=0, model learning rate=optimal;, score=(train=0.991, test=0.991) total

[CV 9/10; 6/30] START model__alpha=1e-06, model__class_weight=balanced, mode

[CV 9/10; 6/30] END model__alpha=1e-06, model__class_weight=balanced, model_ eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.995)

[CV 5/10; 8/30] START model alpha=1e-06, model class weight=balanced, mode

[CV 5/10; 8/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant;, score=(train=0.998, test=0.997) t

[CV 8/10; 9/30] START model_alpha=1e-06, model_class_weight=balanced, mode

[CV 8/10; 9/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.998, test=0.997) tot

l eta0=0.01, model learning rate=constant

l__eta0=1.0, model__learning_rate=constant

1 eta0=5, model__learning_rate=constant

total time=

```
al time=
          1.0s
[CV 4/10; 11/30] START model alpha=1e-05, model class weight=balanced, mod
el eta0=0.01, model learning rate=constant
[CV 4/10; 11/30] END model alpha=1e-05, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.997)
total time=
             1.0s
[CV 10/10; 12/30] START model alpha=1e-05, model class weight=balanced, mo
del__eta0=0.1, model__learning_rate=constant
[CV 10/10; 12/30] END model alpha=1e-05, model class weight=balanced, mode
1__eta0=0.1, model__learning_rate=constant;, score=(train=0.997, test=0.996)
total time=
              0.7s
[CV 6/10; 14/30] START model alpha=1e-05, model class weight=balanced, mod
   _eta0=5, model__learning_rate=constant
[CV 6/10; 14/30] END model alpha=1e-05, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.996, test=0.996) to
tal time=
            0.8s
[CV 3/10; 16/30] START model__alpha=0.0001, model__class_weight=balanced, mo
del__eta0=0.01, model__learning_rate=constant
[CV 3/10; 16/30] END model_alpha=0.0001, model_class_weight=balanced, mode
1__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.995
) total time=
               0.9s
[CV 9/10; 17/30] START model_alpha=0.0001, model_class_weight=balanced, mo
del__eta0=0.1, model__learning_rate=constant
[CV 9/10; 17/30] END model alpha=0.0001, model class weight=balanced, mode
1__eta0=0.1, model__learning_rate=constant;, score=(train=0.996, test=0.995)
total time=
             0.7s
[CV 5/10; 19/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=5, model__learning_rate=constant
[CV 5/10; 19/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=5, model learning rate=constant;, score=(train=0.994, test=0.994) t
otal time=
            0.8s
[CV 2/10; 21/30] START model alpha=0.001, model class weight=balanced, mod
el eta0=0.01, model learning rate=constant
[CV 2/10; 21/30] END model alpha=0.001, model class weight=balanced, model
 eta0=0.01, model learning rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.8s
[CV 8/10; 22/30] START model_alpha=0.001, model_class_weight=balanced, mod
el__eta0=0.1, model__learning_rate=constant
[CV 8/10; 22/30] END model _alpha=0.001, model _class_weight=balanced, model
```

eta0=0.1, model learning rate=constant;, score=(train=0.990, test=0.992)

[CV 3/10; 24/30] START model alpha=0.001, model class weight=balanced, mod

[CV 3/10; 24/30] END model__alpha=0.001, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.990, test=0.989) to

[CV 2/10; 26/30] START model alpha=0.01, model class weight=balanced, mode

[CV 2/10; 26/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.990, test=0.992)

[CV 6/10; 27/30] START model alpha=0.01, model class weight=balanced, mode

[CV 6/10; 27/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant;, score=(train=0.991, test=0.990) t

[CV 2/10; 29/30] START model__alpha=0.01, model__class_weight=balanced, mode

total time=

tal time=

total time=

otal time=

0.7s

0.9s

0.7s

0.8s

el eta0=5, model learning rate=constant

eta0=0.01, model_learning_rate=constant

_eta0=0.1, model__learning_rate=constant

```
1 eta0=5, model learning rate=constant
[CV 2/10; 29/30] END model alpha=0.01, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.693, test=0.693) tot
al time=
          1.0s
[CV 8/10; 30/30] START model_alpha=0.01, model_class_weight=balanced, mode
1__eta0=10, model__learning_rate=constant
[CV 8/10; 30/30] END model alpha=0.01, model class weight=balanced, model
eta0=10, model_learning_rate=constant;, score=(train=0.719, test=0.719) to
tal time=
           0.8s
[CV 9/10; 1/30] START model__alpha=1e-06, model__class_weight=balanced, mode
1__eta0=0, model__learning_rate=optimal
[CV 9/10; 1/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.998, test=0.996) tota
l time=
         1.4s
[CV 10/10; 3/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=0, model learning rate=optimal
[CV 10/10; 3/30] END model__alpha=0.0001, model__class_weight=balanced, mode
1 eta0=0, model learning rate=optimal;, score=(train=0.996, test=0.996) to
tal time=
           0.7s
[CV 7/10; 5/30] START model alpha=0.01, model class weight=balanced, model
 eta0=0, model learning rate=optimal
[CV 7/10; 5/30] END model alpha=0.01, model class weight=balanced, model
eta0=0, model_learning_rate=optimal;, score=(train=0.990, test=0.992) total
       0.6s
time=
[CV 4/10; 7/30] START model__alpha=1e-06, model__class_weight=balanced, mode
l eta0=0.1, model learning rate=constant
[CV 4/10; 7/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0.1, model learning rate=constant;, score=(train=0.997, test=0.998) t
otal time=
            0.8s
[CV 7/10; 8/30] START model alpha=1e-06, model class weight=balanced, mode
l eta0=1.0, model learning rate=constant
[CV 7/10; 8/30] END model__alpha=1e-06, model__class_weight=balanced, model_
eta0=1.0, model learning rate=constant;, score=(train=0.998, test=0.998) t
otal time=
            0.8s
[CV 5/10; 10/30] START model alpha=1e-06, model class weight=balanced, mod
el__eta0=10, model__learning_rate=constant
[CV 5/10; 10/30] END model alpha=1e-06, model class weight=balanced, model
eta0=10, model learning rate=constant;, score=(train=0.991, test=0.988) t
otal time=
            1.0s
[CV 9/10; 11/30] START model alpha=1e-05, model class weight=balanced, mod
   eta0=0.01, model learning rate=constant
[CV 9/10; 11/30] END model alpha=1e-05, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.995)
total time=
             1.0s
```

- [CV 6/10; 13/30] START model__alpha=1e-05, model__class_weight=balanced, mod el eta0=1.0, model learning rate=constant
- [CV 6/10; 13/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant;, score=(train=0.997, test=0.997) total time= 0.8s
- [CV 1/10; 15/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=10, model__learning_rate=constant
- [CV 1/10; 15/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.996, test=0.997) t otal time= 0.8s
- [CV 6/10; 16/30] START model__alpha=0.0001, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant
- [CV 6/10; 16/30] END model__alpha=0.0001, model__class_weight=balanced, mode

```
1 eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.996
) total time=
               0.9s
[CV 2/10; 18/30] START model alpha=0.0001, model class weight=balanced, mo
del__eta0=1.0, model__learning_rate=constant
[CV 2/10; 18/30] END model_alpha=0.0001, model_class_weight=balanced, mode
1__eta0=1.0, model__learning_rate=constant;, score=(train=0.996, test=0.996)
total time=
             0.7s
[CV 8/10; 19/30] START model_alpha=0.0001, model_class_weight=balanced, mo
del eta0=5, model learning rate=constant
[CV 8/10; 19/30] END model_alpha=0.0001, model_class_weight=balanced, mode
1 eta0=5, model learning rate=constant;, score=(train=0.996, test=0.996) t
            0.7s
otal time=
[CV 1/10; 21/30] START model alpha=0.001, model class weight=balanced, mod
el eta0=0.01, model learning rate=constant
[CV 1/10; 21/30] END model alpha=0.001, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.990, test=0.992)
             0.7s
total time=
[CV 7/10; 22/30] START model alpha=0.001, model class weight=balanced, mod
el eta0=0.1, model learning rate=constant
[CV 7/10; 22/30] END model alpha=0.001, model class weight=balanced, model
 eta0=0.1, model__learning_rate=constant;, score=(train=0.990, test=0.992)
total time=
             0.7s
[CV 8/10; 24/30] START model alpha=0.001, model class_weight=balanced, mod
el eta0=5, model learning rate=constant
[CV 8/10; 24/30] END model_alpha=0.001, model_class_weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.728, test=0.729) to
tal time=
           0.7s
[CV 10/10; 25/30] START model alpha=0.001, model class weight=balanced, mo
del eta0=10, model learning rate=constant
[CV 10/10; 25/30] END model alpha=0.001, model class weight=balanced, mode
1 eta0=10, model learning rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.9s
[CV 8/10; 27/30] START model alpha=0.01, model class weight=balanced, mode
l__eta0=0.1, model__learning_rate=constant
[CV 8/10; 27/30] END model alpha=0.01, model class weight=balanced, model
_eta0=0.1, model__learning_rate=constant;, score=(train=0.990, test=0.992) t
otal time=
            0.8s
[CV 3/10; 29/30] START model_alpha=0.01, model_class_weight=balanced, mode
1__eta0=5, model__learning_rate=constant
[CV 3/10; 29/30] END model alpha=0.01, model class weight=balanced, model
eta0=5, model learning rate=constant;, score=(train=0.989, test=0.990) tot
al time=
           1.0s
[CV 9/10; 30/30] START model alpha=0.01, model class weight=balanced, mode
l__eta0=10, model__learning_rate=constant
[CV 9/10; 30/30] END model alpha=0.01, model class weight=balanced, model
_eta0=10, model__learning_rate=constant;, score=(train=0.729, test=0.726) to
tal time=
           0.6s
[CV 5/10; 1/30] START model_alpha=1e-06, model_class_weight=balanced, mode
1 eta0=0, model learning rate=optimal
[CV 5/10; 1/30] END model__alpha=1e-06, model__class_weight=balanced, model__
eta0=0, model learning rate=optimal;, score=(train=0.999, test=0.998) tota
l time=
         1.3s
[CV 5/10; 3/30] START model_alpha=0.0001, model_class_weight=balanced, mod
el__eta0=0, model__learning_rate=optimal
[CV 5/10; 3/30] END model alpha=0.0001, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.996, test=0.996) tot
al time=
          0.7s
```

```
[CV 10/10; 4/30] START model__alpha=0.001, model__class_weight=balanced, mod el eta0=0, model learning rate=optimal
```

- [CV 10/10; 4/30] END model__alpha=0.001, model__class_weight=balanced, model__eta0=0, model__learning_rate=optimal;, score=(train=0.991, test=0.990) tot al time= 0.7s
- [CV 8/10; 6/30] START model__alpha=1e-06, model__class_weight=balanced, model eta0=0.01, model learning rate=constant
- [CV 8/10; 6/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996) total time= 1.0s
- [CV 9/10; 8/30] START model__alpha=1e-06, model__class_weight=balanced, model eta0=1.0, model learning rate=constant
- [CV 9/10; 8/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant;, score=(train=0.998, test=0.995) t otal time= 0.8s
- [CV 3/10; 10/30] START model__alpha=1e-06, model__class_weight=balanced, mod el__eta0=10, model__learning_rate=constant
- [CV 3/10; 10/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.998, test=0.998) t otal time= 0.9s
- [CV 6/10; 11/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=0.01, model__learning_rate=constant
- [CV 6/10; 11/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996) total time= 1.0s
- [CV 3/10; 13/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=1.0, model__learning_rate=constant
- [CV 3/10; 13/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant;, score=(train=0.997, test=0.996) total time= 0.7s
- [CV 8/10; 14/30] START model__alpha=1e-05, model__class_weight=balanced, mod el__eta0=5, model__learning_rate=constant
- [CV 8/10; 14/30] END model__alpha=1e-05, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.996, test=0.996) to tal time= 0.7s
- [CV 2/10; 16/30] START model__alpha=0.0001, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant
- [CV 2/10; 16/30] END model__alpha=0.0001, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.996, test=0.996) total time= 0.9s
- [CV 8/10; 17/30] START model__alpha=0.0001, model__class_weight=balanced, model eta0=0.1, model learning rate=constant
- [CV 8/10; 17/30] END model__alpha=0.0001, model__class_weight=balanced, model__eta0=0.1, model__learning_rate=constant;, score=(train=0.996, test=0.996) total time= 0.7s
- [CV 4/10; 19/30] START model__alpha=0.0001, model__class_weight=balanced, model eta0=5, model learning rate=constant
- [CV 4/10; 19/30] END model__alpha=0.0001, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.991, test=0.991) t otal time= 0.7s
- [CV 10/10; 20/30] START model__alpha=0.0001, model__class_weight=balanced, m odel eta0=10, model learning rate=constant
- [CV 10/10; 20/30] END model__alpha=0.0001, model__class_weight=balanced, mod el__eta0=10, model__learning_rate=constant;, score=(train=0.729, test=0.728) total time= 0.8s
- [CV 6/10; 22/30] START model__alpha=0.001, model__class_weight=balanced, mod el__eta0=0.1, model__learning_rate=constant

```
[CV 6/10; 22/30] END model alpha=0.001, model class weight=balanced, model
 _eta0=0.1, model__learning_rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.6s
[CV 2/10; 24/30] START model_alpha=0.001, model_class_weight=balanced, mod
el__eta0=5, model__learning_rate=constant
[CV 2/10; 24/30] END model alpha=0.001, model class weight=balanced, model
 eta0=5, model_learning_rate=constant;, score=(train=0.990, test=0.992) to
tal time=
           0.9s
[CV 8/10; 25/30] START model alpha=0.001, model class weight=balanced, mod
el__eta0=10, model__learning_rate=constant
[CV 8/10; 25/30] END model__alpha=0.001, model__class_weight=balanced, model
eta0=10, model learning rate=constant;, score=(train=0.990, test=0.991) t
otal time=
            1.1s
[CV 4/10; 28/30] START model alpha=0.01, model class weight=balanced, mode
l eta0=1.0, model learning rate=constant
[CV 4/10; 28/30] END model alpha=0.01, model class weight=balanced, model
_eta0=1.0, model__learning_rate=constant;, score=(train=0.991, test=0.991) t
otal time=
            0.8s
[CV 10/10; 29/30] START model__alpha=0.01, model__class_weight=balanced, mod
   _eta0=5, model__learning_rate=constant
[CV 10/10; 29/30] END model alpha=0.01, model class weight=balanced, model
 eta0=5, model learning rate=constant;, score=(train=0.693, test=0.692) to
tal time= 1.0s
[CV 6/10; 2/30] START model__alpha=1e-05, model__class_weight=balanced, mode
l__eta0=0, model__learning_rate=optimal
[CV 6/10; 2/30] END model alpha=1e-05, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.997, test=0.996) tota
l time= 1.0s
[CV 1/10; 3/30] START model alpha=0.0001, model class weight=balanced, mod
el eta0=0, model learning rate=optimal
[CV 1/10; 3/30] END model alpha=0.0001, model class weight=balanced, model
eta0=0, model learning rate=optimal;, score=(train=0.996, test=0.997) tot
[CV 3/10; 4/30] START model__alpha=0.001, model__class_weight=balanced, mode
1_eta0=0, model_learning_rate=optimal
[CV 3/10; 4/30] END model_alpha=0.001, model_class_weight=balanced, model_
eta0=0, model_learning_rate=optimal;, score=(train=0.991, test=0.990) tota
[CV 1/10; 6/30] START model_alpha=1e-06, model_class_weight=balanced, mode
1 eta0=0.01, model learning rate=constant
[CV 1/10; 6/30] END model alpha=1e-06, model class weight=balanced, model
eta0=0.01, model learning rate=constant;, score=(train=0.996, test=0.997)
total time=
            0.9s
[CV 7/10; 7/30] START model__alpha=1e-06, model__class_weight=balanced, mode
```

[CV 7/10; 7/30] END model__alpha=1e-06, model__class_weight=balanced, model_ eta0=0.1, model learning rate=constant;, score=(train=0.997, test=0.997) t

[CV 3/10; 9/30] START model_alpha=1e-06, model_class_weight=balanced, mode

[CV 3/10; 9/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=5, model__learning_rate=constant;, score=(train=0.997, test=0.996) tot

[CV 10/10; 10/30] START model__alpha=1e-06, model__class_weight=balanced, mo

[CV 10/10; 10/30] END model__alpha=1e-06, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.998, test=0.997)

l__eta0=0.1, model__learning_rate=constant

1__eta0=5, model__learning_rate=constant

del__eta0=10, model__learning_rate=constant

0.9s

otal time=

al time=

```
total time=
              0.7s
[CV 5/10; 12/30] START model alpha=1e-05, model class weight=balanced, mod
el eta0=0.1, model learning rate=constant
[CV 5/10; 12/30] END model alpha=1e-05, model class weight=balanced, model
eta0=0.1, model_learning_rate=constant;, score=(train=0.996, test=0.996)
total time=
              0.6s
[CV 1/10; 14/30] START model alpha=1e-05, model class weight=balanced, mod
el__eta0=5, model__learning_rate=constant
[CV 1/10; 14/30] END model alpha=1e-05, model class weight=balanced, model
__eta0=5, model__learning_rate=constant;, score=(train=0.996, test=0.997) to
tal time=
            0.7s
[CV 8/10; 15/30] START model alpha=1e-05, model class weight=balanced, mod
   _eta0=10, model__learning_rate=constant
[CV 8/10; 15/30] END model alpha=1e-05, model class weight=balanced, model
eta0=10, model learning rate=constant;, score=(train=0.990, test=0.992) t
otal time=
             0.8s
[CV 4/10; 17/30] START model__alpha=0.0001, model__class_weight=balanced, mo
del__eta0=0.1, model__learning_rate=constant
[CV 4/10; 17/30] END model alpha=0.0001, model class weight=balanced, mode
1__eta0=0.1, model__learning_rate=constant;, score=(train=0.996, test=0.997)
             0.7s
total time=
[CV 10/10; 18/30] START model_alpha=0.0001, model_class_weight=balanced, m
odel__eta0=1.0, model__learning_rate=constant
[CV 10/10; 18/30] END model__alpha=0.0001, model__class_weight=balanced, mod
el__eta0=1.0, model__learning_rate=constant;, score=(train=0.996, test=0.996
) total time=
               0.7s
[CV 6/10; 20/30] START model alpha=0.0001, model class weight=balanced, mo
del eta0=10, model learning rate=constant
[CV 6/10; 20/30] END model alpha=0.0001, model class weight=balanced, mode
1 eta0=10, model learning rate=constant;, score=(train=0.991, test=0.990)
total time=
             0.8s
[CV 9/10; 21/30] START model__alpha=0.001, model__class_weight=balanced, mod
el eta0=0.01, model learning rate=constant
[CV 9/10; 21/30] END model alpha=0.001, model class weight=balanced, model
 eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.987)
total time=
             0.8s
[CV 5/10; 23/30] START model_alpha=0.001, model_class_weight=balanced, mod
```

[CV 5/10; 23/30] END model__alpha=0.001, model__class_weight=balanced, model eta0=1.0, model learning rate=constant;, score=(train=0.991, test=0.989)

[CV 1/10; 25/30] START model alpha=0.001, model class weight=balanced, mod

[CV 1/10; 25/30] END model__alpha=0.001, model__class_weight=balanced, model__eta0=10, model__learning_rate=constant;, score=(train=0.990, test=0.992) t

[CV 10/10; 26/30] START model alpha=0.01, model class weight=balanced, mod

[CV 10/10; 26/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=0.01, model__learning_rate=constant;, score=(train=0.991, test=0.990)

[CV 6/10; 28/30] START model alpha=0.01, model class weight=balanced, mode

[CV 6/10; 28/30] END model__alpha=0.01, model__class_weight=balanced, model__eta0=1.0, model__learning_rate=constant;, score=(train=0.991, test=0.990) t

[CV 1/10; 30/30] START model__alpha=0.01, model__class_weight=balanced, mode

el__eta0=1.0, model__learning_rate=constant

el eta0=10, model learning rate=constant

el__eta0=0.01, model__learning_rate=constant

_eta0=1.0, model__learning_rate=constant

0.7s

1.1s

0.7s

0.8s

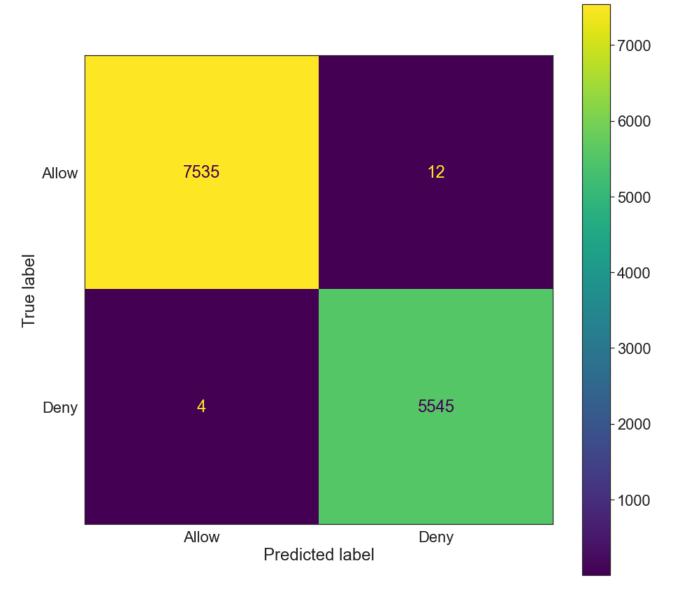
total time=

otal time=

total time=

otal time=

```
l eta0=10, model learning rate=constant
         [CV 1/10; 30/30] END model__alpha=0.01, model__class_weight=balanced, model_
         _eta0=10, model__learning_rate=constant;, score=(train=0.728, test=0.729) to
         tal time=
                   1.1s
Out[38]: | GridSearchCV
         ▶ estimator: SVC
                ▶ SVC
In [39]: # Confirm number of support vectors
         len(grid_search.best_estimator_.n_support_)
Out[39]:
In [40]: # Evaluate the training accuracy of the model
         y_hat_rf_train = grid_search.predict(X_train)
         accuracy_score(y_hat_rf_train, y_train)
         0.9989118399450193
Out[40]:
In [41]: # Build confusion matrix
         confusion_matrix(y_train, y_hat_rf_train)
         disp = ConfusionMatrixDisplay.from_estimator(grid_search, X_test, y_test)
```



```
In [42]: # Evaluate the test accuracy of the model
    y_hat_rf_test = grid_search.predict(X_test)
    accuracy_score(y_hat_rf_test, y_test)
```

Out[42]: 0.9987782529016493

Recall: 0.998409964224195 Precision: 0.997840561454022

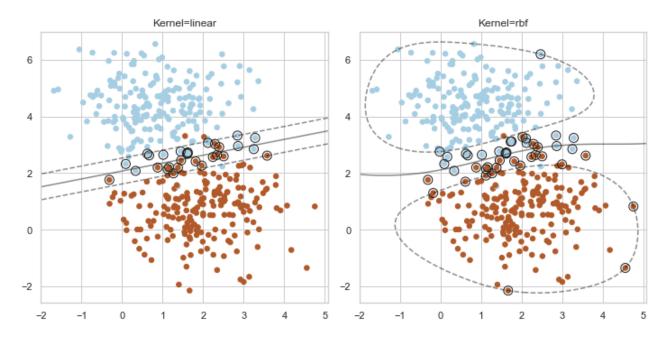
Evaluate Hyperparameters

```
In [44]: p = ['linear', 'rbf']

for i in p:
    model = SVC(kernel = i)
    model.fit(X_train, y_train)
    y_hat_rf_test = model.predict(X_test)
    print(i,":", accuracy_score(y_hat_rf_test, y_test))
```

linear : 0.9987782529016493
rbf : 0.778787416004887

```
In [45]: sns.set(font scale = 1)
          sns.set_style(style = "whitegrid")
          # Prepare test and train data
         X = df.loc[:, df.columns != 'target'].values
         y = df['target'].values
          X_train, X_test, y_train, y_test = train_test_split(
              X, y, test_size = 0.3, random_state = 42)
          X, y = make_blobs(n_samples=400, centers=2, random_state=0)
          plt.figure(figsize=(10, 5))
          for i, K in enumerate(['linear','rbf']):
              clf = SVC(C = 1, kernel= K , random state = 42).fit(X, y)
              decision function = clf.decision function(X)
              support vector indices = np.where(
                  np.abs(decision function) <= 1 + 1e-15)[0]</pre>
              support vectors = X[support vector indices]
              plt.subplot(1, 2, i + 1)
              plt.scatter(X[:, 0],
                          X[:, 1],
                          c = y
                          s = 30,
                          cmap = plt.cm.Paired)
              ax = plt.gca()
              xlim = ax.get_xlim()
              ylim = ax.get_ylim()
              xx, yy = np.meshgrid(np.linspace(xlim[0], xlim[1], 50),
                                   np.linspace(ylim[0], ylim[1], 50))
              Z = clf.decision function(np.c [xx.ravel(), yy.ravel()])
              Z = Z.reshape(xx.shape)
              plt.contour(xx,
                          уу,
                          colors = 'k',
                          levels = [-1, 0, 1],
                          alpha = 0.5,
                          linestyles = ['--', '-', '--'])
              plt.scatter(support_vectors[:, 0],
                          support_vectors[:, 1],
                          s = 100,
                          linewidth = 1,
                          facecolors = 'none',
                          edgecolors = 'k')
              plt.title("Kernel=" + str(K))
          plt.tight layout()
          plt.show()
```

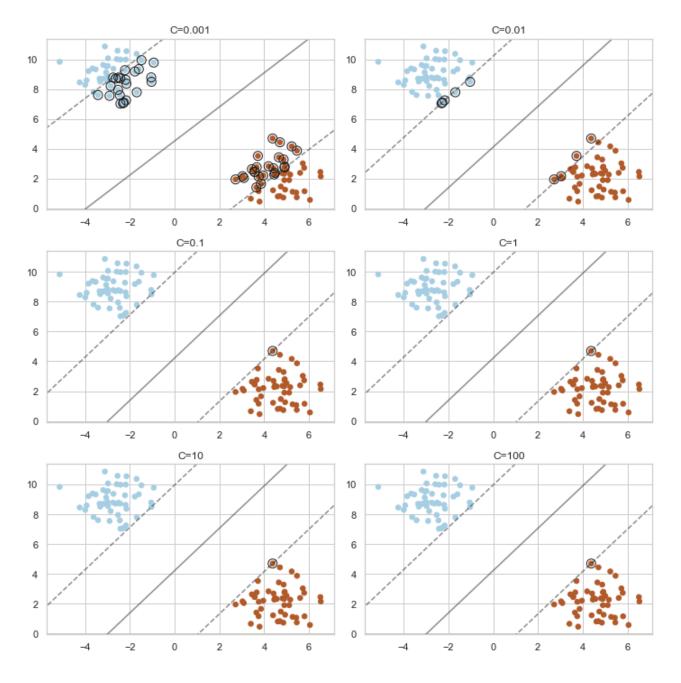


```
In [46]: C = [0.001, 0.01, 0.1, 1, 10, 100]
    for i in C:
        model = SVC(C = i)
        model.fit(X_train, y_train)
        y_hat_rf_test = model.predict(X_test)
        print(str(i), ":", accuracy_score(y_hat_rf_test, y_test))
```

0.001 : 0.574628385257585 0.01 : 0.574628385257585 0.1 : 0.574628385257585 1 : 0.8639788230502953 10 : 0.9350437792710242 100 : 0.9725615964162085

Linear Kernel

```
In [47]: sns.set(font scale = 1)
          sns.set_style(style = "whitegrid")
          X, y = make blobs(n samples = 100,
                            centers = 2,
                            random_state = 42)
         plt.figure(figsize=(10, 10))
          for i, C in enumerate([0.001, 0.01, 0.1, 1, 10, 100]):
              clf = SVC(C = C,
                        kernel = 'linear',
                        random_state = 42).fit(X, y)
              decision_function = clf.decision_function(X)
              support_vector_indices = np.where(
                  np.abs(decision function) <= 1 + 1e-15)[0]</pre>
              support_vectors = X[support_vector_indices]
              plt.subplot(3, 2, i + 1)
              plt.scatter(X[:, 0],
                          X[:, 1],
                          c = y
                          s = 30,
                          cmap = plt.cm.Paired)
              ax = plt.gca()
              xlim = ax.get_xlim()
              ylim = ax.get_ylim()
              xx, yy = np.meshgrid(np.linspace(xlim[0], xlim[1], 50),
                                   np.linspace(ylim[0], ylim[1], 50))
              Z = clf.decision function(np.c [xx.ravel(), yy.ravel()])
              Z = Z.reshape(xx.shape)
              plt.contour(xx,
                          уу,
                          colors = 'k',
                          levels = [-1, 0, 1],
                          alpha = 0.5,
                          linestyles=['--', '-', '--'])
              plt.scatter(support_vectors[:, 0],
                          support_vectors[:, 1],
                          s = 100,
                          linewidth = 1,
                          facecolors = 'none',
                          edgecolors = 'k')
              plt.title("C=" + str(C))
          plt.tight layout()
          plt.show()
```



RBF Kernel

```
In [48]: sns.set(font scale = 1)
          sns.set_style(style = "whitegrid")
          X, y = make blobs(n samples = 100,
                            centers = 2,
                            random_state = 42)
         plt.figure(figsize=(10, 10))
          for i, C in enumerate([0.001, 0.01, 0.1, 1, 10, 100]):
              clf = SVC(C = C,
                        kernel = 'rbf',
                        random_state = 42).fit(X, y)
              decision_function = clf.decision_function(X)
              support_vector_indices = np.where(
                  np.abs(decision function) <= 1 + 1e-15)[0]</pre>
              support_vectors = X[support_vector_indices]
              plt.subplot(3, 2, i + 1)
              plt.scatter(X[:, 0],
                          X[:, 1],
                          c = y
                          s = 30,
                          cmap = plt.cm.Paired)
              ax = plt.gca()
              xlim = ax.get_xlim()
              ylim = ax.get_ylim()
              xx, yy = np.meshgrid(np.linspace(xlim[0], xlim[1], 50),
                                   np.linspace(ylim[0], ylim[1], 50))
              Z = clf.decision function(np.c [xx.ravel(), yy.ravel()])
              Z = Z.reshape(xx.shape)
              plt.contour(xx,
                          уу,
                          colors = 'k',
                          levels = [-1, 0, 1],
                          alpha = 0.5,
                          linestyles=['--', '-', '--'])
              plt.scatter(support_vectors[:, 0],
                          support_vectors[:, 1],
                          s = 100,
                          linewidth = 1,
                          facecolors = 'none',
                          edgecolors = 'k')
              plt.title("C=" + str(C))
          plt.tight layout()
          plt.show()
```

