j_version

October 31, 2022

```
[]: import scipy.io as sio
     import matplotlib.pyplot as plt
     import numpy as np
     import sklearn.discriminant_analysis
     from sklearn import svm
     from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
     from function_plot import Load_mat_single
     from function_plot import mat_to_array
     from function_plot import plot_confusion_matrix
     from function_plot import train_test
     import time
     # Load data
     path_good = 'data//baseline_20220915_sv.mat'
     path_bad = 'data//fault7_20220915_sv.mat'
     mat_contents_good = Load_mat_single(path_good)
     good_data = mat_to_array(mat_contents_good)
     mat_contents_bad = Load_mat_single(path_bad)
     bad_data = mat_to_array(mat_contents_bad)
     show_time = True
     # constuct the data
     X = np.concatenate((good_data, bad_data))
     n_sample = good_data.shape[0]
     n_feature = good_data.shape[1]
     Y = np.zeros(n_sample)
     Y = np.concatenate((Y, np.ones(n_sample)))
     # PCA
```

```
print("Start PCA process...")
X_{mean} = X - np.mean(X)
C_x = np.dot(X_mean.T, X_mean)
SS pca, V = np.linalg.eig(C x)
# from large to small
sortIndex = np.flip(np.argsort(SS_pca))
dimension = good_data.shape[1]
VSorted = np.empty((dimension, 0))
for i in range(dimension):
   VSorted = np.append(
        VSorted, V[:, sortIndex[i]].reshape(dimension, 1), axis=1)
classificationError_lda_pca = np.zeros(5,)
classificationError_svm_pca = np.zeros(5,)
Score_Sorted = np.dot(X, VSorted)
train_index = np.arange(0, n_sample*0.75).astype(int).tolist() + \
   np.arange(n_sample, n_sample+n_sample*0.75).astype(int).tolist()
test_index = np.arange(n_sample*0.75, n_sample).astype(int).tolist() + \
   np.arange(n_sample+n_sample*0.75, n_sample+n_sample).astype(int).tolist()
# data for training and testing
X_train = Score_Sorted[train_index, :]
X_test = Score_Sorted[test_index, :]
Y_train = Y[train_index]
Y_test = Y[test_index]
# Classifier 1 LDA with PCA
for numDims in range(4, 9):
   Score_Reduced = X_train[:, 0:numDims]
```

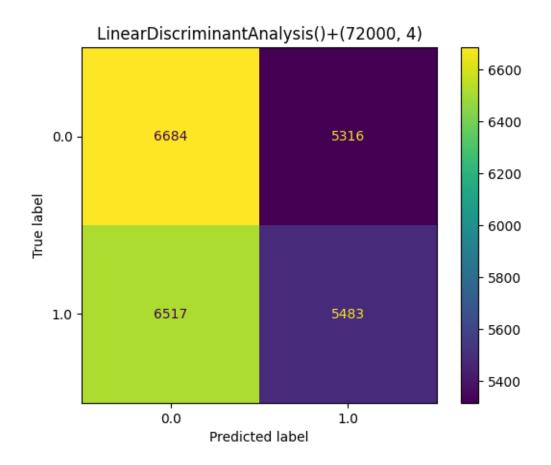
```
lda_pca = sklearn.discriminant_analysis.LinearDiscriminantAnalysis()
   X_test_temp = X_test[:, 0:numDims]
   error, prediction_lda_pca = train_test(
       Score_Reduced, Y_train, X_test_temp, Y_test, lda_pca, show_time)
   classificationError_lda_pca[numDims-4] = error
   print("====== Confusion matrix for LDA with PCA, Reduced score shape is ⊔
 Score_Reduced.shape))
   plot_confusion_matrix(Y_test, prediction_lda_pca, lda_pca, Score_Reduced)
# Classifier 2 SVM with PCA
for numDims in range(4, 9):
   Score_Reduced = X_train[:, 0:numDims]
   clf_svm_pca = svm.SVC(kernel='linear')
   X_test_temp = X_test[:, 0:numDims]
   error, prediction_svm_pca = train_test(
       Score_Reduced, Y_train, X_test_temp, Y_test, clf_svm_pca, show_time)
   classificationError_svm_pca[numDims-4] = error
   print("====== Confusion matrix for SVM with PCA, Reduced score shape is ⊔
 Score Reduced.shape))
   plot_confusion_matrix(Y_test, prediction_svm_pca, lda_pca, Score_Reduced)
# # Feature selection-backward search
print("Start Feature Selection process...")
X_train_fs = X[train_index, :]
X_test_fs = X[test_index, :]
n_train = X_train_fs.shape[0]
n_test = X_test_fs.shape[0]
```

```
final_dimension = 5
# Classifier 1 LDA with Feature selection
print('Start Feature Selection with LDA...')
removed = []
index_all = [0, 1, 2, 3, 4, 5, 6, 7]
remaining = index_all[:]
classificationError_lda_fs = n_test*np.ones(final_dimension)
lda_fs = sklearn.discriminant_analysis.LinearDiscriminantAnalysis()
error_temp, prediction = train_test(
   X_train_fs, Y_train, X_test_fs, Y_test, lda_fs, show_time)
classificationError_lda_fs[0] = error_temp
print("====== Confusion matrix for LDA with FS, the training shape is {}⊔
 ⇒======= ".format(
   X_train_fs.shape))
plot_confusion_matrix(Y_test, prediction, lda_fs, X_train_fs)
for iteration in range(final_dimension-1):
   error_inside = n_test*np.ones(n_feature)
   for idx, item in enumerate(remaining):
       temp_removed = removed[:]
       temp removed.append(item)
        # print(temp_removed)
       Xtrain_temp = np.delete(X_train_fs, temp_removed, axis=1)
       Xtest_temp = np.delete(X_test_fs, temp_removed, axis=1)
       lda_temp = sklearn.discriminant_analysis.LinearDiscriminantAnalysis()
        error_temp, item_ignored = train_test(Xtrain_temp, Y_train,
                                              Xtest_temp, Y_test, lda_temp)
        error_inside[idx] = error_temp
   worst_i = np.argmin(error_inside)
```

```
worst_item = remaining[worst_i]
   removed.append(worst_item)
   print("The removed colomns", removed)
   remaining.remove(worst_item)
   print("========= The remained colomns {} =========\n".
 →format(remaining))
    # calculate best remaining
   X_train_selection = np.delete(X_train_fs, removed, 1)
   X_test_selection = np.delete(X_test_fs, removed, 1)
   lda_fs = sklearn.discriminant_analysis.LinearDiscriminantAnalysis()
   error, prediction_lda_fs = train_test(X_train_selection, Y_train,
                                         X_test_selection, Y_test, lda_fs,_
 ⇒show_time)
   classificationError_lda_fs[iteration+1] = error
   print("====== Confusion matrix for LDA with FS, the training shape is {}⊔
 ⇒======== ".format(
       X_train_selection.shape))
   plot_confusion_matrix(Y_test, prediction_lda_fs, lda_fs, X_train_selection)
# Classifier 2 SVM with Feature selection
print('Start Feature Selection with SVM...')
classificationError_svm_fs = n_test*np.ones(final_dimension)
removed = []
remaining = index_all[:]
clf_svm_fs = svm.SVC(kernel='linear')
error_temp, prediction = train_test(
   X_train_fs, Y_train, X_test_fs, Y_test, clf_svm_fs)
classificationError_svm_fs[0] = error_temp
```

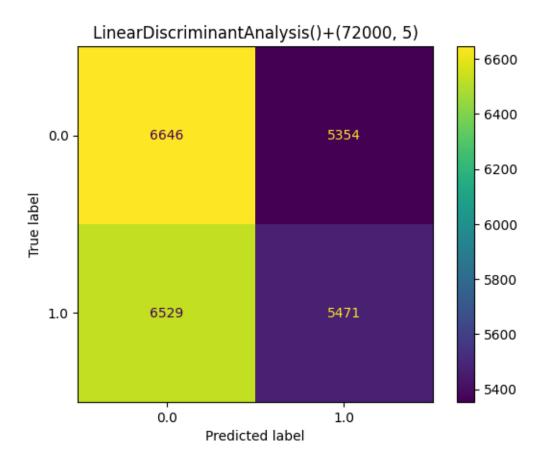
```
print("====== Confusion matrix for SVM with FS, the training shape is {}_{\sqcup}
 →======= ".format(
   X_train_fs.shape))
plot_confusion_matrix(Y_test, prediction, clf_svm_fs, X_train_fs)
for iteration in range(final_dimension-1):
   error_inside = n_test*np.ones(n_feature)
   for idx, item in enumerate(remaining):
       temp_removed = removed[:]
       temp_removed.append(item)
        # print(temp_removed)
       Xtrain_temp = np.delete(X_train_fs, temp_removed, axis=1)
       Xtest_temp = np.delete(X_test_fs, temp_removed, axis=1)
       svm_temp = svm.SVC(kernel='linear')
        error_temp, item_ignored = train_test(Xtrain_temp, Y_train,
                                              Xtest_temp, Y_test, svm_temp)
        error_inside[idx] = error_temp
   worst_i = np.argmin(error_inside)
   worst_item = remaining[worst_i]
   removed.append(worst_item)
   print("The removed colomns", removed)
   remaining.remove(worst_item)
   print("The remained colomns {}\n".format(remaining))
   X_train_selection = np.delete(X_train_fs, removed, 1)
   X_test_selection = np.delete(X_test_fs, removed, 1)
   svm fs = svm.SVC(kernel='linear')
   error, prediction_svm_fs = train_test(X_train_selection, Y_train,
                                          X_test_selection, Y_test, svm_fs,
 ⇒show_time)
   classificationError_svm_fs[iteration+1] = error
```

```
print("====== Confusion matrix for SVM with FS, the training shape is {}⊔
  ⇒======= ".format(
        X_train_selection.shape))
    plot_confusion_matrix(Y_test, prediction_svm_fs, svm_fs, X_train_selection)
plt.figure()
plt.scatter([8, 7, 6, 5, 4], np.flip(classificationError_lda_pca),
            c='b', marker='*', label="PCA+LDA")
plt.scatter([8, 7, 6, 5, 4], classificationError_lda_fs,
            c='r', marker='o', label="Feature Selection+LDA")
plt.xlabel('Dimension')
plt.ylabel('Error')
plt.title('Classifier 1 LDA')
plt.legend()
plt.figure()
plt.scatter([8, 7, 6, 5, 4], np.flip(classificationError_svm_pca),
            c='b', marker='*', label="PCA+SVM")
plt.scatter([8, 7, 6, 5, 4], classificationError_svm_fs,
            c='r', marker='o', label="Feature Selection+SVM")
plt.xlabel('Dimension')
plt.ylabel('Error')
plt.title('Classifier 2 SVM')
plt.legend()
0 header
1 __version__
2 __globals__
3 sv
0 __header__
1 __version__
2 __globals__
3 sv
Start PCA process...
The experiment is LinearDiscriminantAnalysis()
The shape of X train is (72000, 4)
The train time is --- 0.04088902 seconds ---
The test time is --- 0.00099659 seconds ---
====== Confusion matrix for LDA with PCA, Reduced score shape is (72000, 4)
========
```



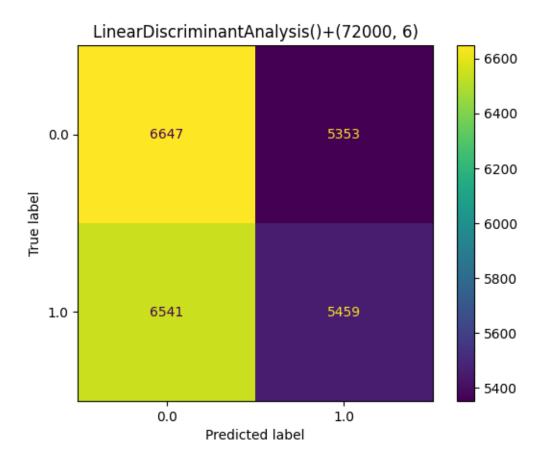
The shape of X_{train} is (72000, 5)

The train time is --- 0.04886866 seconds --The test time is --- 0.00099754 seconds --======= Confusion matrix for LDA with PCA, Reduced score shape is (72000, 5)
========



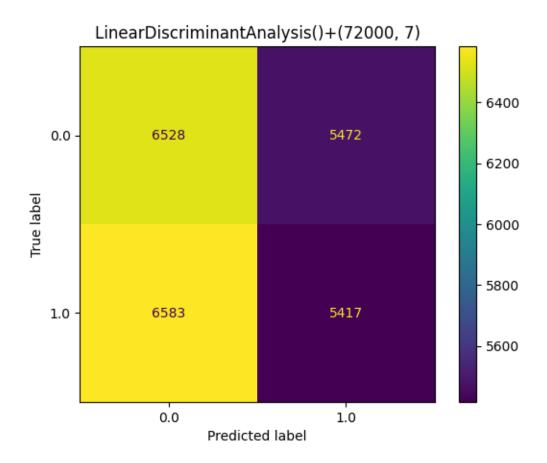
The shape of X_{train} is (72000, 6)

The train time is --- 0.06083798 seconds --The test time is --- 0.00096536 seconds --======= Confusion matrix for LDA with PCA, Reduced score shape is (72000, 6)
========



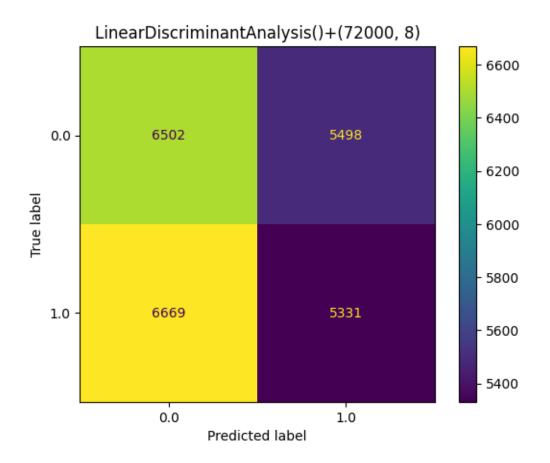
The shape of X_{train} is (72000, 7)

The train time is --- 0.06881666 seconds --The test time is --- 0.00099635 seconds --======= Confusion matrix for LDA with PCA, Reduced score shape is (72000, 7)
========



The shape of X_{train} is (72000, 8)

The train time is --- 0.07682896 seconds --The test time is --- 0.00101137 seconds --======= Confusion matrix for LDA with PCA, Reduced score shape is (72000, 8)
========



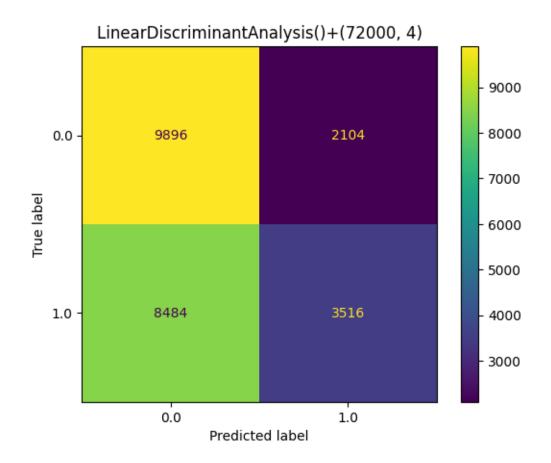
```
The experiment is SVC(kernel='linear')

The shape of X_train is (72000, 4)

The train time is --- 101.14606404 seconds ---

The test time is --- 24.38094330 seconds ---

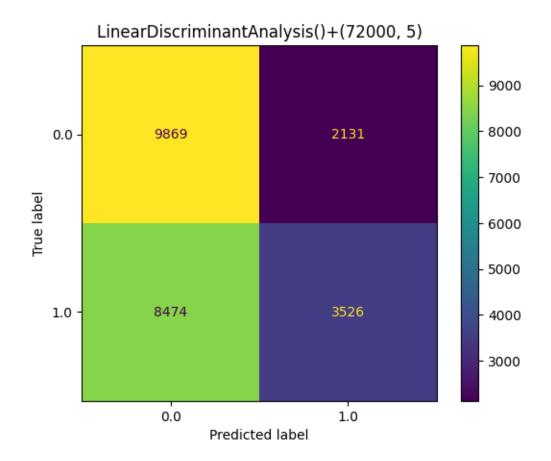
======== Confusion matrix for SVM with PCA, Reduced score shape is (72000, 4)
```



The experiment is SVC(kernel='linear')

The shape of X_train is (72000, 5)

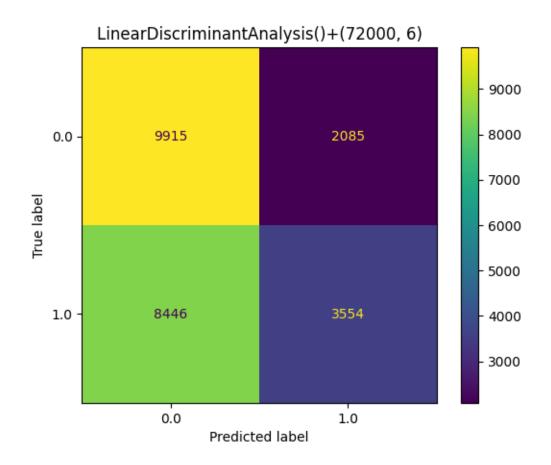
The train time is --- 102.90769792 seconds --
The test time is --- 25.33850074 seconds --
======== Confusion matrix for SVM with PCA, Reduced score shape is (72000, 5)



The experiment is SVC(kernel='linear')

The shape of X_train is (72000, 6)

The train time is --- 105.95546889 seconds --
The test time is --- 25.70138359 seconds --
======== Confusion matrix for SVM with PCA, Reduced score shape is (72000, 6)



```
The experiment is SVC(kernel='linear')

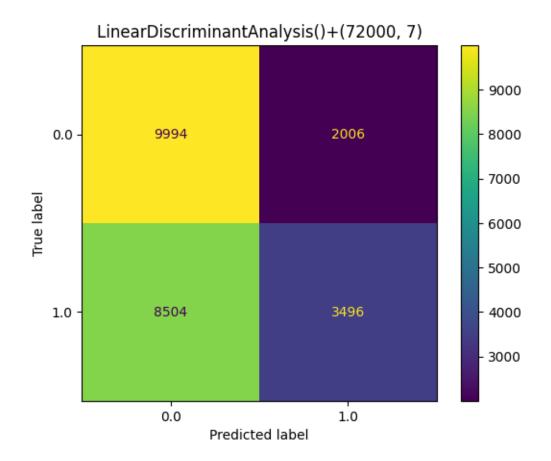
The shape of X_train is (72000, 7)

The train time is --- 115.95480776 seconds ---

The test time is --- 27.22708750 seconds ---

======== Confusion matrix for SVM with PCA, Reduced score shape is (72000, 7)

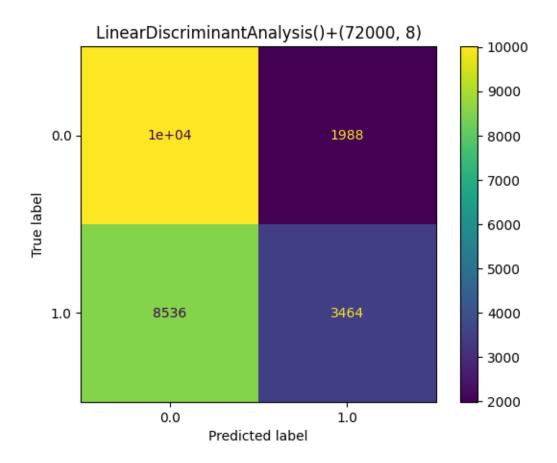
=========
```



The experiment is SVC(kernel='linear')

The shape of X_train is (72000, 8)

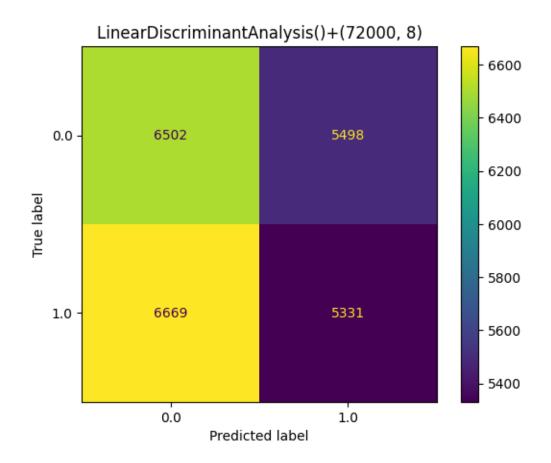
The train time is --- 116.09141397 seconds --
The test time is --- 27.43732071 seconds --
======== Confusion matrix for SVM with PCA, Reduced score shape is (72000, 8)

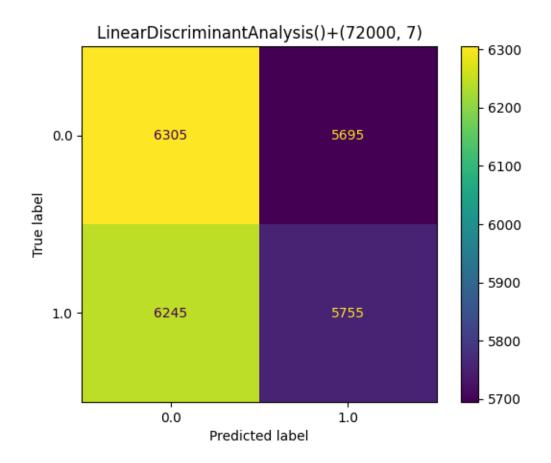


Start Feature Selection process...
Start Feature Selection with LDA...
The experiment is LinearDiscriminantAnalysis()

The shape of X_{train} is (72000, 8)

The train time is --- 0.07081032 seconds --The test time is --- 0.00099874 seconds --======= Confusion matrix for LDA with FS, the training shape is (72000, 8)

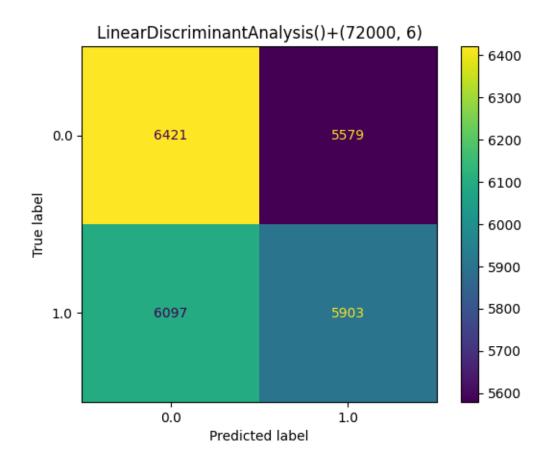




The experiment is LinearDiscriminantAnalysis()

The shape of X_train is (72000, 6)

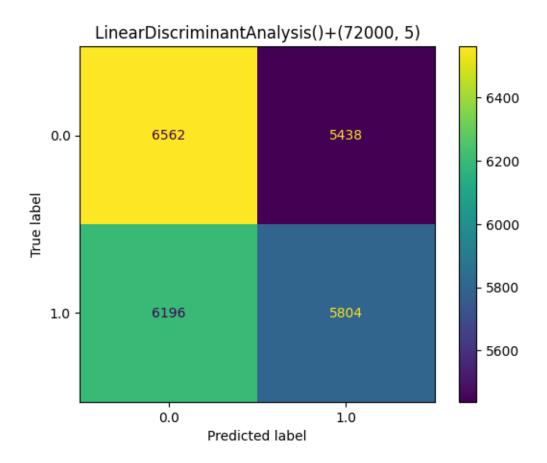
The train time is --- 0.05185819 seconds --- The test time is --- 0.00100064 seconds --- ======= Confusion matrix for LDA with FS, the training shape is (72000, 6) ==========



The experiment is LinearDiscriminantAnalysis()

The shape of X_train is (72000, 5)

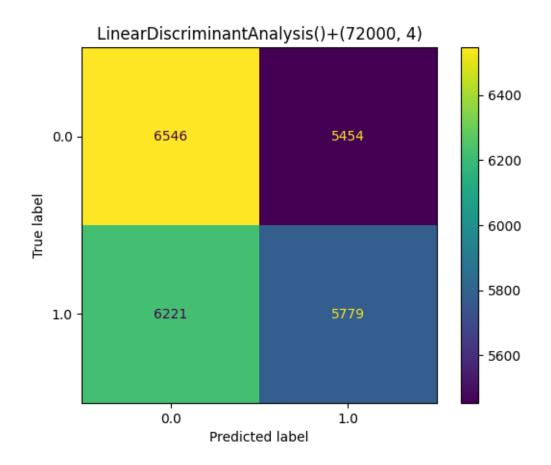
The train time is --- 0.04687524 seconds --- The test time is --- 0.00099683 seconds --- ======= Confusion matrix for LDA with FS, the training shape is (72000, 5) =========



The experiment is LinearDiscriminantAnalysis()

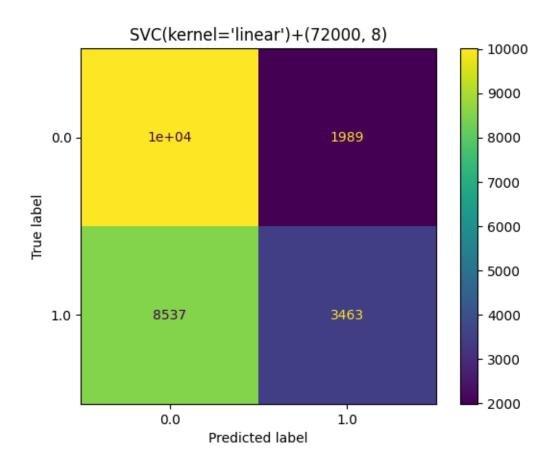
The shape of X_train is (72000, 4)

The train time is --- 0.03690171 seconds --- The test time is --- 0.00102901 seconds ---

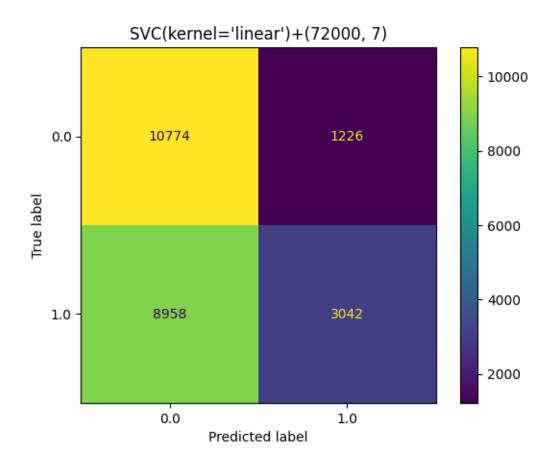


Start Feature Selection with SVM...

======= Confusion matrix for SVM with FS, the training shape is (72000, 8)



```
The removed colomns [3]
The remained colomns [0, 1, 2, 4, 5, 6, 7]
The experiment is SVC(kernel='linear')
The shape of X_train is (72000, 7)
The train time is --- 109.88118410 seconds ---
The test time is --- 27.06184483 seconds ---
========= Confusion matrix for SVM with FS, the training shape is (72000, 7)
========
```

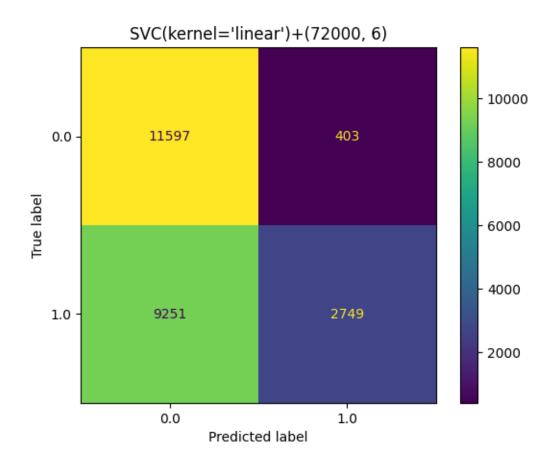


```
The removed colomns [3, 4]
The remained colomns [0, 1, 2, 5, 6, 7]

The experiment is SVC(kernel='linear')

The shape of X_train is (72000, 6)

The train time is --- 105.73561668 seconds ---
The test time is --- 25.08102965 seconds ---
========= Confusion matrix for SVM with FS, the training shape is (72000, 6)
=========
```

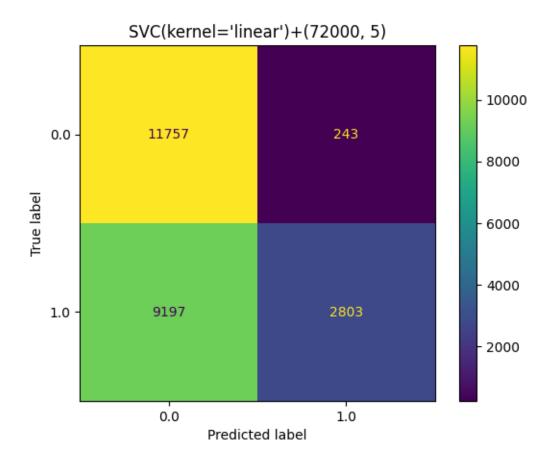


```
The removed colomns [3, 4, 1]
The remained colomns [0, 2, 5, 6, 7]

The experiment is SVC(kernel='linear')

The shape of X_train is (72000, 5)

The train time is --- 104.75766206 seconds ---
The test time is --- 25.33646107 seconds ---
========= Confusion matrix for SVM with FS, the training shape is (72000, 5)
=========
```

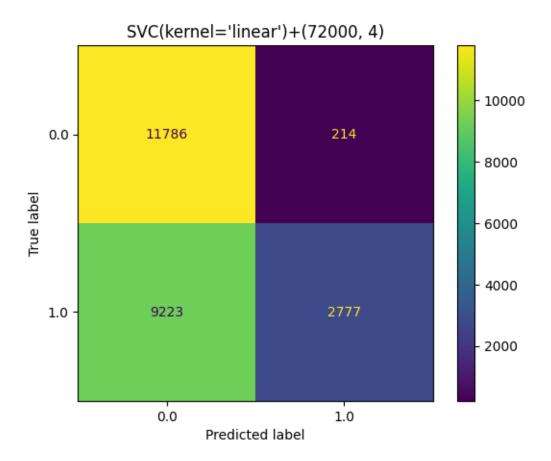


```
The removed colomns [3, 4, 1, 0]
The remained colomns [2, 5, 6, 7]

The experiment is SVC(kernel='linear')

The shape of X_train is (72000, 4)

The train time is --- 143.62490559 seconds ---
The test time is --- 24.23975325 seconds ---
========= Confusion matrix for SVM with FS, the training shape is (72000, 4)
=========
```



[]: <matplotlib.legend.Legend at 0x1d79b6c2040>

