## Carga Librerias

```
In [1]: # Librerias
         import warnings
         import pandas as pd
         import numpy as np
         import matplotlib
         import matplotlib.pyplot as plt
         import plotly.express as px
         import seaborn as sns
         from sklearn.preprocessing import MinMaxScaler
         from sklearn.model selection import train test split, GridSearchCV
         from sklearn.decomposition import PCA
         from sklearn import svm
         from sklearn.linear model import LogisticRegression
         from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier
         from sklearn.tree import DecisionTreeClassifier
         from xgboost import XGBClassifier
         from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay, accuracy_score, precision_score, recall_s
         warnings.filterwarnings('ignore')
In [2]: def plot_metrics(model, x, y, pred_y):
             print('Accuracy:\t', accuracy_score(y, pred_y))
print('Precision:\t', precision_score(y, pred_y))
print('Recall:\t\t', recall_score(y, pred_y))
print('F1 score:\t', f1_score(y, pred_y))
              print('ROC AUC score:\t', roc_auc_score(y, model.predict_proba(x)[:, 1]))
              fig, ax = plt.subplots(1)
              matplotlib.rc('figure', figsize=(15, 5))
              ax.plot(y)
              ax.plot(pred_y)
              plot_roc_curve(model, x, y)
```

# Carga Datos

```
DATA_DIR = "C:/Users/NetRunner/OneDrive/UOC/Semestre 6/TFM/MultipleDatasets"
In [3]:
        train data cut = pd.read csv(f"{DATA DIR}/train data.csv")
        test_data_cut = pd.read_csv(f"{DATA_DIR}/test_data.csv")
        train data uncut = pd.read csv(f"{DATA DIR}/train data uncut.csv")
        test data uncut = pd.read csv(f"{DATA DIR}/test data uncut.csv")
        # X_train = pd.read_csv(f"{DATA_DIR}/X_train.csv")
        # y_train = pd.read_csv(f"{DATA_DIR}/y_train.csv")
# X_test = pd.read_csv(f"{DATA_DIR}/X_test.csv")
        # y_test = pd.read_csv(f"{DATA_DIR}/y_test.csv")
       data_cut = pd.concat([train_data_cut, test_data_cut])
        data_uncut = pd.concat([train_data_uncut, test_data_uncut])
        'error1_count', 'error2_count', 'error3_count', 'error4_count',
                  'error5 count']
        label = ['State']
        data_cut = data_cut[features+label]
        data uncut = data uncut[features+label]
```

#### Normalización de datos MinMax

```
In [5]: feature_scaler_cut = MinMaxScaler(feature_range=(0,1))
    label_scaler_cut = MinMaxScaler(feature_range=(0,1))
    feature_scaler_cut.fit(data_cut[features])
    label_scaler_cut.fit(data_cut[label].values.reshape(-1,1))
Out[5]: MinMaxScaler()
```

#### Dataset Train/Test

```
In [8]: train norm cut = data norm cut[:len(train data cut)]
         test_norm_cut = data_norm_cut[len(train_data_cut):(len(train_data_cut)+len(test_data_cut))]
         X_train_cut = train_norm_cut.loc[:, train_norm_cut.columns != 'State']
y_train_cut = train_norm_cut['State'].values
         X test cut = test norm cut.loc[:, test norm cut.columns != 'State']
         y test_cut = test_norm_cut['State'].astype(int)
         print('X_train:\t', X_train_cut.shape)
         print('y_train:\t', y_train_cut.shape)
print('X_test:\t\t', X_test_cut.shape)
print('y_test:\t\t', y_test_cut.shape)
                             (17280, 22)
         X train:
         y_train:
                             (17280,)
         X test:
                             (3600, 22)
                             (3600,)
         y test:
In [9]: train_norm_uncut = data_norm_uncut[:len(train_data_uncut)]
         test norm uncut = data norm uncut[len(train data uncut):(len(train data uncut)+len(test data uncut))]
         X_train_uncut = train_norm_uncut.loc[:, train_norm_uncut.columns != 'State']
         y train uncut = train norm uncut['State'].astype(int)
         X test uncut = test norm uncut.loc[:, test norm uncut.columns != 'State']
         y_test_uncut = test_norm_uncut['State'].astype(int)
         print('X_train_uncut:\t', X_train_uncut.shape)
         print('y_train_uncut:\t', y_train_uncut.shape)
print('X_test_uncut:\t', X_test_uncut.shape)
         print('y test uncut:\t', y test uncut.shape)
         X train uncut:
                             (40397, 22)
         y train uncut:
                             (40397,)
                             (8305, 22)
         X test uncut:
                             (8305,)
         y test uncut:
```

### Train/Test variable control

```
In [10]: # Cut cycles
    X_train = X_train_cut
    y_train = y_train_cut
    X_test = X_test_cut
    y_test = y_test_cut

# Uncut cycles
# X_train = X_train_uncut
# y_train = y_train_uncut
# X_test = X_test_uncut
# y_test = y_test_uncut
```

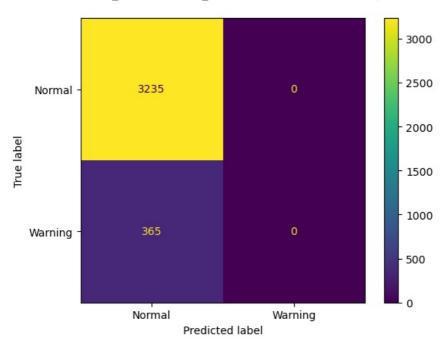
# Logistic Regression

```
In [11]: parameters = {'solver':['newton-cg', 'lbfgs', 'liblinear'],'C':[0.001, 0.01, 0.1, 1, 10]}
    grid_search = GridSearchCV(LogisticRegression(), parameters, verbose=0)
    grid_search.fit(X_train, y_train)
    grid_df = pd.DataFrame(grid_search.cv_results_['params'])
    grid_df['mean_test_score'] = grid_search.cv_results_['mean_test_score']
    grid_df['rank'] = grid_search.cv_results_['rank_test_score']
    grid_df.sort_values(by='rank').head()
```

```
solver mean_test_score rank
Out[11]:
           3 0.010 newton-cg
                                       0.899884
           4 0.010
                         lbfgs
                                       0.899884
           5 0.010
                                       0.898958
                                                    3
                       liblinear
           0 0.001 newton-cg
                                       0.898611
                                                    4
           1 0.001
                         lbfgs
                                       0.898611
```

```
In [12]: model_lr = LogisticRegression(solver='newton-cg', C=0.01).fit(X_train, y_train)
    pred_lr = model_lr.predict(X_test)
    lr_cm = confusion_matrix(y_test, pred_lr)
    ConfusionMatrixDisplay(lr_cm, display_labels=['Normal', 'Warning']).plot()
```

~sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x1e09a866190>

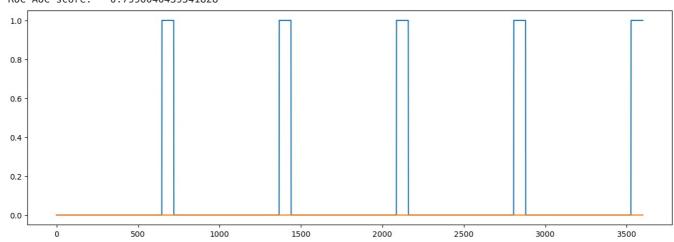


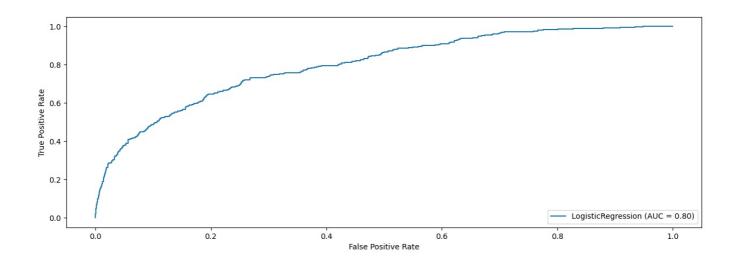
#### In [14]: plot\_metrics(model\_lr, X\_test, y\_test, pred\_lr)

Accuracy: 0.898611111111111

Precision: 0.0 Recall: 0.0 F1 score: 0.0

ROC AUC score: 0.7990040439541828

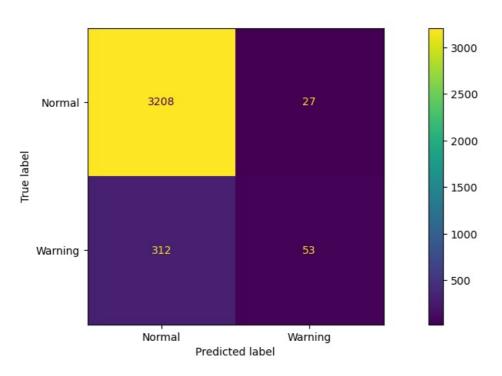




### **SVM**

```
In [14]: parameters = {'kernel': ['poly', 'rbf', 'sigmoid'], 'C':[0.001, 0.01, 0.1, 1, 10]}
           grid_search = GridSearchCV(svm.SVC(), parameters, verbose=0)
           grid search.fit(X train, y train)
           grid_df = pd.DataFrame(grid_search.cv_results_['params'])
grid_df['mean_test_score'] = grid_search.cv_results_['mean_test_score']
grid_df['rank'] = grid_search.cv_results_['rank_test_score']
           grid_df.head().sort_values(by='rank').head()
                     kernel mean_test_score rank
Out[14]:
           3 0.010
                                     0.904398
                        poly
                                                 1
           0.001
                        poly
                                     0.899826
                                                 2
           1 0.001
                                     0.898611
                                                 3
                         rbf
           2 0.001 sigmoid
                                     0.898611
                                                 3
           4 0.010
                                     0.898611
                                                 3
In [15]:
           model_svm = svm.SVC(kernel='poly', C=0.01, probability=True).fit(X_train, y_train)
           pred_svm = model_svm.predict(X_test)
           svm_cm = confusion_matrix(y_test, pred_svm)
           ConfusionMatrixDisplay(svm_cm, display_labels=['Normal', 'Warning']).plot()
```

Out[15]: <sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x1e09d30faf0>

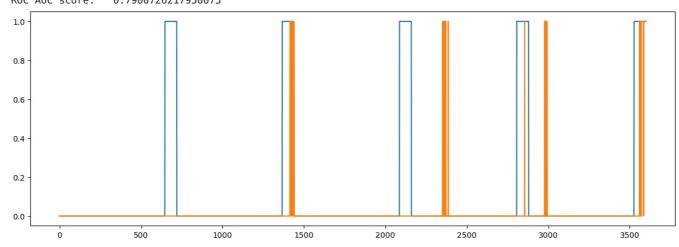


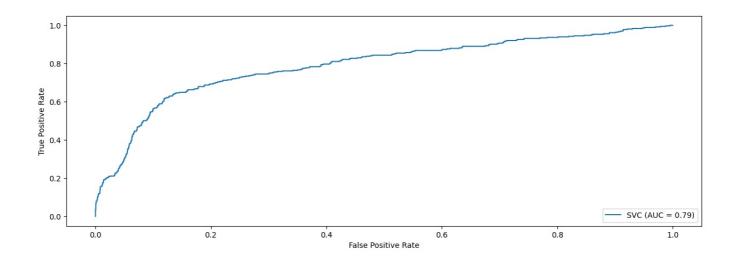
In [16]: plot\_metrics(model\_svm, X\_test, y\_test, pred\_svm)

0.9058333333333334 0.6625

Accuracy: Precision:

Recall: 0.14520547945205478 F1 score: 0.23820224719101124 ROC AUC score: 0.7900726217950075





## Random Forest Classifier

```
In [17]: parameters = {'max_depth': [6,7,8,9,10,11,12], 'n_estimators': [10,50,100,200]}
    grid_search = GridSearchCV(RandomForestClassifier(), parameters, verbose=0)
    grid_search.fit(X_train, y_train)
    grid_df = pd.DataFrame(grid_search.cv_results_['params'])
    grid_df['mean_test_score'] = grid_search.cv_results_['mean_test_score']
    grid_df['rank'] = grid_search.cv_results_['rank_test_score']
    grid_df.sort_values(by='rank').head()
```

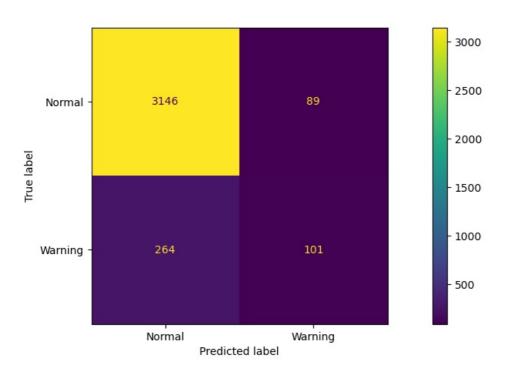
Out[17]:		max_depth	n_estimators	mean_test_score	rank
	0	6	10	0.883275	1
	2	6	100	0.882986	2
	7	7	200	0.881539	3
	3	6	200	0.881366	4

100

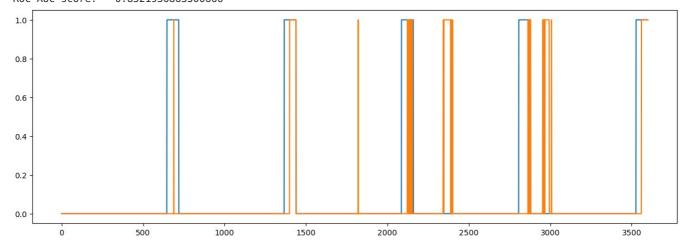
```
In [17]: model_rf = RandomForestClassifier(max_depth=6, n_estimators=10).fit(X_train, y_train)
    pred_rf = model_rf.predict(X_test)
    rf_cm = confusion_matrix(y_test, pred_rf)
    ConfusionMatrixDisplay(rf_cm, display_labels=['Normal', 'Warning']).plot()
```

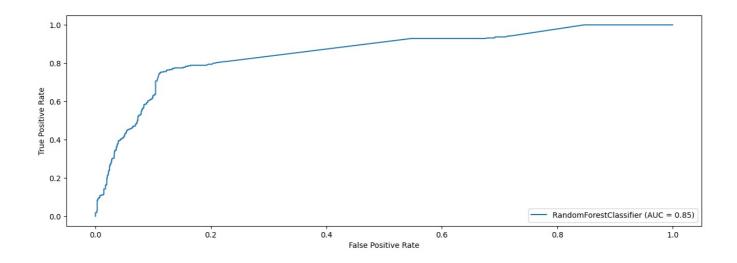
Out[17]: <sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x1e09d406b50>

0.879919



In [18]: plot\_metrics(model\_rf, X\_test, y\_test, pred\_rf)





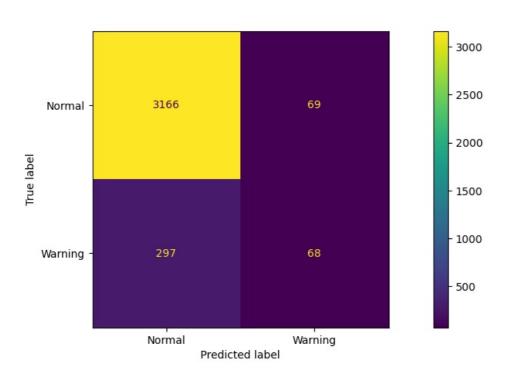
# **Gradient boosting**

```
In [20]: parameters = {'learning_rate': [0.001, 0.01,0.1,1,2], 'n_estimators': [10,50,100,200]}
    grid_search = GridSearchCV(GradientBoostingClassifier(), parameters, verbose=0)
    grid_search.fit(X_train, y_train)
    grid_df = pd.DataFrame(grid_search.cv_results_['params'])
    grid_df['mean_test_score'] = grid_search.cv_results_['mean_test_score']
    grid_df['rank'] = grid_search.cv_results_['rank_test_score']
    grid_df.sort_values(by='rank').head()
```

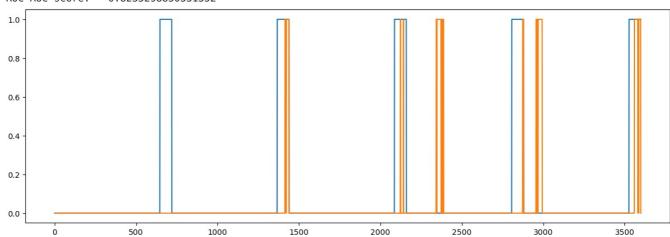
Out[20]:		learning_rate	n_estimators	mean_test_score	rank
	6	0.010	100	0.908275	1
	8	0.100	10	0.906192	2
	7	0.010	200	0.903009	3
	0	0.001	10	0.898611	4
	1	0.001	50	0.898611	4

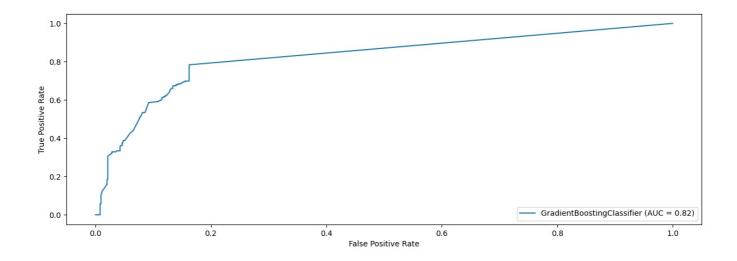
```
In [21]: model_gb = GradientBoostingClassifier(learning_rate=0.010, n_estimators=100).fit(X_train, y_train)
    pred_gb = model_gb.predict(X_test)
    gb_cm = confusion_matrix(y_test, pred_gb)
    ConfusionMatrixDisplay(gb_cm, display_labels=['Normal', 'Warning']).plot()
```

Out[21]: <sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x1e09e1c4dc0>



In [22]: plot\_metrics(model\_gb, X\_test, y\_test, pred\_gb)





# **XGBoosting**

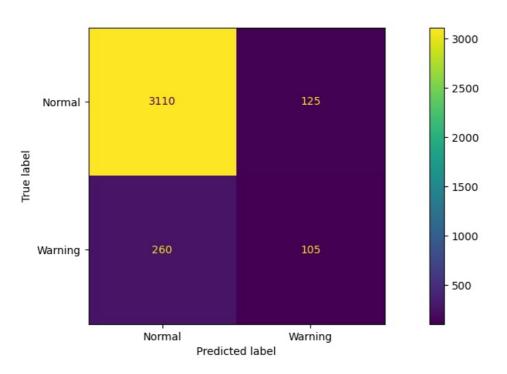
0u

```
parameters = {'booster': ['gbtree', 'dart'], 'reg_alpha': [0, 0.5, 1, 5], 'reg_lambda': [0, 0.5, 1, 5], 'eval_m
    grid_search = GridSearchCV(XGBClassifier(use_label_encoder=False), parameters, verbose=0)
    grid_search.fit(X_train, y_train.astype(int))
    grid_df = pd.DataFrame(grid_search.cv_results_['params'])
    grid_df['mean_test_score'] = grid_search.cv_results_['mean_test_score']
    grid_df['rank'] = grid_search.cv_results_['rank_test_score']
    grid_df.sort_values(by='rank').head()
```

t[23]:		booster	eval_metric	reg_alpha	reg_lambda	mean_test_score	rank
	22	dart	mlogloss	0.5	1.0	0.874306	1
	6	gbtree	mlogloss	0.5	1.0	0.874306	1
	27	dart	mlogloss	1.0	5.0	0.868229	3
	11	gbtree	mlogloss	1.0	5.0	0.868229	3
	28	dart	mlogloss	5.0	0.0	0.867650	5

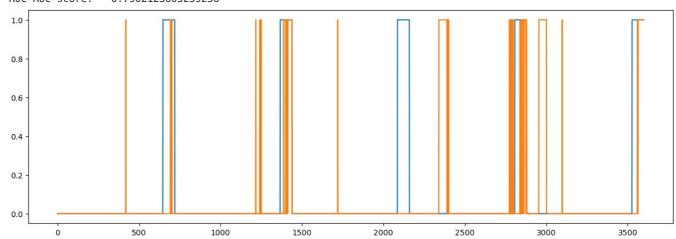
```
In [23]: model_xgb = XGBClassifier(booster='dart', reg_alpha=0.5, reg_lambda=1, eval_metric='mlogloss').fit(X_train, y_t
    pred_xgb = model_xgb.predict(X_test)
    xgb_cm = confusion_matrix(y_test, pred_xgb)
    ConfusionMatrixDisplay(xgb_cm, display_labels=['Normal', 'Warning']).plot()
```

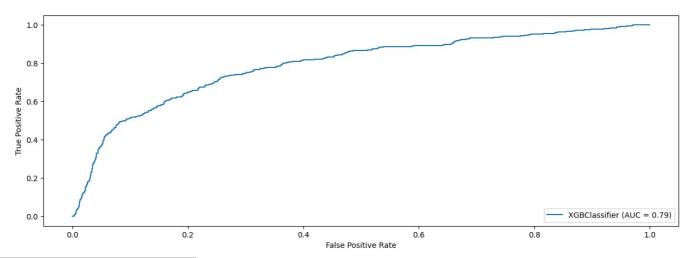
Out[23]: <sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x1e09e3d7d60>



In [24]: plot\_metrics(model\_xgb, X\_test, y\_test, pred\_xgb)

Accuracy: 0.893055555555556
Precision: 0.45652173913043476
Recall: 0.2876712328767123
F1 score: 0.3529411764705882
ROC AUC score: 0.7902123605259256





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