

# Pycaret is a low-code python library for ML

## 1 Imports

```
In [1]: import pandas as pd  
from pycaret.classification import *
```

## 2 Load your data (example)

```
In [3]: data = pd.read_csv("Thyroid_Diff.csv")
```

## 3 Setup the PyCaret experiment

```
In [5]: clf_setup = setup(  
    data=data,  
    target='Recurred',      # or your target column name  
    session_id=123,  
    fold=5  
)
```

	Description	Value
<b>0</b>	Session id	123
<b>1</b>	Target	Recurrent
<b>2</b>	Target type	Binary
<b>3</b>	Target mapping	No: 0, Yes: 1
<b>4</b>	Original data shape	(383, 17)
<b>5</b>	Transformed data shape	(383, 50)
<b>6</b>	Transformed train set shape	(268, 50)
<b>7</b>	Transformed test set shape	(115, 50)
<b>8</b>	Numeric features	1
<b>9</b>	Categorical features	15
<b>10</b>	Preprocess	True
<b>11</b>	Imputation type	simple
<b>12</b>	Numeric imputation	mean
<b>13</b>	Categorical imputation	mode
<b>14</b>	Maximum one-hot encoding	25
<b>15</b>	Encoding method	None
<b>16</b>	Fold Generator	StratifiedKFold
<b>17</b>	Fold Number	5
<b>18</b>	CPU Jobs	-1
<b>19</b>	Use GPU	False
<b>20</b>	Log Experiment	False
<b>21</b>	Experiment Name	clf-default-name
<b>22</b>	USI	bfa0

#### 4 Find the best model and store it in `best_model`

```
In [21]: from pycaret.classification import compare_models, tune_model, save_model

# 1) Find the best baseline model (already have this)
best_model = compare_models(sort='AUC')

# 2) Tune that best model, still optimizing for AUC
best_tuned_model = tune_model(best_model, optimize='AUC')
```

```
# 3) save the tuned best model as a .pkl file
save_model(best_tuned_model, 'best_tuned_thyroid_model')
```

	Model	Accuracy	AUC	Recall	Prec.	F1	Kappa	MCC	TT (Sec)
<b>lightgbm</b>	Light Gradient Boosting Machine	0.9626	0.9941	0.9626	0.9632	0.9622	0.9058	0.9074	0.3100
<b>rf</b>	Random Forest Classifier	0.9551	0.9907	0.9551	0.9552	0.9546	0.8868	0.8880	0.3360
<b>et</b>	Extra Trees Classifier	0.9551	0.9893	0.9551	0.9558	0.9546	0.8869	0.8889	0.3020
<b>gbc</b>	Gradient Boosting Classifier	0.9477	0.9889	0.9477	0.9481	0.9474	0.8698	0.8709	0.2560
<b>lr</b>	Logistic Regression	0.9476	0.9834	0.9476	0.9475	0.9465	0.8659	0.8681	2.3240
<b>ridge</b>	Ridge Classifier	0.9477	0.9751	0.9477	0.9484	0.9466	0.8660	0.8692	0.1880
<b>svm</b>	SVM - Linear Kernel	0.8769	0.9749	0.8769	0.9005	0.8743	0.6968	0.7252	0.1860
<b>lda</b>	Linear Discriminant Analysis	0.9477	0.9710	0.9477	0.9484	0.9466	0.8660	0.8692	0.1780
<b>ada</b>	Ada Boost Classifier	0.9440	0.9674	0.9440	0.9451	0.9436	0.8603	0.8626	0.2520
<b>nb</b>	Naive Bayes	0.9253	0.9402	0.9253	0.9279	0.9239	0.8105	0.8162	0.1840
<b>qda</b>	Quadratic Discriminant Analysis	0.8807	0.9307	0.8807	0.8862	0.8692	0.6645	0.6914	0.1800
<b>dt</b>	Decision Tree Classifier	0.9402	0.9260	0.9402	0.9425	0.9399	0.8511	0.8543	0.1900
<b>knn</b>	K Neighbors Classifier	0.8693	0.9095	0.8693	0.8769	0.8618	0.6502	0.6728	1.9740
<b>dummy</b>	Dummy Classifier	0.7164	0.5000	0.7164	0.5133	0.5981	0.0000	0.0000	0.1800

	<b>Accuracy</b>	<b>AUC</b>	<b>Recall</b>	<b>Prec.</b>	<b>F1</b>	<b>Kappa</b>	<b>MCC</b>
<b>Fold</b>							
<b>0</b>	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000
<b>1</b>	0.9444	0.9795	0.9444	0.9440	0.9438	0.8586	0.8596
<b>2</b>	0.9630	0.9967	0.9630	0.9630	0.9630	0.9112	0.9112
<b>3</b>	0.9434	0.9684	0.9434	0.9430	0.9428	0.8577	0.8586
<b>4</b>	0.9811	0.9956	0.9811	0.9816	0.9809	0.9526	0.9536
<b>Mean</b>	0.9664	0.9880	0.9664	0.9663	0.9661	0.9160	0.9166
<b>Std</b>	0.0218	0.0121	0.0218	0.0220	0.0220	0.0550	0.0547

Fitting 5 folds for each of 10 candidates, totalling 50 fits

Original model was better than the tuned model, hence it will be returned. NOTE: The display metrics are for the tuned model (not the original one).

Transformation Pipeline and Model Successfully Saved

```
Out[21]: (Pipeline(memory=Memory(location=None),
                    steps=[('label_encoding',
                             TransformerWrapperWithInverse(exclude=None, include=None,
                                                             transformer=LabelEncoder())),
                           ('numerical_imputer',
                             TransformerWrapper(exclude=None, include=['Age'],
                                                               transformer=SimpleImputer(add_indicator=False,
                                                               copy=True,
                                                               fill_value=None,
                                                               keep_empty_features
                                                               =False,
                                                               missing_values=nan,
                                                               strategy='mean'))...
                           LGBMClassifier(boosting_type='gbdt', class_weight=None,
                                          colsample_bytree=1.0, importance_type='split',
                                          learning_rate=0.1, max_depth=-1,
                                          min_child_samples=20, min_child_weight=0.001,
                                          min_split_gain=0.0, n_estimators=100, n_jobs=-1,
                                          num_leaves=31, objective=None, random_state=123,
                                          reg_alpha=0.0, reg_lambda=0.0, subsample=1.0,
                                          subsample_for_bin=200000, subsample_freq=0))],
                  verbose=False),
             'best_tuned_thyroid_model.pkl')
```

In [13]: models()

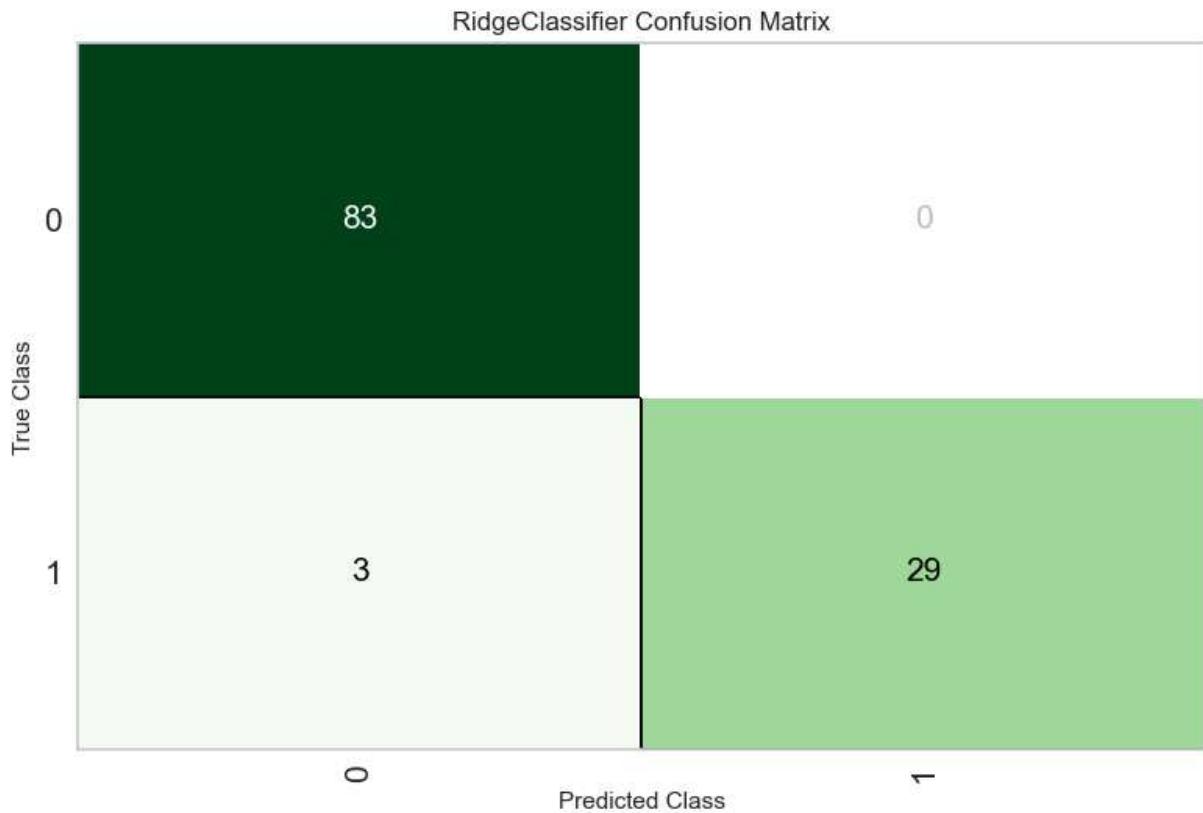
Out[13]:

ID	Name	Reference	Turbo
<b>lr</b>	Logistic Regression	sklearn.linear_model._logistic.LogisticRegression	True
<b>knn</b>	K Neighbors Classifier	sklearn.neighbors._classification.KNeighborsCl...	True
<b>nb</b>	Naive Bayes	sklearn.naive_bayes.GaussianNB	True
<b>dt</b>	Decision Tree Classifier	sklearn.tree._classes.DecisionTreeClassifier	True
<b>svm</b>	SVM - Linear Kernel	sklearn.linear_model._stochastic_gradient.SGDC...	True
<b>rbfsvm</b>	SVM - Radial Kernel	sklearn.svm._classes.SVC	False
<b>gpc</b>	Gaussian Process Classifier	sklearn.gaussian_process._gpc.GaussianProcessC...	False
<b>mlp</b>	MLP Classifier	sklearn.neural_network._multilayer_perceptron....	False
<b>ridge</b>	Ridge Classifier	sklearn.linear_model._ridge.RidgeClassifier	True
<b>rf</b>	Random Forest Classifier	sklearn.ensemble._forest.RandomForestClassifier	True
<b>qda</b>	Quadratic Discriminant Analysis	sklearn.discriminant_analysis.QuadraticDiscrim...	True
<b>ada</b>	Ada Boost Classifier	sklearn.ensemble._weight_boosting.AdaBoostClas...	True
<b>gbc</b>	Gradient Boosting Classifier	sklearn.ensemble._gb.GradientBoostingClassifier	True
<b>lda</b>	Linear Discriminant Analysis	sklearn.discriminant_analysis.LinearDiscrimina...	True
<b>et</b>	Extra Trees Classifier	sklearn.ensemble._forest.ExtraTreesClassifier	True
<b>lightgbm</b>	Light Gradient Boosting Machine	lightgbm.sklearn.LGBMClassifier	True
<b>dummy</b>	Dummy Classifier	sklearn.dummy.DummyClassifier	True

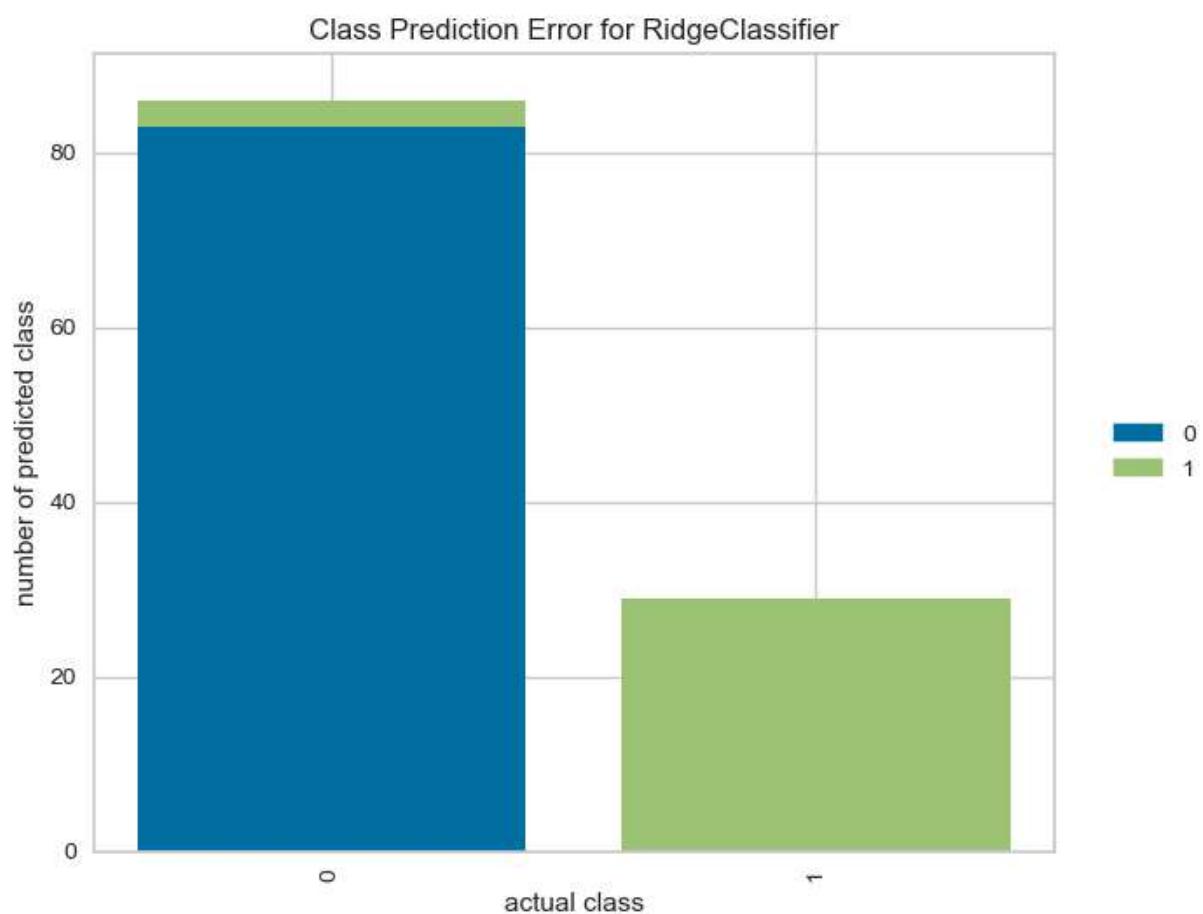
In [14]: `print(model_ridge_classifier)`

```
RidgeClassifier(alpha=1.0, class_weight=None, copy_X=True, fit_intercept=True,
               max_iter=None, positive=False, random_state=123, solver='auto',
               tol=0.0001)
```

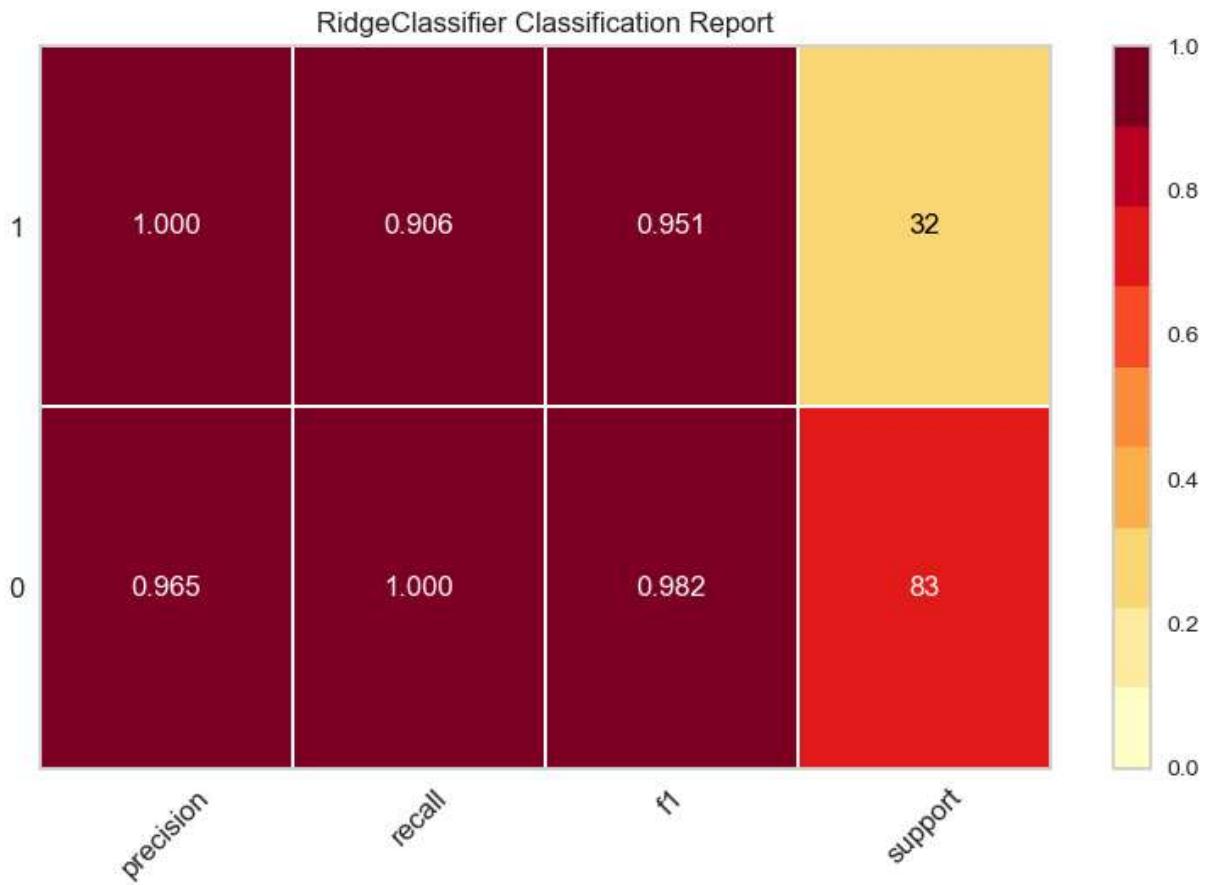
In [15]: `# Confusion matrix`  
`plot_model(tune_ridge_classifier, plot='confusion_matrix')`



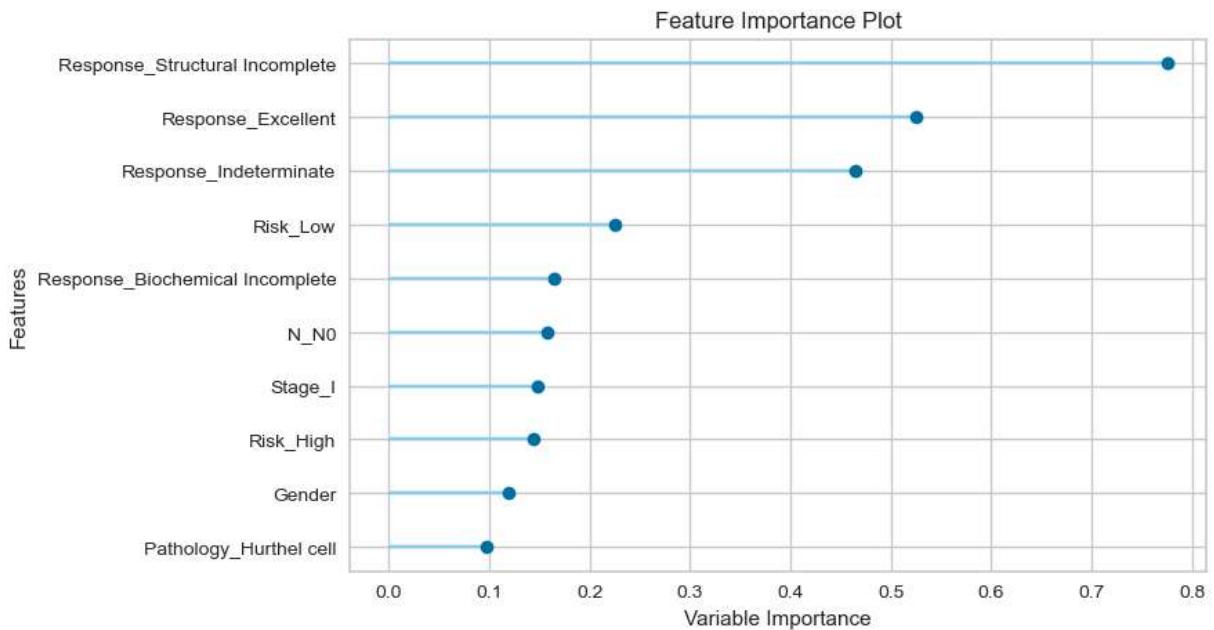
```
In [16]: # Classification error  
plot_model(tune_ridge_classifier, plot='error')
```



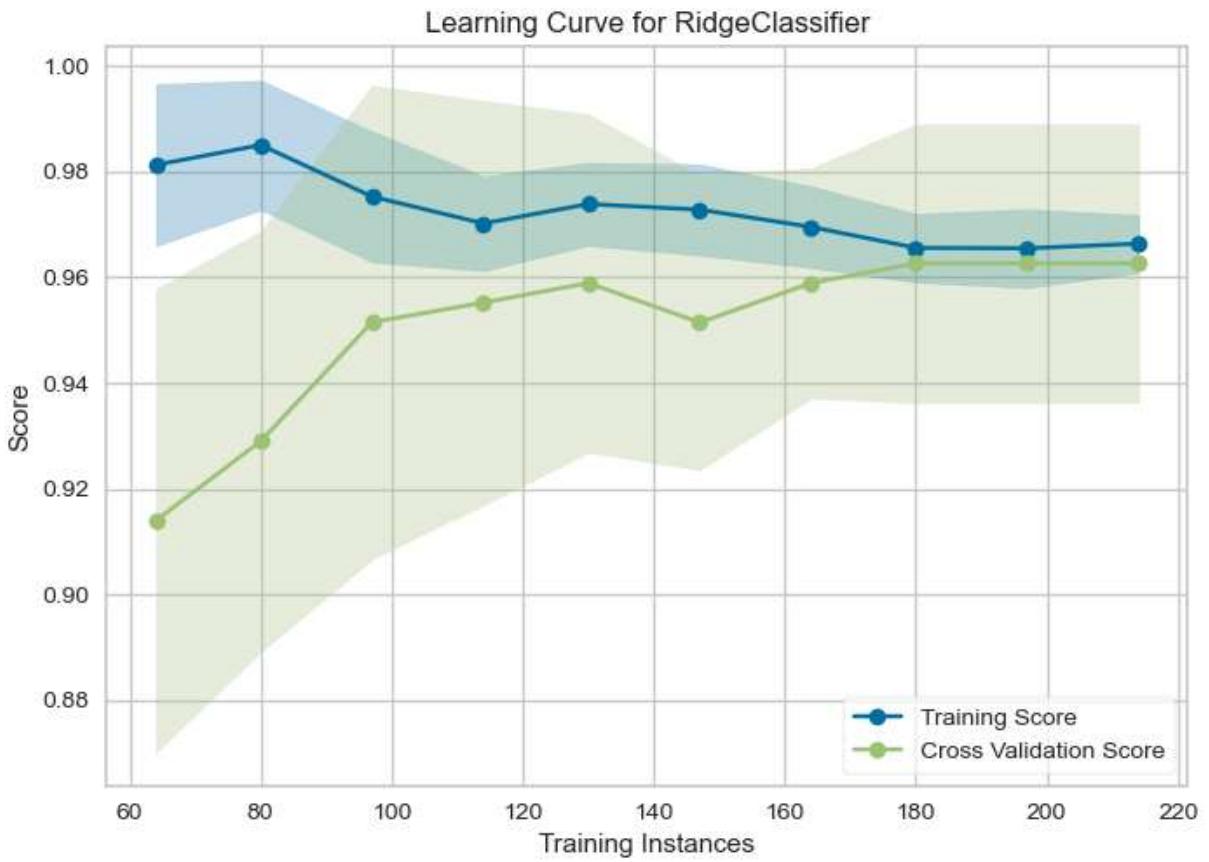
```
In [17]: # Classification report (precision, recall, F1, etc.)  
plot_model(tune_ridge_classifier, plot='class_report')
```



```
In [18]: # Feature importance  
plot_model(tune_ridge_classifier, plot='feature')
```



```
In [19]: # Learning curve  
plot_model(tune_ridge_classifier, plot='learning')
```



```
In [20]: evaluate_model(tune_ridge_classifier)
```

```
interactive(children=(ToggleButtons(description='Plot Type:', icons=''), options= ('Pipeline Plot', 'pipelin...
```

```
In [ ]:
```